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# **Analyzing Brexit: Implications for the Electricity System of Great Britain**

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Abstract: The UK's exit from the European Union (EU) has potential ramifications for the country's electricity sector, given its increasing interlinkage with other EU electricity systems. Brexit could hamper the development toward higher market integration and the realization of new interconnector projects. Moreover, a fall in the value of the Pound, resulting from Brexit in the medium term, could also affect the electricity trading structure. Combining a European electricity market model and a multi-criteria decision analysis tool, this study assesses the implications of Brexit for the electricity market of Great Britain (hereafter GB) for 2030, from the perspective of (i) political decision makers, (ii) electricity consumers, and (iii) producers. Results indicate that the implications of Brexit depend on the future development of the GB electricity system and on the objectives of the respective stakeholders. Possible opportunities brought by Brexit under a low-carbon trajectory contrast with greater challenges and tradeoffs between stakeholders under alternative power system development paths. Despite increased British autonomy in energy and climate matters, there remains interdependency between British and EU energy policy.

**Keywords:** Brexit; electricity markets; interconnectors; TOPSIS

### 1. Introduction

Britain's vote in June 2016 to leave the European Union (EU) represents a dramatic shift in the country's relationship with the rest of the EU, following over 40 years in the Union. There have been efforts to evaluate the impact of Brexit across various sectors of the British economy and society [1–5]. However, the potential results of a reduction in integration with the EU for these sectors, following Brexit, remain unclear.

One sector that could be influenced by Brexit is the Great Britain (GB) (In this paper, we only consider the power system of England, Scotland, and Wales, and not Northern Ireland. Northern Ireland and the Republic of Ireland share one electricity grid and engage in a single electricity market. Analyzing the impact of reduced expansion of interconnector capacity between the EU and GB on the joint Irish electricity system is beyond the scope of this paper) electricity sector, as it is a system that has experienced increasing integration with other electricity systems in the EU. Indeed, there has been attention in the literature on the implications of reduced integration for UK energy policy and security [6,7]: This integration has come e.g., in the form of interconnectors linking GB's electricity grid with those of neighboring EU countries, enabling the trading of electricity. The European Commission has set a target for net transfer capacities to equal 10% of all installed capacity by 2020 and 15% by 2030 [8], which means that the trade of electricity through interconnectors will form an important component in the ongoing transformation of the British and other EU member states' power systems

Energies **2019**, 12, 3212 2 of 27

toward 2030. Brexit could affect the integration of GB's electricity system into those of other EU member states and may have consequences for its future trajectory. This paper seeks to investigate the possible consequences of Brexit for the electricity system of GB.

The Brexit vote coincides with wider changes in the GB electricity system [9]. In relation to the composition of the electricity mix, there has been a shift away from coal: GB's aging coal power plant fleet will be phased out by the mid-2020s [10], and this has also been accompanied by a greater focus on combined cycle gas turbine (CCGT) plants, renewable energy sources (RES), nuclear power and imports of power through interconnectors [11]. Further interconnector projects with France and Ireland in addition to new interconnector projects with Belgium, Norway, Denmark [12], and even Iceland [13] are being developed. There is currently 4 GW of interconnection capacity installed between GB and its neighbors; a further 7 GW are planned between now and 2022 [12]. A total of 12 GW of interconnection capacity would amount to just over 12% of total generation capacity in relation to a total installed generation capacity of 81 GW at the end of 2017 [14]. This would represent progression toward the EU target for interconnection. Uncertainty caused by Brexit could raise doubts about these projects, and ultimately lead to lower interconnector capacity than anticipated. At the same time, the Brexit vote entailed a decline of the British Pound and it has never regained its pre-Brexit level [15].

Applying the Electricity Market Model for Europe (EMME), we address the question of the effects of Brexit on the GB electricity system by analyzing the implications of a reduced expansion of net transfer capacity (NTC) between the GB grid and EU neighboring grids combined with a fall in the value of the Pound. In this paper, the level of NTC is not modeled in the optimization; it is assumed that the NTC expansion is politically determined, and therefore exogenous to the model.

The model results will be categorized with respect to three different stakeholders within the electricity system: (i) political decision makers, (ii) electricity producers, and (iii) electricity consumers. The results for these three groups depend on the transition pathway that the GB electricity sector will follow up to 2030; therefore, we consider and discuss different conceivable pathways in our analysis. Applying the multi-criteria decision analysis (MCDA) tool TOPSIS (Technique for Order Preference by Similarity to Ideal Solution), the modeling results will be evaluated from the viewpoint of the three respective interest groups. MCDA techniques can complement other analytical tools. Both energy market modeling frameworks and MCDA methods are commonly used for scenario assessment and energy planning (see e.g., [16–18]). McKenna et al. [19] introduced an integrated participatory approach for energy generation scenarios for small communities. In doing so, they coupled an MCDA with a linear optimization modeling framework. Ribeiro et al. [20] electricity production scenarios for Portugal by ranking different development paths generated with an optimization model with an MCDA. Ram et al. [21] combined MCDA with scenario planning, as MCDA can be used to handle competing objectives whereas, unlike MCDA, scenario planning tools are effective at dealing with uncertainty. Energy scenario tools may offer a higher, more general level of analysis, aimed at strategic guidance and can be combined with MCDA approaches which, on the other hand, provide a detailed evaluation of a particular alternative from the perspective of different actors [22]. Furthermore, MCDA can act as an interface between complex models and practical decision problems. For instance, in evaluating future trajectories for the Portuguese power system, Ribeiro, Ferreira, and Araújo [20] coupled scenarios resulting from a mixed integer linear programming model with an MCDA tool. In doing so, they identified a willingness to increase the overall cost of the system among all the respondents as long as factors other than economic ones are integrated into the decision; MCDA, in this case, has revealed the importance of non-economic factors to the decision problem.

Furthermore, MCDA tools offer an alternative to the monetary evaluation of environmental aspects, as e.g., within the cost–benefit analysis or integrated assessment models (IAM) (see Ürge-Vorsatz, et al. [23]). Scholten et al. [24] compared the application of MCDA tools to IAM in the context of water supply planning. They found that both methodological frameworks identified the identical best alternative, even though the rankings among the alternatives differed.

Energies **2019**, 12, 3212 3 of 27

Saarikoski et al. [25] and Browne and Ryan [26] compared CBA (Cost-Benefit Analysis) and MCDA approaches.

The structure of this paper is organized as follows: Section 2 will briefly present the related scientific literature dealing with possible implications of Brexit on the GB energy system. Particular focus will be paid to the structure of the GB energy landscape and its interlinkages with the European Internal Energy Market (IEM). In this context, the impact of international NTCs is highlighted. Furthermore, the literature review considers possible consequences of Brexit with regard to future British climate and energy policy. Subsequently, in Section 3, the methodological framework is presented, detailing the specifications of the dispatch model, the scenarios used in this analysis, and the indicators forming the basis of the assessment. The modeling results for the three main actors mentioned above and the MCDA results will be presented in Section 4. Finally, policy-oriented conclusions will be drawn in Section 5. The analysis presented here is differentiated according to three possible trajectories for the UK power system to 2030, based on scenarios from ENTSO-E (European Network of Transmission System Operators for Electricity) [11]: (i) a Blue and Yellow scenario that is closer to the status quo, and (ii) a Green scenario reflecting policy with a stronger focus on the development of renewable energy capacities.

# 2. The Trend toward Greater Integration of the EU's Electricity Systems

# 2.1. Interconnectors: Role in Single Electricity Markets

Increasingly, the European Union has been promoting a common energy policy, based on the three core priorities of energy security, affordability, and environmental protection [27]. The IEM underlies this common energy policy with the aim that electricity can be traded among EU member states without restrictions, creating a more competitive electricity market, benefiting from greater security of supply [28–31], which is made possible through physical interconnections between electricity grids [32].

With regard to the multifarious challenges of security of electricity supply, the role of interconnectors has been discussed frequently [9,33]. Historically, cross-border interconnectors were intended to reinforce the security of supply, as other countries could help, via electricity exports, if a member state's electricity system faced disruption [34]. However, increasingly, there has been a focus on how interconnection can increase trade among member states and, in doing so, improve the efficiency of markets and, ultimately, social welfare [35]. A crucial argument in favor of interconnectors is that they further the integration of intermittent power resources [35–37], as they can help to reduce the impact of renewable penetration on price volatility [38,39].

Connecting two electricity markets generally results in greater price convergence between the two markets and lower price volatility [39]. This, in effect, creates both "winners and losers", as the fall in electricity prices in one region, thanks to the interconnectors, penalizes electricity producers, whereas the rise in electricity prices in the other region adversely affects electricity consumers in that region [34,40]. Furthermore, interconnectors bring about technical benefits to the power system [41], especially in situations of high renewable power penetration [42,43]. For instance, Lynch, et al. [44] argue that interconnection reduces total system costs in cases where there is high wind penetration, as the total capacities of both conventional and renewable generation technologies can be reduced.

There has already been an ongoing discussion of the impact of interconnection on electricity markets and  $CO_2$  emissions [45–47]. Mezősi, Pató, and Szabó [38] argued that the effect of interconnectors on  $CO_2$  emissions is dependent on the carbon price and the ratio between coal prices and gas prices. Indeed, they found that complying with the 10% interconnection target by 2020 would lead to an increase in  $CO_2$  emissions, if the low  $CO_2$  price environment prevailed. Considering the UK's decision to introduce a  $CO_2$  price floor (CPF), the question arises as to how a unilateral minimum price for carbon emissions interacts with electricity trade and the European Emission Trading Scheme (ETS). The CPF constitutes a minimum price for  $CO_2$  emissions, with a fixed development corridor over the next years. It consists of two parts (see Hirst [48]): (i) the ETS certificate price and (ii) the Carbon

Energies **2019**, 12, 3212 4 of 27

Support Price. The Carbon Support Price tops up the ETS price in accordance with the CPF corridor set by the British Government. In 2030, the CPF was originally set to be £70 [48].

Investments in interconnectors are long-term in nature (between 25–40 years), increasing the significance of uncertainty for the planning process of new projects [49]. Moreover, in general, there are insufficient incentives to invest in interconnectors, compounded by resistance from domestic firms to more competition in the domestic market [34,47]. Furthermore, investment in interconnectors is complicated by long planning and construction times, onerous permit procedures, and the possibility that interconnector capacities will be under-used and, therefore, become stranded [34]. Barriers to investment in interconnectors are exacerbated by the presence of a "regulatory gap" in which there is no regulator responsible for cross-border infrastructures [40]. The uncertainty arising from Brexit could increase the reluctance to invest in further interconnector projects. Therefore, this work offers a detailed analysis of possible implications of a reduced interconnection development for the GB power system.

## 2.2. Brexit: Implications for the GB Electricity System

It remains unclear what the UK's trading relationship with the EU will look like after Brexit. At the time of the Withdrawal Agreement, the will was expressed to develop a "framework" that would enable "technical cooperation between electricity and gas network operators" and ensure the security of supply and efficient trade over interconnectors; however, the British Parliament subsequently rejected the agreement [50]. If the Withdrawal Agreement is revived, it is not clear to what extent such a framework would emulate the conditions the UK enjoyed as a member of the IEM. Energy UK [51] emphasizes the importance of minimizing uncertainty, given the "long investment timelines" concerning energy assets, such as interconnectors.

Brexit has led to a decline in the value of the Pound, and this should be factored into any (medium-term) economic analysis of Brexit. In the weeks following the referendum, the Pound's value fell substantially, losing almost 15% and, in the months to October 2016, the nominal effective exchange rate of the Pound, measured against a selection of foreign currencies, depreciated by nearly 20% [52]. Since 48% of the UK's exports go to the European Single Market [53], the impact of restrictions and greater costs in trading with the EU would be considerable. There are indications that the Pound's loss in value could be sustained, as both trade between the UK and the EU in addition to foreign direct investment (FDI) are likely to be adversely affected by Brexit [54–56]. By 2030, Ebell and Warren [55] estimated that if the UK follows a Swiss arrangement with the EU, trade between the UK and the EU will be between 13–17% lower than it would have been, had the UK remained in the EU and, if the UK has a World Trade Organization (WTO) arrangement, then trade is estimated to be between 20–30% lower. These conditions would, in the long-term, depress the value of the Pound. A lower value of the Pound is likely to make imports of electricity more expensive, and conversely increase the competitiveness of GB electricity exports to the EU.

Both climate and energy policies have substantial links with EU policy, and could be affected by lower integration with the EU. In terms of climate policy, the UK's current CPF is contingent on the EU ETS—it tops up, essentially, the emission certificate price, and if the UK were to be out of the EU ETS, it would need to replace this with a UK ETS or a carbon tax [4]. Taking a more optimistic view of Brexit, Pollitt [5] suggested that the impacts on the GB energy sector will be rather contained, and that leaving the EU could bring opportunities for the sector. He refers to the fact that fossil fuel prices are set internationally, whilst the UK has diversified its suppliers of fossil fuels in addition to the interconnection capacities being currently low, as indicative of the limited negative impacts of Brexit. Outside of the EU, the UK would have greater freedom to pursue alternative options in market design that are inconsistent with EU policy, such as reconsidering the unbundling of generation from supply and the financing of nuclear generation over long-term contracts between generators and suppliers [5].

As regards the risks of Brexit, energy security is a concern. This is because there may be less cooperation in terms of exchanging electricity through the interconnectors. Furthermore, "increased coordination costs" between the UK and EU countries could lead to delays in interconnector investments

Energies **2019**, 12, 3212 5 of 27

and reduced NTC growth [5]. This would force GB to resort to more expensive domestic back-up capacity in order to meet security of supply needs. A reduced expansion of interconnectivity could weaken the taming effect that interconnection has on very high peak power prices and lead to potential revenue opportunities for operators of large-scale storage facilities; conditions for such facilities in Great Britain have been less attractive than expected in recent years [57]. Indeed, Pye, Mathieu, and Deane [58] warned that the uncertainty associated with Brexit may lead to nervousness on the part of investors as regards investing in interconnectors, with French regulators already re-examining the rationale behind a new French–British interconnector project.

In summary, a greater integration of EU electricity systems is considered advantageous for energy security, the competitiveness of power markets, and should lead, ultimately, to lower system costs. Trajectories of the UK's power system toward 2030 have taken into account greater levels of cross-border interconnection (see e.g., [11,59]), and since Brexit raises questions about future interconnection projects, the possible impacts of lower interconnection must be examined in the course of the evaluation of Brexit's impacts.

# 3. Research Approach

## 3.1. Model Specifications

The EMME is a dispatch model of the EU electricity market. This model, featuring 28 individual member states of the EU and Switzerland, is a linear optimization model. Power flows among member states are restricted by NTCs between the respective markets. The model features a full list of types of generation technologies, their respective capacities and vintage structures, as well as differentiated variable costs in each country. The generation capacities, energy carrier prices, CO<sub>2</sub> certificate prices and the power demand are exogenous model inputs and demand is given in hourly intervals. The availability of variable and non-dispatchable RES is determined by time series.

The model takes the following objective function as its basis:

$$\min Z = \sum_{h,i,d} [Pr(h,i,d) \cdot Cst(i,d)] + \sum_{h,d,k} Im(h,d,k) \cdot T$$
(1)

subject to:

$$\sum_{h,i,d} Pr(h,i,d) + \sum_{h,d,k} Im(h,d,k) = Dm(h,d)$$
(2)

$$Pr(h, i, d) \le Cp(h, i, d)$$
 (3)

$$Im(h,d,k) \le NTC(d,k) \tag{4}$$

with:

- i: generation technology type
- *h*: specific hour of the year [-]
- *T*: transport costs for imports and exports [€/MWh]
- *d* and *k*: country indexes [-]
- Cst: variable generation costs [€/MWh]
- *Pr*: electricity production [MWh]
- *Cp*: generation capacity [MW]
- Im: electricity imports from country k to country d [MWh]
- Dm: electricity demand [MWh]
- NTC: net transfer capacity between two markets [MW]

Energies **2019**, 12, 3212 6 of 27

Producers' surpluses are calculated as the spread between the wholesale price and variable electricity generation costs:

$$PS(i,d) = \sum_{h} ((X(h,d) - Cst(i,d)) \cdot Pr(h,i,d))$$
(5)

with

- *PS*: producer surplus [€]
- X: wholesale price [€/MWh]

Changes in consumers' surpluses are expressed as the absolute change in average wholesale prices multiplied by electricity consumption. Prices in the Yellow scenario are used as a reference value when calculating the magnitude of differences in wholesale prices.

$$CS(d) = \left(X_{Ref}(d) - \frac{1}{8760} \cdot \sum_{h} X(h, d)\right) \cdot DM \tag{6}$$

with

- *CS*: change in consumers' surplus [€/a]
- $X_{Ref}$ : average reference wholesale price of Yellow\_100\_0 [ $\epsilon$ /MWh]
- *DM*<sub>Ref</sub>: overall electricity demand [MWh/a]

In a subsequent step, the model results will be used to determine the overall levelized costs of electricity (LCOE) of the UK generation fleet *avg\_LCOE* with:

$$avg\_LCOE = \frac{\left(\sum_{i} ANF_{n,p} \cdot \frac{I}{t_{voll}} + \frac{O\&M_{fix}}{t_{voll}} + O\&M_{var} + FC + EC\right) \cdot Pr(h, i, d)}{\sum_{i} Pr(h, i, d)}$$
(7)

and:

$$ANF_{n,p} = \frac{(1+p)^n \cdot p}{(1+p)^n - 1} \tag{8}$$

with:

- I: overnight investment costs [€/MW]
- *ANF*: annuity [€]
- $O\&M_{fix}$ : fixed operation and maintenance costs [€/MW]
- O&M<sub>var</sub>: variable operation and maintenance costs [€/MWh]
- t<sub>voll</sub>: achieved full load hours [h]
- *FC*: fuel costs [€/MWh]
- *EC*: costs for emission certificates [€/MWh]
- *p*: interest rate [-]
- *n*: expected lifetime of power plant [a]

With respect to the electricity production of each power plant type, the respective environmental implications are calculated by:

$$Env_{x}(d) = \sum_{i} \left( LCA_{x,i} \cdot \sum_{h} Pr(h,i,d) \right)$$
(9)

with

Energies **2019**, 12, 3212 7 of 27

- $Env_x$ : environmental impact x
- $LCA_{x,i}$ : specific environmental impact factor x for power plant type i

The calculation of environmental impacts is based on LCA (Life Cycle Analysis) data for power plant types in the UK provided by "probas", which is a database for environmental management systems made available by the German Federal Ministry for the Environment, Nature Conservation, and Nuclear Safety.

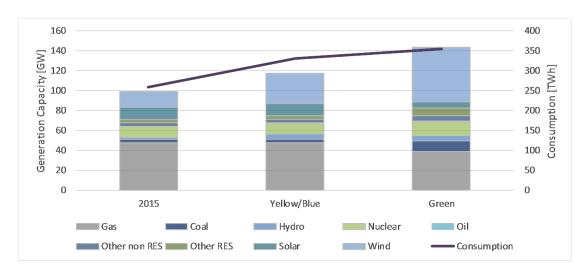
#### 3.2. Scenarios

The modeling approach in the paper studies three different scenarios for the power system in 2030, namely: Yellow, Blue, and Green. The scenarios are based on the capacity, CO<sub>2</sub> price, and fuel cost assumptions contained within European Transmission System Operators' Ten-Year Network Development Plan (TYNDP) data [11] (the data refer to 'Vision 1' and 'Vision 3', respectively). Adjustments were made to integrate the UK's goal for a complete coal phase out by 2022. The Climate Change Act (2008) and the introduction of the Carbon Price Floor are examples of the UK unilaterally adopting strong climate policies and indicate a "Green" trajectory. However, the UK would no longer participate in the EU ETS, with which the Carbon Price Floor is linked, and would no longer be subject to the EU Renewables Directive. Thus, the UK would have to replace these instruments [4]. This gives rise to the opportunity to deviate from the Green trajectory, and possible deviations are characterized by the Yellow and Blue scenarios.

The Yellow scenario is characterized by a lower carbon price of  $\leqslant 30$  per ton of  $\mathrm{CO}_2$ , and a moderate expansion of renewable power. As discussed before, Britain introduced a minimum price for emissions certificates in the form of a CPF. In general, the Yellow and Blue scenarios show the same characteristics with the exception of greenhouse gas (GHG) certificate prices in the UK. Yellow assumes a weaker enforcement of British environmental policy, which results in a CPF lower than (or equal to) the overall EU market price (i.e., the EU ETS price will apply for British generators). Blue represents a stronger commitment toward national climate mitigation policies, and thus shows a CPF well above the EU market price for  $\mathrm{CO}_2$  emissions.

In the Green scenario, amid the higher price for emission certificates, there is a stronger deployment of low-carbon electricity generation units.

Figure 1 shows the exogenous input parameters of overall installed capacity in GW (left axis) and electricity demand in TWh (right axis) of the Yellow, Blue, and Green scenarios in 2030 compared with 2015 for the UK (based on [49], and [11]).



**Figure 1.** Assumptions on changes in installed capacity. Source: Own compilation based on [60] and [11].

Energies **2019**, 12, 3212 8 of 27

Data relating to wind and PV profiles and capacity factors of different generation technologies were taken from internal data in the IEK-STE (Institute of Energy and Climate Research– Systems Analysis and Technology Evaluation). Information about the net transfer capacity of Great Britain was found in OFGEM (Office of Gas and Electricity Markets) data and ENTSO-E (European Network of Transmission System Operators of Electricity) sources.

As discussed in Section 1, this paper includes a differentiated analysis of a decreased development of the NTCs for the UK. While the already existing NTCs of 4 GW will be held constant, the 7 GW of NTCs planned for the future will be reduced by 50% or 35%, respectively (the capacity evaluation underlying the Green scenario did not allow for a reduction of NTC expansion by 50%. Section 4 will provide further details). Furthermore, the risk of a devaluation of the British Pound resulting from Brexit will be analyzed by introducing a devaluation of the Pound by 10%. This results in four different variations for each scenario, as depicted in Table 1.

Scenario Variation		NTC Expansion			
Scenario variation	S	As Planned	Reduced Increase		
Devaluation of Currency	No	Yellow_100_0 Blue_100_0 Green_100_0	Yellow_50_0 Blue_50_0 Green_65_0		
Devaluation of Currency –	Yes	Yellow_100_10 Blue_100_10 Green_100_10	Yellow_50_10 Blue_50_10 Green_65_10		

**Table 1.** Scenario variations. Source: Own compilation.

Table 2 shows the key parameters for the scenarios applied.

	Yel	low	В	lue	Green		
Factor	Yellow_100_0 Yellow_65_10 Yellow_65_0 Yellow_100_10		Blue_100_0	Blue_100_0 Blue_100_0		Green_100_10	
			Blue_65_0 Blue_65_0		Green_65_0	Green_65_10	
Certificate price [Euro/t CO <sub>2</sub> ]	UK: 30, EU: 30		UK: 90, EU: 30		UK: 90, EU: 90		
Electricity demand UK [TWh]	330		330		355		
Fuel prices:							
Oil [Euro/GJ]:	57.87		57.87		41.73		
Coal [Euro/GJ]	12.53		12.53		7.96		
Gas [Euro/GJ]	37.00	39.54	37.00	39.54	28.48	30.44	
Uranium [Euro/GI]	1.36	1.42	1.36	1.36 1.42		1.42	

Table 2. Key parameters for the scenarios. Source: Own compilation, based on ENTSO-E [11].

## 3.3. System Performance and Evaluation of Modeling Results

MCDA approaches are commonly used for evaluating energy policies, scenarios, and technologies [61–63]. For the evaluation of the presented scenarios, TOPSIS will be applied, which is a variant of MCDA that has been used frequently in energy planning and analysis (see e.g., [64–66]) (in order to provide insights on the robustness of the MCDA results with regard to the chosen MCDA tool, Appendix C offers a system performance analysis using the Preference Ranking Organization Method for Enrichment Evaluation (PROMETHEE) method). MCDA offers a systematic approach to complex decision problems involving multiple perspectives, uncertainty, and competing goals [67]. Firstly, in the system performance assessment, each scenario is evaluated against a set of indicators, and the scores on each indicator are normalized to enable comparability. Secondly, an evaluation matrix (Table 3) is constructed, given the importance attached by each actor to each indicator. The weightings were derived from the judgment of the researchers informed by an understanding of the main issues

Energies **2019**, 12, 3212 9 of 27

for stakeholders in the energy sector. The actor-specific utility function results from the sum of actor-specific weightings for the indicators, which are given in Table 3, and the calculation of which is detailed in Equations (10)–(12), multiplied by the scores for each scenario on each indicator. As shown in greater detail in Appendix A, based on this utility function, it is possible to calculate, from the perspective of the particular actor, how far a scenario is from a positive ideal scenario and a negative ideal scenario. The highest scoring scenario (given by Equation (A5) in Appendix A) gives the highest utility for the particular actor, under the assumption that utility rises or falls at a constant level, according to changes in performance of a scenario on a given indicator [68].

**Table 3.** Relevance of parameters for each actor. Source: Own compilation. LCOE: levelized costs of electricity.

	Criteria Evaluation of Actor:						
Indicator	Political Decision Makers  Electricity Produce		<b>Electricity Consumers</b>				
Net imports	3	2	1				
Consumer surplus	2	1	3				
Avg. wholesale price	1	2	3				
LCOE	1	3	2				
Avg. price spread	2	3	1				
CO <sub>2</sub> emissions	3	1	2				
Environmental impacts	2	1	3				
Share of dom. production	3	2	1				
Producer surplus	2	3	1				

The system performance assessment will rely on the following indicators:

- (1) Net imports
- (2) Consumers' surplus
- (3) Average wholesale price
- (4) LCOE
- (5) Average price spread
- (6) CO<sub>2</sub> emissions
- (7) Environmental impacts
- (8) Share of domestic production
- (9) Producers' surplus

TOPSIS evaluates the distance of a given alternative to a hypothetical positive-ideal and negative-ideal alternative [69] (see Appendix A for detailed information on the approach used to evaluate the scenarios). The TOPSIS tool has a wide range of applications [70]. TOPSIS is used in the main article, as it is a well-known technique [70] that is simple and considered more familiar to practitioners than the PROMETHEE method [71]. However, the PROMETHEE method is used in Appendix C to validate the results of the TOPSIS analysis. In the field of energy research, TOPSIS has been used, among other things, to evaluate energy and environmental policy [72], optimal sites for thermal power plants [73], electric vehicle charging stations [74,75], and for rating renewable energy sources [76].

In order to depict the objectives of different actors within the electricity system, we initially attached preferences for each actor in relation to each parameter. Table 3 shows the evaluation matrix, with a ranking of each actor group's interest in the respective MCDA criteria (the values provided in Table 3 are indicative and serve the purpose of illustration. Several studies provide a discussion on deriving such evaluations through applying other tools (e.g., the analytic hierarchy process) for certain stakeholder groups or panels (see e.g., [77–79])). The value 3 indicates highest interest of the stakeholder group in the specific criteria, whereas 1 indicates lowest interest.

Energies **2019**, 12, 3212 10 of 27

From these preferences, five different weighting factors for each parameter and for each actor within the MCDA were derived: (i) 'Equal weights', (ii) 'Political decision makers' objective', (iii) Producers' objective, (iv) Consumers' objective, and (v) Average actors' weights. 'Equal weights' will assign a weighting factor of:

$$w_i = \frac{1}{m} \tag{10}$$

with  $w_i$ : the weighting factor for indicator i, and m: the number of indicators to every criterion. Due to the uncertainty of the assigned interests of the respective stakeholders, this weighting provides a benchmark to the other weighting factors.

The weighting approaches 'Political decision makers' objective', 'Producers' objective', and 'Consumers' objective' reflect the individual ranking of each stakeholder group. The weighting factors are calculated by dividing the criteria evaluation i of each actor a, as depicted, by the sum of actor-specific criteria evaluations:

$$w_i^a = \frac{r_i^a}{\sum_{i=1}^m r_i^a} \tag{11}$$

For 'average actors' weights', we sum up the stakeholder-specific weightings and divide the sum by the number of actors:

$$\overline{w}_i = \sum_{a}^{k} \left( \frac{r_i^a}{\sum_{i=1}^{m} r_i^a} \right) / k \tag{12}$$

where *k*: the number of actors.

#### 4. Results and Discussion

## 4.1. Computation of the Indicators

In 2030, the modeling results show that the volume of production and the relative share of different generation sources vary according to the scenario pursued and its respective variants regarding the NTC development and exchange rates (see Figure 2) (the dimensions "Full NTC" and "Reduced NTC" refer to an NTC expansion as planned, or a reduced NTC development, respectively. "Default" describes a default exchange rate, and "Devaluated" describes a devaluated currency. These abbreviations will be used throughout the text).

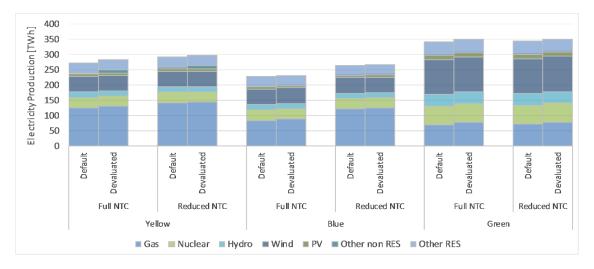
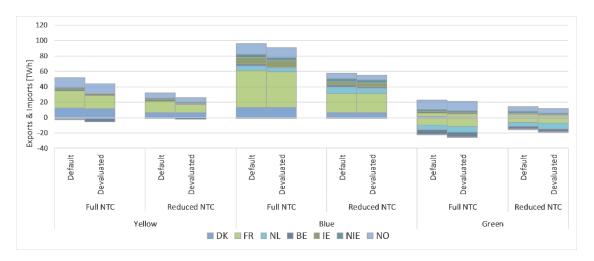


Figure 2. 2030 Power mix across scenarios and variants. Source: Own calculation.

In 2015, net imports of electricity accounted for 5.7% of domestic consumption (own calculation, based on [60] and [80]). For the Yellow scenario, this proportion increases to 15.5% in 2030. Due to the asymmetric financial burden for GB generators in the Blue scenario, net imports rise drastically,

Energies **2019**, 12, 3212 11 of 27

reaching 29.7% of domestic electricity consumption in 2030. Within the Green scenario, a general shift in the import–export ratio can be observed: while imports decrease drastically, exports to neighboring markets increase, leading to nearly balanced net imports. In all the scenarios under consideration, Norway plays a major role as an electricity exporter, which stems from its significant hydroelectric park and volume of hydro pump storage. For the Green and Yellow scenarios, it can be observed that there is a general shift in the import structure. In the Yellow scenario, Norway replaces the Netherlands as the second largest exporter to GB. Electricity imports from the Netherlands fall to less than 0.4% of overall imports (compared to just under 38% of all imports in 2017 [14]). In the Green scenario, Norway becomes the largest importer, followed by France; the Netherlands become a net importer of electricity from GB. Figure 3 depicts the changes in the electricity trade structure, with imports displayed on the positive ordinate axis and exports displayed on the negative ordinate axis.



**Figure 3.** Electricity trade. Source: Own calculation.

In general, for the Blue scenario, a devaluation of the Pound does not significantly affect the overall electricity imports to GB. This is mainly attributable to the CPF, which increases the relative price at which GB (fossil) generators are able to supply electricity. Hence, the further increase in variable costs for fuel acquisition only marginally influences the cost structure. The Yellow scenario shows the strongest sensitivity to a devaluation of the Pound.

Green is only moderately affected by a devaluation of the Pound. Due to the high share of variable renewable energies in the system, imports are primarily used in periods of low wind and solar availability.

Although imports in the Green scenario represent only a fraction of the imports in the Yellow and Blue scenarios, modeling results indicate that electricity transfers represent a more critical role for the Green scenario; imports are characterized by an acute need for power transfers at certain hours in the year in which variable RES are unavailable. Whilst in the Yellow and Blue scenarios, a lower expansion of future NTCs in Great Britain by 50% is possible, for the Green scenario, the maximum feasible reduction is 35%. This indicates that in the Green scenario, in order to provide a sufficient security of supply, the GB electricity system relies on either a strong interconnection with neighboring electricity markets, or a significant deployment of further flexibility options (e.g., gas power plants, electricity storages, demand side management, etc.).

An examination of CO<sub>2</sub> intensity of electricity generation, displayed in Table 4, reveals that the significant reduction in emissions in the Blue scenario, compared to the Yellow scenario, derives partly from the general decrease of domestic electricity production. The reduced expansion of NTCs means resorting to a greater use of older power plants, leading to a rapid increase in CO<sub>2</sub> emissions. Likewise, under the devalued currency, domestic generation, including that from less efficient older plants, increases, leading to greater emissions. Due to the CPF, this effect is stronger in the Blue scenario than

Energies **2019**, 12, 3212 12 of 27

in Yellow. In absolute terms, greenhouse gas emissions in the Green scenario are quite stable and show no strong interrelation with changes in NTC or exchange rate volatility.

<b>Table 4.</b> Sensitivity of CO <sub>2</sub> emissions with regard to reduced net transfer capacity (NTC) expansion
and depreciation of the Pound. Source: Own calculation.

	Expansion of NTC	as Planned	Reduced Increases in NTC Expansion		
	Default Exchange Rate	Devaluated Currency	Default Exchange Rate	Devaluated Currency	
CO <sub>2</sub> emissions Yellow	44.41 Mt	+14.28%	+15.47%	+27.43%	
CO <sub>2</sub> emissions Blue	27.04 Mt	+5.86%	+46.92%	+50.00%	
CO <sub>2</sub> emissions Green	23.42 Mt	+12.79%	+4.30%	+15.55%	
CO <sub>2</sub> intensity Yellow	163.21 g/kWh	+10.06%	+8.07%	+16.46%	
CO <sub>2</sub> intensity Blue	119.30 g/kWh	+3.71%	+25.88%	+27.52%	
CO <sub>2</sub> intensity Green	68.42 g/kWh	+10.10%	+3.88%	+12.70%	

Table 5 presents the average wholesale prices. Despite the high price of CO<sub>2</sub> certificates, the Green scenario shows the lowest market prices, which is predominantly due to the high share of renewables. Due to the low volume of overall imports, the sensitivity toward changes in the NTC development is marginal. In fact, wholesale prices show a small decrease as a result of reduced NTC growth. Especially in the Blue scenario, a decreased expansion of NTC leads to significant price increases. Due to the additional financial stress for British generators, which is caused by the higher carbon price level, imports represent a considerable share of the electricity supply. This is due to the loss of competitiveness of British fossil fuel generators. Consequently, a decrease in NTC growth necessitates a stronger integration of older, less efficient gas power plants. The high price for CO<sub>2</sub> emissions has a greater effect on those power plants, leading to the higher increase in the average wholesale price. For all the scenarios, the effect of the depreciation of the British Pound on wholesale prices is rather strong, which can be traced back to two main drivers. Overall expenses for fuel purchase for power plants will rise, increasing the marginal costs of fossil and nuclear power plants. Furthermore, imports from neighboring states become relatively more expensive. Due to the relatively costly imports, the utilization rate of older and less efficient gas-fired power plants rises, resulting in higher wholesale prices.

Table 5. Changes in wholesale prices. Source: Own calculation.

	Expansion of NT	C as Planned	Reduced Increases in NTC Expansion		
Scenario	Default Exchange Rate	Devaluated Currency	Default Exchange Rate	Devaluated Currency	
Yellow	72.96 [€/MWh]	+6.23%	+2.81%	+8.58%	
Blue	88.14 [€/MWh]	+5.35%	+6.27%	+11.34%	
Green	67.95 [€/MWh]	+6.03%	-0.22%	+5.38%	

As shown in Figure 4, the Blue and Yellow scenarios show decreasing consumer surpluses under the reduced development of NTCs. The high share of RES in the Green scenario exerts downward pressure on the wholesale prices, and as a result leads to a slight increase in consumer surplus. Whilst rents show no significant sensitivity toward a reduced NTC development in the Green scenario, the Blue scenario is strongly affected by a reduced NTC growth. Due to the interrelation of a drop in the exchange rate and wholesale prices as described above, a depreciation of the Pound leads to a significant decrease in consumer surplus for all the scenarios.

Energies **2019**, 12, 3212 13 of 27

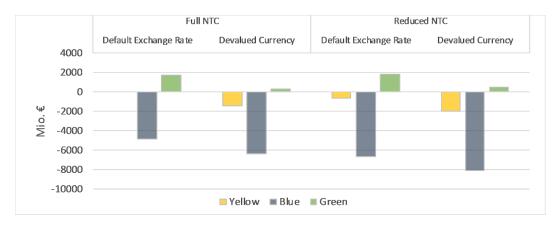
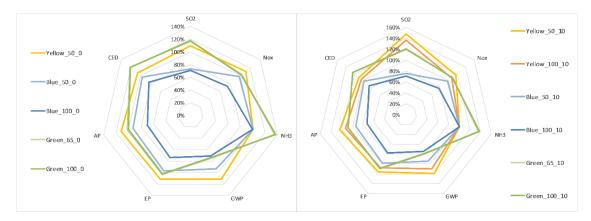


Figure 4. Changes in consumer surplus in comparison to the Yellow scenario. Source: Own calculation.

Assuming preferences as presented in Table 3, consumers will prefer electricity systems with a low environmental impact. The most notable emissions resulting from electricity generation include: sulfur dioxide ( $SO_2$ ), ammonia ( $NH_3$ ), and nitrogen oxides ( $NO_X$ ). Compound indicators also highlight environmental or climatic damage, such as global warming potential (GWP), eutrophication (EP), and acidification potential (EP). Figure 5 shows the emissions of electricity generation in the UK in comparison to the Yellow scenario, with no reduction of net transfer capacities (indexed Yellow = 100%).



**Figure 5.** Environmental impacts from electricity generation in comparison to Yellow\_100\_0. Source: Own calculation.

In the case of no reduction in the NTC development, the Blue scenario shows the strongest reduction in emissions. However, this has to be considered in the light of the overall electricity generation structure in the Blue scenario, with a significant amount of the reduced environmental impact stemming from the overall reduction in electricity generation (see Section 4.1). This can be observed in the sharp increase between Blue\_100\_0 and Blue\_50\_0. Since the electricity generation structure for the Green scenario is relatively rigid (variable RES and nuclear power plants are at the lower end of the merit order, and operate nearly on their maximum utilization rate), emissions hardly change in the case of reduced NTC growth. While the Green scenario shows a significant drop in GWP, other emissions do not drop drastically. In order to adjust the emission data to the framework of this study, the emissions presented in Figure 5 will be condensed to a single indicator 'environmental impact' by applying TOPSIS with equal weights for the depicted impacts. This will allow for the subsequent evaluation of the scenario ranking in Section 4.2.

The average price spread gives an estimation of the difference between GB's and the connected electricity markets' wholesale prices, and indirectly, of the producers' opportunities to sell electricity to foreign markets. As discussed in Section 2.1, a higher level of interconnectivity should in general lead to a smaller divergence in prices. Indeed, observing the interaction between the average price

Energies **2019**, 12, 3212 14 of 27

spread of GB and the connected markets of Norway, Denmark, France, Belgium, the Netherlands, and Northern Ireland, this assumption holds for the Yellow and Blue scenarios (see Table 6).

	Expansion of N	TC as Planned	Reduced Increases in NTC Expansion		
	Default Exchange Rate	Devaluated Currency	Default Exchange Rate	Devaluated Currency	
Yellow	13.94	13.04	16.58	15.31	
Blue	28.08	26.97	34.65	33.29	
Green	7.14	5.61	6.94	5.07	

**Table 6.** Average price spread in € per MWh. Source: Own calculation.

Prima facie, examining the changes of the price spread in the Green scenarios shows a counter-intuitive structure: a reduced expansion of NTC reduces the price spread between GB and the connected neighboring electricity markets. There are two effects that lead to this development. As discussed in the analysis of wholesale prices, Green\_65\_0 actually shows a decrease in wholesale prices in comparison to Green\_100\_0. Simultaneously, other markets in Europe show an increase in wholesale prices in case of a lower NTC development. The model results suggest that this stems from GB's function as the 'intermediary' between Norway and the markets in France, Belgium, and the Netherlands. As a result, the price spread between GB and the connected markets in continental Europe decrease in the Green scenario, in the case of a reduction in the growth of NTC.

Despite the higher import ratio of the Blue scenarios, producers' surpluses are significantly higher than in the Yellow scenarios (see Figure 6), and even more so at a devalued exchange rate (as depicted by the black bars).

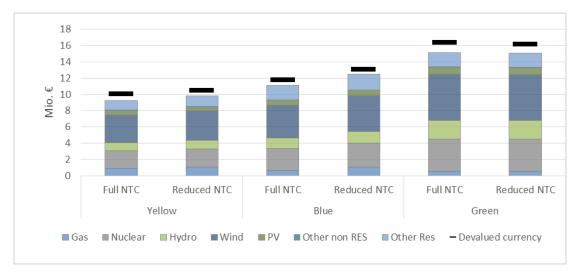


Figure 6. Interrelation of producer surplus and exchange rate volatility. Source: Own calculation.

In the Blue scenario, higher wholesale prices allow for increasing profit opportunities for low carbon generation technologies in particular. As a result, low carbon options, such as nuclear power plants, or wind farms show a significant increase in producer surpluses. On the other hand, the Green scenario reveals the highest level of producer surpluses. It should be noted that the Green scenario shows the highest level of consumer surpluses. Again, especially low-carbon generation technologies account for the major share of overall surpluses, while fossil fuel power plants do not create meaningful surpluses.

Figure 7 shows the average LCOE of the British electricity system, with the black bars indicating increases due to an additional devaluation of the exchange rate. The Yellow scenario shows the lowest LCOE for the scenarios under consideration and a moderate sensitivity toward changes in the NTCs.

Energies **2019**, 12, 3212 15 of 27

In the case of no reduction of the NTCs, the implementation of a high CO<sub>2</sub> price in the Blue scenario leads to an increase of 9.6% in the average LCOE in comparison to the Yellow scenario. However, the LCOE in the Blue scenario drastically increases in the case of a reduced development of transfer capacities. As discussed in the previous section, the CPF shifts electricity production to neighboring countries. Whilst the increase of LCOE in the Blue scenario is almost exclusively driven by higher variable costs for CO<sub>2</sub> emissions, for the Green scenario, several drivers can be determined. First of all, the increase in LCOE stems partly from the higher overnight costs for RES. However, the higher LCOE in the Green scenario in comparison to the Yellow scenario cannot be explained by higher investment costs for RES capacities alone. There are two other drivers that show increasing effects. As in the Blue scenario, a high price for CO<sub>2</sub> certificates drives up variable generation costs. However, since the overall electricity system in the Green scenario is less CO<sub>2</sub> intensive, this effect is not as strong as in the Blue scenario. The merit order effect reduces the full load hours of gas-fired power plants. As a result, the digression of fixed costs (overnight investment and operation and maintenance (O&M) costs) for generation units drops, which increases the average LCOE.

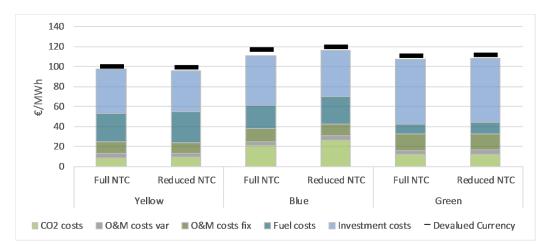


Figure 7. Average levelized costs of electricity generation. Source: Own calculation.

Within all the scenarios under consideration, a depreciation of the British Pound does not significantly impact the LCOE. Although domestic electricity production is stimulated by the depreciation of the British Pound and the average utilization rate of gas-fired power plants increases, the higher digression of fixed costs, driven by more full load hours, does not offset the higher costs for fuel supply and imported power plant components.

# 4.2. MCDA

This section will combine the results presented in the previous chapter and jointly evaluate them using the MCDA tool described in Section 3.1, with the indicators and their results listed in Table 7.

The weighting factors expressed by Equations (11)–(13) that were used for the MCDA analysis are shown in Table 8. Applying those factors, the TOPSIS method arrives at the rankings presented in Table 8.

Energies **2019**, 12, 3212 16 of 27

	<b>Table 7.</b> Compilation of multi-criteria	decision analysis	(MCDA) parameters.	Source: own compilation.
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		Expansion of NTC as Planned					Reduced Increases in NTC Expansion					
	Defa	ult Exchan	ge Rate	Deva	Devaluated Currency		Default Exchange Rate			Deva	Devaluated Currency	
	Yellow	Blue	Green	Yellow	Blue	Green	Yellow	Blue	Green	Yellow	Blue	Green
Net Imports [TWh]	49.91	95.43	0.76	39.18	90.33	-4.72	30.99	57.19	-0.61	23.98	55.11	-6.87
Con. Surplus [Mio. €]	0.00	-4883.83	1730.28	-1463.04	-6400.05	312.17	-659.13	-6662.97	1780.89	-2014.77	-8099.56	465.22
Avg. wholesale price [€/MWh]	72.96	88.14	67.95	77.50	92.85	72.05	75.00	93.67	67.81	79.22	98.13	71.61
LCOE [€/MWh]	97.10	111.25	107.32	99.51	116.22	110.75	96.11	115.83	108.49	98.74	119.51	111.27
Avg. price spread [€/MWh]	13.94	28.08	7.14	13.05	26.97	5.61	16.58	34.65	6.94	15.31	33.29	5.07
CO <sub>2</sub> [Mio t]	44.41	27.04	23.42	50.75	28.63	26.42	51.28	39.73	24.43	56.59	40.56	27.06
Environmental impact [-]	0.53	1.00	0.44	0.31	0.95	0.37	0.37	0.66	0.44	0.22	0.65	0.37
Share of dom. Production [%]	0.85	0.70	0.99	0.88	0.72	1.01	0.90	0.82	0.99	0.93	0.83	1.01
Prod surplus [Mio. €]	9.27	11.13	15.17	10.01	11.77	16.35	9.81	12.52	15.10	10.47	13.05	16.18

Table 8. Sets for different weighting factors. Source: Own compilation.

Weighting Method Indicator	Equal Weights	Political Decision Makers' Objective	Producers' Objective	Consumers' Objective	Average Actors' Weight
Net imports	0.111	0.158	0.111	0.059	0.109
Consumers' surplus	0.111	0.105	0.056	0.176	0.112
Avg. wholesale price	0.111	0.053	0.111	0.176	0.113
LCOE	0.111	0.053	0.167	0.118	0.112
Avg. price spread	0.111	0.105	0.167	0.059	0.110
CO <sub>2</sub> emissions	0.111	0.158	0.056	0.118	0.110
Environmental impact	0.111	0.105	0.056	0.176	0.112
Share of dom. production	0.111	0.158	0.111	0.059	0.109
Producers' surplus	0.111	0.105	0.167	0.059	0.110

Table 9 shows the results of the TOPSIS analysis, with the colors corresponding to the scenarios as explained in the Color Key. They show a clear pattern with regard to the ranking of the three main scenarios. While the Green scenarios prevail for all the weighting approaches, the Blue scenarios rank last. However, the layout of this paper does not allow for a comprehensive comparison of the three scenario groups—Yellow, Blue and Green. In particular, the fact that the Green sub-scenarios only allow a 35% reduction in NTC expansion makes a ranking between the other scenarios with a reduction in expanded capacity of 50% less meaningful (however, even with a reduction of 35% of NTC expansion in all the scenarios, the pattern presented in Table 9 does not change significantly. Appendix B presents further information). Therefore, the conclusions provided in this section will focus more strongly on the patterns within the respective scenario groups.

The results reveal that the impact of a reduction in interconnectivity is nuanced in that it (i) depends on the transition scenario that the UK pursues, and (ii) is different for each stakeholder.

A reduced interconnectivity of the GB electricity market does not necessarily negatively impact the observed stakeholder groups; in the Green scenario, there can be advantages from reduced interconnection, with a positive effect on the overall evaluation (i.e., 'Equal' and 'Average actors' weight'). In case of a development of the electricity market in accordance with the Green scenario, producers do not necessarily prefer a full development of future NTC projects, but possible advantages from a devaluation of the Pound, do not offset their negative impacts, as both Green variants featuring devaluation rank last within their respective groups. In contrast, from the perspective of consumers, a full development of NTC and no devaluation constitutes the best option in all three scenario groups. Between the variants of the Green scenario with no devaluation and those with a devalued Pound, there is a considerable drop in the scenario valuation. This can be mostly explained by the reduction in consumers' surpluses that occur due to a devaluation of the Pound. Thus, from a consumer perspective, a devaluation of the Pound leads to a more severe decrease in the overall valuation than a reduction in the NTC development. However, under the Green scenario, reduced growth in NTC capacity leads to

Energies **2019**, 12, 3212 17 of 27

risks to the stability of the system, since although imports are lower overall, there are times of the year in which imports are more acutely important to meeting demand. This is why the power system is no longer viable in the Green scenarios when NTCs fall below 65% of their potential level. In the Green scenario, opportunities for lower overall reliance on imports (although there are points of time at which imports have greater importance than in the other scenarios) and opportunities for exports must be contrasted with this greater risk to the stability of the system resulting from reduced growth in NTCs.

Rank	Rank Equal		Actors' Perspective: Political Decision Makers	Actors' Perspective: PRODUCERS	Actors' Perspective: Consumers
1	0.785	0.783	0.811	0.850	0.738
2	0.784	0.782	0.811	0.850	0.735
3	0.752	0.750	0.789	0.846	0.679
4	0.748	0.745	0.785	0.843	0.673
5	0.596	0.599	0.573	0.593	0.666
6	0.581	0.581	0.552	0.582	0.590
7	0.540	0.540	0.543	0.580	0.531
8	0.530	0.528	0.527	0.562	0.483
9	0.376	0.379	0.344	0.266	0.468
10	0.337	0.339	0.322	0.262	0.390
11	0.294	0.293	0.318	0.260	0.285
12	0.272	0.270	0.309	0.258	0.237
			Color Key:		
C		Expansion of	NTC as planned	Reduced incr	reases in NTC
Scer	nario	Default	Devaluated	Default	Devaluated
Yel	low				
Bl	lue				
Gr	een				

**Table 9.** TOPSIS analysis of modeling results. Source: Own calculation.

In the case of the Yellow scenario, Table 9 reveals that all the weightings with the exception of 'Actors' Perspective: Producers' and 'Actors' Perspective: Political Decision Makers' evaluate an environment with full NTC development and no devaluation of the Pound as the most favorable option. The producer-oriented weighting shows that from a generator's viewpoint, a development with reduced NTC expansion and devaluation would be preferred. A similar picture emerges when assessing the Blue scenario. However, in contrast to the Yellow scenario, the lower expansion in NTC has a stronger effect on the evaluation of the Blue scenario. As a result, the Blue scenarios with reduced NTC development rank last among those scenario variants, with the exception of the producers' viewpoint. The risks in the Blue and Yellow scenarios relate to loss of export opportunities for the producers, loss of electricity market competitiveness for the consumers, and loss of energy independence opportunities, from the point of view of political decision makers. Reflecting on the rationale behind interconnectors, Section 2.1 discussed their role in enhancing the competitiveness of electricity markets and in improving the stability within connected markets. It would appear that in the Green scenario, this role of ensuring stability is much more important, whereas for the Yellow and Blue scenarios, the contribution of interconnectors is chiefly to ensure price convergence between less competitive electricity markets (the GB market) and more competitive markets (EU markets).

## 5. Conclusions

By introducing the consideration of stakeholders' viewpoints to the assessment of quantified power system scenarios with an MCDA tool, this paper provides a differentiated analysis of the effects of Brexit on the GB power system. The applied methodology facilitates a versatile scenario analysis that takes into consideration the diversity of interests of the affected groups.

Energies **2019**, 12, 3212 18 of 27

The analysis presented shows that Brexit is not inherently negative for the GB electricity market. At the same time, the results show that the effectiveness of future policy measures aiming at the attenuation of possible negative impacts strongly depend on the design of the power system. In the Green scenario, implications of Brexit could bring about positive developments for the GB power system, but the results indicate that the market is in need of additional flexibility options in case of a reduced NTC development. If the electricity system moves more strongly toward an energy landscape resembling the Yellow scenario, focus on currency stability could be more important than a strong drive toward further interconnection. On the other hand, the MCDA outputs show that conflicts of interest between the considered stakeholder groups may arise. This becomes especially apparent for producers' and consumers' interests in the Yellow scenario, where the ranking of the scenario variants diverges substantially.

In the Blue scenario, given that the higher carbon price (e.g., due to the CPF) leads to a considerable increase in costs, a tradeoff between environmental protection and consumer welfare emerges. In the longer term, in the Blue and Yellow scenarios, the UK may have to decide if it should build greater domestic power capacities to compensate for a reduced ability to import more competitively priced power. The impact of Brexit on the sustainability of the GB and EU power systems remains unclear. There is debate in Section 2.1 on the influence of interconnection on overall  $CO_2$  emissions and, principally, the carbon leakage effect. In the Blue scenario, the domestic environmental impact of GB's power system is reduced the most, but this is due to the fall in domestic generation, so the emissions are likely to be produced elsewhere.

This work focuses on the implications of Brexit on the electricity market with consideration of different stakeholders' viewpoints. However, it excludes the extensive interactions of the economy-wide effects of Brexit with the power system. While econometric assessments already addressed issues concerning Brexit (see e.g., Raddant [81]), to the authors' knowledge, there exists no contributions that carry out a coupled approach of econometric modeling frameworks with a power system model. As a starting point for new research activities, this link could provide new valuable insights into the topic.

Further research avenues may include analyzing the implications of different forms of Brexit on the power system—i.e., comparing the impacts on the power system of a Brexit in which the UK retains a closer relationship with the EU compared to a Brexit in which the UK's relationship with the EU is more distant. The nature of possible advantages for the power system, which Brexit may entail, as referred to in Section 2.2, contrasted with potential wider challenges, such as the effects of changes in access to EU energy research and EU finance, also warrant further investigation. Pollitt [5] and Dutton and Lockwood [40] argued that Brexit might deliver attractive opportunities, which is to some degree congruent with the analysis presented in this work. Nevertheless, a higher level of autonomy in designing environmental and energy policy does not necessarily mean that British policies can be regarded as fully independent from European energy policy. This is exemplified by the Blue scenario: a CPF that strongly diverges from ETS prices may lead to asymmetric market conditions by preventing a level playing field for generators.

**Author Contributions:** This research article is a combined effort of TU Bergakademie Freiberg and Forschungszentrum Jülich. The contributions of the respective authors can be categorized as follows: Conceptualization and an initial literature review was done by C.B. and P.M. Together, they designed the original draft of the paper. A more extensive review and assessment of literature was conducted by C.B. The methodology can be separated into two parts. The modeling approach relies on the contributions of S.V. and P.M., while the resources for the modeling approach were provided by S.V. The MCDA results were computed by Ball and Mayer. The visualizations were processed by C.B. and P.M. The project was supervised by W.K. and D.R. Writing the article was a joint effort of all contributing authors. The final remarks and editing of W.K., D.R. and S.V. were used to finalize the article.

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Energies **2019**, 12, 3212 19 of 27

Conflicts of Interest: The authors declare no conflict of interest.

# Appendix A. TOPSIS Method

The approach follows the procedure presented by [69]. TOPSIS evaluates the MCDA alternative with regard to their distance to a positive-ideal and negative-ideal alternative. In order to do so, each performance indicator  $d_{i,j}$  of all n criteria  $C_j$  with j = 1, ..., n has to be transformed into a decision matrix M for each k alternative  $S_i$ , where i = 1, ..., k.

$$C_1 C_2 \dots C_{j\dots} C_n$$
 (A1)

Due to the heterogeneity of the computed MCDA criteria, the matrix is normalized with:

$$p_{i,j} = \frac{d_{i,j}}{\sqrt{\sum_{i=1}^{k} d_{i,j}^2}} \quad \forall d_{i,j} \neq 0, \ \forall \ j = 1, \dots, n, \text{ and } \forall \ i = 1, \dots, k$$
(A2)

or

$$p_{i,j} = 0$$
  $\forall d_{i,j} = 0, \ \forall \ j = 1, ..., \ n \text{ and } \forall \ i = 1, ..., \ k$  (A3)

The normalized criteria are weighted with the factors described with Equations (10)–(12). This results in the normalized and weighted Matrix  $\overline{M}$ :

$$C_{1} C_{2} \dots C_{j...} C_{n} \qquad \overline{M} = \begin{bmatrix} w_{1}p_{11} & w_{2}p_{12} & \dots & w_{j}p_{1j} & \dots & w_{n}p_{1n} \\ w_{1}p_{21} & w_{2}p_{22} & \dots & & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ w_{1}p_{i1} & w_{2}p_{i2} & \dots & w_{j}p_{ij} & \dots & w_{n}p_{in} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ w_{1}p_{k1} & w_{2}p_{k2} & \dots & w_{3}p_{kj} & \dots & w_{n}p_{kn} \end{bmatrix}$$
(A4)

Table A1 shows the compilation of MCDA criteria after normalization.

Table A1. Normalized MCDA criteria. Source: own calculation.

Criteria/Scenario	Net Imports	Con. Rent	Avg. Wholesale Price	LCOE	Weightend Price Spread	CO <sub>2</sub> [Mio t]	GWP [Mio t]	Share of Dom. Production	Prod Surplus
Yellow_100_0	0.292	0.000	0.262	0.260	0.200	0.334	0.264	0.274	0.209
Blue_100_0	0.558	-0.356	0.316	0.297	0.403	0.203	0.501	0.228	0.251
Green_100_0	0.004	0.126	0.244	0.287	0.103	0.176	0.223	0.320	0.342
Yellow_100_10	0.229	-0.107	0.278	0.266	0.187	0.381	0.153	0.284	0.226
Blue_100_10	0.529	-0.467	0.333	0.311	0.387	0.215	0.477	0.233	0.265
Green_100_10	-0.028	0.023	0.259	0.296	0.081	0.198	0.186	0.328	0.369
Yellow_50_0	0.181	-0.048	0.269	0.257	0.238	0.385	0.187	0.292	0.221
Blue_50_0	0.335	-0.486	0.336	0.310	0.497	0.298	0.332	0.266	0.282
Green_65_0	-0.004	0.130	0.243	0.290	0.100	0.183	0.218	0.321	0.340
Yellow_50_10	0.140	-0.147	0.284	0.264	0.220	0.425	0.113	0.299	0.236
Blue_50_10	0.322	-0.591	0.352	0.320	0.478	0.305	0.324	0.268	0.294
Green_65_10	-0.040	0.034	0.257	0.298	0.073	0.203	0.185	0.328	0.365

Each scenario is evaluated with regard to a distance dimension *A*:

$$A_i = \frac{D_i^-}{D_i^- + D_i^+} \tag{A5}$$

where *D* describes the Euclidian distance toward both alternatives, with:

$$D_i^+ = \sqrt{\sum_{j=1}^n \left( w_i p_{ij} - w_i p_j^+ \right)^2} \qquad \forall i = 1, \dots, k \text{ and } \forall j = 1, \dots, n$$
 (A6)

Energies **2019**, 12, 3212 20 of 27

$$D_i^- = \sqrt{\sum_{j=1}^n \left( w_i p_{ij} - w_i p_j^- \right)^2} \qquad \forall i = 1, \dots, k \text{ and } \forall j = 1, \dots, n$$
 (A7)

where

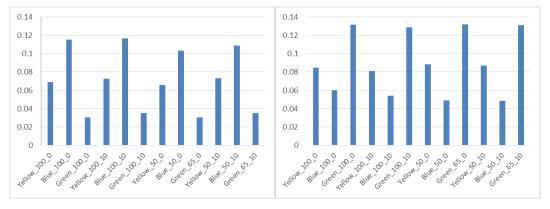
- $D_i^+$ : Euclidian distance to positive-ideal alternative
- $D_i^-$ : Euclidian distance to negative-ideal alternative
- *p*<sup>-</sup>: negative ideal reference point
- $p^+$ : positive-ideal reference point

Table A2 presents the positive-ideal and negative ideal alternatives toward which the Euclidian distance of the scenarios under consideration is calculated.

Table A2. Positive-ideal and negative-ideal alternatives. Source: Own compilation.

	Net Imports	Con. Rent	Avg. Wholesale Price	LCOE	Weightend Price Spread	CO <sub>2</sub> [Mio t]	GWP [Mio t]	Share of Dom. Production	Prod Surplus
Positive Ideal	-0.002	0.023	0.043	0.030	0.004	0.021	0.088	0.019	0.022
Negative Ideal	0.033	-0.104	0.062	0.038	0.029	0.050	0.020	0.013	0.012

The Euclidian distance of the scenarios under consideration of the positive-ideal and negative-ideal solutions are presented in Figure A1:



**Figure A1.** Euclidian distance of scenarios and their variants to positive-ideal (**left**) and negative-ideal (**right**) alternatives. Source: Own compilation.

# Appendix B. Robustness of Results toward the NTC Development Assumptions

In order to provide insights on the robustness of the presented data and results, calculations were applied, with harmonized reductions of the future NTC expansions across all the scenarios and their variants. Table A3 shows the TOPSIS results for an extension of interconnectors reduced by 35%, with the colors corresponding to the scenarios as explained in the Color Key.

With respect to the sequence of Yellow, Blue, and Green, no change can be observed compared to a mixed reduction of the NTC expansion (Yellow and Blue scenarios by 50% and Green scenario by 35%), as presented in Table 9. In addition, the ranking of the respective variants across the different weighting factors do not show significant differences. The only exception is 'Actors' three scenarios shows a divergent pattern. However, this does not contradict the general conclusions presented in Sections 4 and 5.

Energies **2019**, 12, 3212 21 of 27

Table A3.	TOPSIS	analysis	of model	ing result	s with	35%	reduction	of NTC	expansion.	Source:
Own comp	ilation.									

Rank	Equal	Avg. Actors Weight	Actors Perspective: Political Decision Maker	Actors Perspective: Producers	Actors Perspective: Consumers
1	0.854	0.853	0.871	0.881	0.829
2	0.832	0.830	0.853	0.873	0.800
3	0.804	0.802	0.834	0.867	0.745
4	0.789	0.787	0.821	0.865	0.727
5	0.600	0.603	0.580	0.580	0.690
6	0.599	0.600	0.553	0.570	0.629
7	0.525	0.525	0.533	0.556	0.519
8	0.522	0.521	0.513	0.542	0.487
9	0.354	0.357	0.329	0.239	0.435
10	0.322	0.324	0.312	0.231	0.371
11	0.306	0.307	0.308	0.230	0.336
12	0.279	0.279	0.296	0.230	0.282
			Colour key:		

C	Expans	sion of NTC as planned	Reduced increases in NTC			
Scenario	Default	Devalued	Default	Devalued		
Yellow						
Blue						
Green						

# Appendix C. Robustness of Results in Comparison to the PROMETHEE Method

An alternative MCDA technique—the PROMETHEE method—is used to gain an impression of the sensitivity of the results from the TOPSIS method to the use of other decision analysis techniques. The PROMETHEE method (Preference Ranking Organization Method for Enrichment Evaluation) is based on outranking flows associated with a particular alternative (in this case, the different scenarios for the GB power system in 2030). The PROMETHEE method is a set of procedures that was developed in the 1980s by Brans, et al. [82] and Behzadia, et al. The authors of [83] conduct a review of different methodologies and applications of the PROMETHEE method.

Outranking flows are calculated from preference functions that evaluate the pairwise difference between the performance scores of scenarios on each criterion. These performance scores are then multiplied by the respective actor-specific weighting for each criterion. The following equation, which was adapted from Mareschal [84], gives the outranking relation:

$$\pi\left(S_{a}, S_{b}\right) = \sum_{K=1}^{n} W_{k} \times P_{k} \tag{A8}$$

The outranking relation  $\pi$  ( $S_a$ ,  $S_b$ ) assesses, for all the criteria (K), the extent to which scenario a ( $S_a$ ) is outranked by scenario b ( $S_b$ ), and corresponds to the sum of all the preference functions ( $P_k$ ) (pairwise differences between scenarios in performance scores on the respective criteria) multiplied by the actor-specific weightings for those criteria ( $W_k$ ).

Following the computation of the outranking relations, it is then possible to build the positive outranking flow and the negative outranking flow for each scenario. The positive outranking flow for scenario a,  $(\phi^+(a))$  consists of the instances in which scenario a performed better than another scenario overall (once all the criteria were considered), and the sum of the magnitude of those superior performances. Conversely, the negative outranking flow for scenario a,  $(\phi^-(a))$  consists of instances in which scenario a performed worse than another scenario overall (once all the criteria were considered) and the sum of the magnitude of those inferior performances. The positive and negative outranking flows, as reproduced from Brans and Mareschal [85], are depicted below:

Energies **2019**, 12, 3212 22 of 27

Positive outranking flow (scenario a outperforming other scenarios):

$$\varphi^{+}(a) = \frac{1}{n-1} \sum \pi(a, x)$$
 (A9)

Negative outranking flow (scenario a outperformed by other scenarios):

$$\varphi^{-}(a) = \frac{1}{n-1} \sum \pi(x, a)$$
 (A10)

The net preference flow is computed from the positive outranking flow minus the negative outranking flow (( $\varphi^+(a) - \varphi^-(a)$ ) [86]; the higher the net preference flow, the better the scenario, from the perspective of any particular actor.

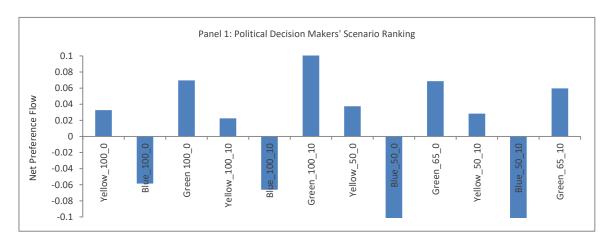
This process was followed for the three weighting factors representing the stakeholders' perspectives, namely: (i) political decision makers, (ii) producers, and (iii) consumers, and the results are depicted in Table A4 below:

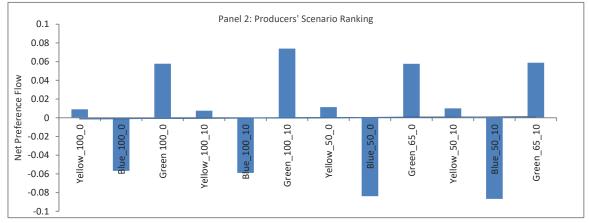
Table A4. Outranking flow results.

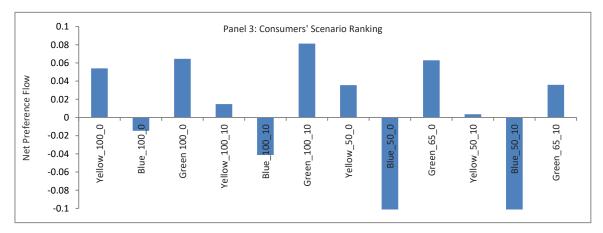
Positive Flow   Negative Flow   Net Flow   Ranking (PDM		lable A	4. Outranking no	w resurts.		
Yellow_100_0		Outranking	Flows (Political De	cision Makers)		
Blue_100_0		Positive Flow	Negative Flow	Net Flow	Ranking (PDM)	
Green 100_0         0.073878944         −0.004282498         0.069596446         2           Yellow_100_10         0.044142223         −0.02168776         0.022454463         8           Blue_100_10         0.015248548         −0.061395048         −0.0661465         10           Green_100_10         0.120986423         0         0.120986423         1           Yellow_50_0         0.051555197         −0.014085363         0.037469834         5           Blue_50_0         0.000869473         −0.153290427         −0.1522420955         11           Green_65_0         0.073089011         −0.004440485         0.068648526         3           Yellow_50_10         0.046594172         −0.018255031         0.02839142         7           Blue_50_10         0         0.066356837         −0.006684543         0.059672294         4           Outranking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         −0.0179437         0.00906281         7           Blue_100_0         0.059216051         −0.001701775         −0.05676717         9           Green 100_0         0.059363627         −0.018843734         0.00751989         8 <td>Yellow_100_0</td> <td>0.048785448</td> <td>-0.016063755</td> <td>0.032721694</td> <td>6</td>	Yellow_100_0	0.048785448	-0.016063755	0.032721694	6	
Yellow_100_10         0.044142223         -0.02168776         0.022454463         8           Blue_100_10         0.015248548         -0.081395048         -0.0661465         10           Green_100_10         0.120986423         0         0.120986423         1           Yellow_50_0         0.051555197         -0.014085363         0.037469834         5           Blue_50_0         0.000869473         -0.153290427         -0.152420955         11           Green_65_0         0.073089011         -0.004440485         0.068648526         3           Yellow_50_10         0         -0.162854625         12           Green_65_10         0         -0.162854625         12           Green_65_10         0         0.066356837         -0.006684543         0.059672294         4           Dutranking Flows (Producers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.02706509         -0.0179437         0.0096281         7           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.02706509         -0.0179437         0.0056281         7 <td col<="" td=""><td>Blue_100_0</td><td>0.017168488</td><td>-0.07563523</td><td>-0.058466742</td><td>9</td></td>	<td>Blue_100_0</td> <td>0.017168488</td> <td>-0.07563523</td> <td>-0.058466742</td> <td>9</td>	Blue_100_0	0.017168488	-0.07563523	-0.058466742	9
Blue_100_10	Green 100_0	0.073878944	-0.004282498	0.069596446	2	
Green_100_10         0.120986423         0         0.120986423         1           Yellow_50_0         0.051555197         −0.014085363         0.037469834         5           Blue_50_0         0.000869473         −0.153290427         −0.152420955         11           Green_65_0         0.073089011         −0.004440485         0.068648526         3           Yellow_50_10         0         −0.162854625         −0.162854625         12           Green_65_10         0.066356837         −0.006684543         0.059672294         4           Outrarking Flows (Producers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.027066509         −0.0179437         0.00906281         7           Blue_100_0         0.024934607         −0.061701775         −0.05676717         9           Green 100_0         0.059216051         −0.001440161         0.05777589         3           Yellow_100_10         0.026363627         −0.018443734         0.00751989         8           Blue_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         −0.016910624         0.01132526         5	Yellow_100_10	0.044142223	-0.02168776	0.022454463	8	
Yellow_50_0         0.051555197         −0.014085363         0.037469834         5           Blue_50_0         0.000869473         −0.153290427         −0.152420955         11           Green_65_0         0.073089011         −0.004440485         0.068648526         3           Yellow_50_10         0.046594172         −0.018255031         0.028339142         7           Blue_50_10         0         −0.162854625         −0.162854625         12           Green_65_10         0.066356837         −0.006684543         0.059672294         4           Outranking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         −0.0179437         0.00906281         7           Blue_100_0         0.004934607         −0.061701775         −0.05676717         9           Green_100_1         0.052363627         −0.018433734         0.00771989         8           Blue_100_10         0.024363627         −0.018433434         0.007395354         1           Yellow_50_0         0.028262983         −0.016910624         0.01135236         5           Blue_50_0         0.00240301         −0.084095685         −0.08835538         11<	Blue_100_10	0.015248548	-0.081395048	-0.0661465	10	
Blue_50_0	Green_100_10	0.120986423	0	0.120986423	1	
Green_65_0         0.073089011         -0.004440485         0.068648526         3           Yellow_50_10         0.046594172         -0.018255031         0.028339142         7           Blue_50_10         0         -0.162854625         -0.162854625         12           Green_65_10         0.066356837         -0.006684543         0.059672294         4           Outrarking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         -0.0179437         0.00906281         7           Blue_100_0         0.004934607         -0.061701775         -0.05676717         9           Green 100_0         0.059216051         -0.001440161         0.05777589         3           Yellow_100_10         0.026363627         -0.018843734         0.00751989         8           Blue_100_10         0.04395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.00243031         -0.084095685         -0.08385538         11 <td< td=""><td>Yellow_50_0</td><td>0.051555197</td><td>-0.014085363</td><td>0.037469834</td><td>5</td></td<>	Yellow_50_0	0.051555197	-0.014085363	0.037469834	5	
Yellow_50_10         0.046594172         -0.018255031         0.028339142         7           Blue_50_10         0         -0.162854625         -0.162854625         12           Creen_65_10         0.066356837         -0.006684543         0.059672294         4           Outrarking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         -0.0179437         0.00906281         7           Green 100_0         0.029216051         -0.061701775         -0.05676717         9           Green 100_0         0.059216051         -0.001440161         0.05777589         3           Yellow_100_10         0.026363627         -0.018843734         0.00751989         8           Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.022862983         -0.016910624         0.01135236         5           Blue_50_0         0.009240301         -0.084095685         -0.0889538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4 </td <td>Blue_50_0</td> <td>0.000869473</td> <td>-0.153290427</td> <td>-0.152420955</td> <td>11</td>	Blue_50_0	0.000869473	-0.153290427	-0.152420955	11	
Blue_50_10         0         −0.162854625         −0.162854625         12           Green_65_10         0.066356837         −0.006684543         0.059672294         4           Outrarbing Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         −0.0179437         0.00906281         7           Blue_100_0         0.004934607         −0.061701775         −0.05676717         9           Green 100_0         0.059216051         −0.001440161         0.05777589         3           Yellow_100_10         0.026363627         −0.018843734         0.00751989         8           Blue_100_10         0.04395798         −0.063318201         −0.0589224         10           Green_100_10         0.073953543         0         0         0.07395354         1           Yellow_50_0         0.028262983         −0.016910624         0.01135236         5           Blue_50_0         0.00240301         −0.084095685         −0.08385538         11           Green_65_0         0.059179629         −0.001452301         0.05772733         4           Yellow_50_10         0.027481087         −0.017469121         0.01001197		0.073089011	-0.004440485	0.068648526	3	
Green_65_10         0.066356837         −0.006684543         0.059672294         4           Outrarking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         −0.0179437         0.00906281         7           Blue_100_0         0.004934607         −0.061701775         −0.05676717         9           Green 100_0         0.059216051         −0.001440161         0.05777589         3           Yellow_100_10         0.026363627         −0.018843734         0.00751989         8           Blue_100_10         0.04395798         −0.063318201         −0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         −0.016910624         0.01135236         5           Blue_50_0         0.00240301         −0.084095685         −0.08385538         11           Green_65_0         0.059179629         −0.001452301         0.05772733         4           Yellow_50_10         0.027481087         −0.017469121         0.01010197         6           Blue_50_10         0.06013628         −0.001256115         0.0588017         2      <	Yellow_50_10	0.046594172	-0.018255031	0.028339142	7	
Green_65_10         0.066356837         −0.006684543         0.059672294         4           Outrarking Flows (Producers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         −0.0179437         0.00906281         7           Blue_100_0         0.004934607         −0.061701775         −0.05676717         9           Green 100_0         0.059216051         −0.001440161         0.05777589         3           Yellow_100_10         0.026363627         −0.018843734         0.00751989         8           Blue_100_10         0.04395798         −0.063318201         −0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         −0.016910624         0.01135236         5           Blue_50_0         0.00240301         −0.084095685         −0.08385538         11           Green_65_0         0.059179629         −0.001452301         0.05772733         4           Yellow_50_10         0.027481087         −0.017469121         0.01010197         6           Blue_50_10         0.06013628         −0.001256115         0.0588017         2      <	Blue_50_10	0	-0.162854625	-0.162854625	12	
Yellow_100_0         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.027006509         -0.0179437         0.00906281         7           Blue_100_0         0.004934607         -0.061701775         -0.05676717         9           Green 100_0         0.059216051         -0.001440161         0.05777589         3           Yellow_100_10         0.026363627         -0.018843734         0.00751989         8           Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.0282662983         -0.016910624         0.01135236         5           Blue_50_0         0.00024301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.0588017         2           Vellow_100_0         0.05790375		0.066356837	-0.006684543	0.059672294	4	
Yellow_100_0         0.027006509         -0.0179437         0.00906281         7           Blue_100_0         0.004934607         -0.061701775         -0.05676717         9           Green 100_0         0.059216051         -0.001440161         0.05777589         3           Yellow_100_10         0.026363627         -0.018843734         0.00751989         8           Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         0.066738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Dutsurking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow		Outr	anking Flows (Proc	ducers)		
Blue_100_0		Positive Flow	Negative Flow	Net Flow	Ranking	
Blue_100_0	Yellow_100_0	0.027006509	-0.0179437	0.00906281	7	
Green 100_0         0.059216051         -0.001440161         0.05777589         3           Yellow_100_10         0.026363627         -0.018843734         0.00751989         8           Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outranking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0446426         9 </td <td></td> <td>0.004934607</td> <td></td> <td>-0.05676717</td> <td>9</td>		0.004934607		-0.05676717	9	
Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrawing Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10		0.059216051	-0.001440161	0.05777589	3	
Blue_100_10         0.004395798         -0.063318201         -0.0589224         10           Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10	Yellow_100_10	0.026363627	-0.018843734	0.00751989	8	
Green_100_10         0.073953543         0         0.07395354         1           Yellow_50_0         0.028262983         -0.016910624         0.01135236         5           Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrawing Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1		0.004395798	-0.063318201	-0.0589224	10	
Blue_50_0         0.000240301         -0.084095685         -0.08385538         11           Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.004564837         -0.01006055         0.03559912         11	Green_100_10	0.073953543	0	0.07395354	1	
Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.00645932         -0.00166254         0.06293066         3	Yellow_50_0	0.028262983	-0.016910624	0.01135236	5	
Green_65_0         0.059179629         -0.001452301         0.05772733         4           Yellow_50_10         0.027481087         -0.017469121         0.01001197         6           Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.00645932         -0.00166254         0.06293066         3	Blue_50_0	0.000240301	-0.084095685	-0.08385538	11	
Blue_50_10         0         -0.086738998         -0.086739         12           Green_65_10         0.06013628         -0.001256115         0.05888017         2           Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8 <tr< td=""><td></td><td>0.059179629</td><td>-0.001452301</td><td>0.05772733</td><td>4</td></tr<>		0.059179629	-0.001452301	0.05772733	4	
Green_65_10         0.06013628         −0.001256115         0.05888017         2           Outranking Flows (Consumers)           Positive Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         −0.00389236         0.05401139         4           Blue_100_0         0.02450183         −0.03914444         −0.0146426         9           Green 100_0         0.0659491         −0.00139136         0.06455773         2           Yellow_100_10         0.03524032         −0.02046861         0.01477171         7           Blue_100_10         0.01785543         −0.05908365         −0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         −0.01006055         0.03558782         6           Blue_50_0         0.0020616         −0.1380528         −0.1359912         11           Green_65_0         0.0645932         −0.00166254         0.06293066         3           Yellow_50_10         0.03057175         −0.02700461         0.00356713         8           Blue_50_10         0         −0.16073039 <td< td=""><td>Yellow_50_10</td><td>0.027481087</td><td>-0.017469121</td><td>0.01001197</td><td>6</td></td<>	Yellow_50_10	0.027481087	-0.017469121	0.01001197	6	
Outrarking Flows (Consumers)           Positive Flow         Negative Flow         Net Flow         Ranking           Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12	Blue_50_10	0	-0.086738998	-0.086739	12	
Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12	Green_65_10	0.06013628	-0.001256115	0.05888017	2	
Yellow_100_0         0.05790375         -0.00389236         0.05401139         4           Blue_100_0         0.02450183         -0.03914444         -0.0146426         9           Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12		Outra	anking Flows (Cons	sumers)		
Blue_100_0       0.02450183       -0.03914444       -0.0146426       9         Green 100_0       0.0659491       -0.00139136       0.06455773       2         Yellow_100_10       0.03524032       -0.02046861       0.01477171       7         Blue_100_10       0.01785543       -0.05908365       -0.04122822       10         Green_100_10       0.0812541       0       0.0812541       1         Yellow_50_0       0.04564837       -0.01006055       0.03558782       6         Blue_50_0       0.0020616       -0.1380528       -0.1359912       11         Green_65_0       0.0645932       -0.00166254       0.06293066       3         Yellow_50_10       0.03057175       -0.02700461       0.00356713       8         Blue_50_10       0       -0.16073039       -0.16073039       12		Positive Flow	Negative Flow	Net Flow	Ranking	
Green 100_0         0.0659491         -0.00139136         0.06455773         2           Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12	Yellow_100_0	0.05790375	-0.00389236	0.05401139		
Yellow_100_10         0.03524032         -0.02046861         0.01477171         7           Blue_100_10         0.01785543         -0.05908365         -0.04122822         10           Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12	Blue_100_0	0.02450183	-0.03914444	-0.0146426	9	
Yellow_100_10     0.03524032     -0.02046861     0.01477171     7       Blue_100_10     0.01785543     -0.05908365     -0.04122822     10       Green_100_10     0.0812541     0     0.0812541     1       Yellow_50_0     0.04564837     -0.01006055     0.03558782     6       Blue_50_0     0.0020616     -0.1380528     -0.1359912     11       Green_65_0     0.0645932     -0.00166254     0.06293066     3       Yellow_50_10     0.03057175     -0.02700461     0.00356713     8       Blue_50_10     0     -0.16073039     -0.16073039     12	Green 100_0	0.0659491	-0.00139136	0.06455773	2	
Green_100_10         0.0812541         0         0.0812541         1           Yellow_50_0         0.04564837         -0.01006055         0.03558782         6           Blue_50_0         0.0020616         -0.1380528         -0.1359912         11           Green_65_0         0.0645932         -0.00166254         0.06293066         3           Yellow_50_10         0.03057175         -0.02700461         0.00356713         8           Blue_50_10         0         -0.16073039         -0.16073039         12	Yellow_100_10	0.03524032	-0.02046861	0.01477171	7	
Yellow_50_0     0.04564837     -0.01006055     0.03558782     6       Blue_50_0     0.0020616     -0.1380528     -0.1359912     11       Green_65_0     0.0645932     -0.00166254     0.06293066     3       Yellow_50_10     0.03057175     -0.02700461     0.00356713     8       Blue_50_10     0     -0.16073039     -0.16073039     12	Blue_100_10	0.01785543	-0.05908365	-0.04122822	10	
Blue_50_0     0.0020616     -0.1380528     -0.1359912     11       Green_65_0     0.0645932     -0.00166254     0.06293066     3       Yellow_50_10     0.03057175     -0.02700461     0.00356713     8       Blue_50_10     0     -0.16073039     -0.16073039     12	Green_100_10	0.0812541	0	0.0812541	1	
Blue_50_0     0.0020616     -0.1380528     -0.1359912     11       Green_65_0     0.0645932     -0.00166254     0.06293066     3       Yellow_50_10     0.03057175     -0.02700461     0.00356713     8       Blue_50_10     0     -0.16073039     -0.16073039     12	Yellow_50_0	0.04564837	-0.01006055	0.03558782	6	
Yellow_50_10		0.0020616	-0.1380528	-0.1359912	11	
Blue_50_10 0 -0.16073039 -0.16073039 12	Green_65_0	0.0645932	-0.00166254	0.06293066	3	
Blue_50_10 0 -0.16073039 -0.16073039 12	Yellow_50_10	0.03057175	-0.02700461	0.00356713	8	
		0	-0.16073039	-0.16073039	12	
		0.04583739	-0.00992554	0.03591186	5	

Energies **2019**, 12, 3212 23 of 27

In Figure A2, the scenario rankings for the different actors, resulting from the PROMETHEE process, are depicted.







**Figure A2.** Scenario ranking for: (i) political decision makers (Panel 1), (ii) producers (Panel 2) and (iii) consumers (Panel 3).

## References

- 1. House of Commons: DCMS Committee. *The Potential Impact Of Brexit On The Creative Industries, Tourism and The Digital Single Market*; House of Commons: London, UK, 2018; pp. 1–67.
- 2. Cambridge Econometrics. Preparing for Brexit; Cambridge Econometrics: Cambridge, UK, 2018; pp. 1–86.
- 3. Kierzenkowski, R.; Pain, N.; Rusticelli, E.; Zwart, S. The Economic Consequences of Brexit. In *OECD Economic Policy Papers*; OECD: Paris, France, 2016; Volume 16, pp. 1–37.

Energies **2019**, 12, 3212 24 of 27

4. Hepburn, C.; Teytelboym, A. Climate change policy after Brexit. *Oxf. Rev. Econ. Policy* **2017**, *33*, 144–154. [CrossRef]

- 5. Pollitt, M.G. The economic consequences of Brexit: energy. *Oxf. Rev. Econ. Policy* **2017**, *33*, 134–143. [CrossRef]
- 6. Acquah-Andoh, E.; Ifelebuegu, A.O.; Theophilus, S.C. Brexit and UK Energy Security: Perspectives from Unconventional Gas Investment and the Effects of Shale Gas on UK Energy Prices. *Energies* **2019**, *12*, 600. [CrossRef]
- 7. Ifelebuegu, A.; Aidelojie, K.; Acquah-Andoh, E. Brexit and Article 50 of the Treaty of the European Union: Implications for UK Energy Policy and Security. *Energies* **2017**, *10*, 2143. [CrossRef]
- 8. European Commission. Projects of Common Interest. Available online: https://ec.europa.eu/energy/en/topics/infrastructure/projects-common-interest (accessed on 25 July 2017).
- 9. Grubb, M.; Newbery, D. UK Electricity Market Reform and the Energy Transition: Emerging Lessons. *Energy J.* **2018**, 39. [CrossRef]
- 10. DECC. *Updated Energy and Emissions Projections*; Department of Energy & Climate Change: London, UK, 2014; pp. 1–47.
- 11. ENTSO-E. TYNDP 2014 Visions Data. Available online: https://www.entsoe.eu/major-projects/ten-year-network-development-plan/tyndp-2014/Pages/default.aspx (accessed on 1 May 2017).
- 12. Ofgem. Electricity Interconnectors. Available online: https://www.ofgem.gov.uk/electricity/transmission-networks/electricity-interconnectors (accessed on 25 July 2017).
- 13. Dutton, J. *The Politics of Cross-Border Electricity Market Connection*; Working Paper 1602; UK Energy Research Centre: Exeter, UK, 2016; pp. 1–36.
- 14. DBEIS. *Digest of United Kingdom Energy Statistics* 2018; Department for Business, Energy & Industrial Strategy: London, UK, July 2018; pp. 1–41.
- 15. European Central Bank. ECB Euro Reference Exchange Rate: Pound Sterling (GBP). Available online: https://www.ecb.europa.eu/stats/policy\_and\_exchange\_rates/euro\_reference\_exchange\_rates/html/eurofxref-graph-gbp.en.html (accessed on 25 November 2018).
- 16. Stewart, T.J.; French, S.; Rios, J. Integrating multicriteria decision analysis and scenario planning—Review and extension. *Omega* **2013**, *41*, 679–688. [CrossRef]
- 17. Shmelev, S.E.; van den Bergh, J.C. Optimal diversity of renewable energy alternatives under multiple criteria: An application to the UK. *Renew. Sustain. Energy Rev.* **2016**, *60*, 679–691. [CrossRef]
- 18. Montibeller, G.; Gummer, H.; Tumidei, D. Combining scenario planning and multi-criteria decision analysis in practice. *J. Multi-Criteria Decis. Anal.* **2006**, *14*, 5–20. [CrossRef]
- 19. McKenna, R.; Bertsch, V.; Mainzer, K.; Fichtner, W. Combining local preferences with multi-criteria decision analysis and linear optimization to develop feasible energy concepts in small communities. *Eur. J. Oper. Res.* **2018**, *268*, 1092–1110. [CrossRef]
- 20. Ribeiro, F.; Ferreira, P.; Araújo, M. Evaluating future scenarios for the power generation sector using a Multi-Criteria Decision Analysis (MCDA) tool: The Portuguese case. *Energy* **2013**, *52*, 126–136. [CrossRef]
- 21. Ram, C.; Montibeller, G.; Morton, A. Extending the use of scenario planning and MCDA for the evaluation of strategic options. *J. Oper. Res. Soc.* **2011**, *62*, 817–829. [CrossRef]
- 22. Volkart, K.; Weidmann, N.; Bauer, C.; Hirschberg, S. Multi-criteria decision analysis of energy system transformation pathways: A case study for Switzerland. *Energy Policy* **2017**, *106*, 155–168. [CrossRef]
- 23. Úrge-Vorsatz, D.; Herrero, S.T.; Dubash, N.K.; Lecocq, F. Measuring the co-benefits of climate change mitigation. *Ann. Rev. Environ. Resour.* **2014**, *39*, 549–582. [CrossRef]
- 24. Scholten, L.; Maurer, M.; Lienert, J. Comparing multi-criteria decision analysis and integrated assessment to support long-term water supply planning. *PloS ONE* **2017**, *12*, e0176663. [CrossRef] [PubMed]
- 25. Saarikoski, H.; Mustajoki, J.; Barton, D.N.; Geneletti, D.; Langemeyer, J.; Gomez-Baggethun, E.; Marttunen, M.; Antunes, P.; Keune, H.; Santos, R. Multi-Criteria Decision Analysis and Cost-Benefit Analysis: Comparing alternative frameworks for integrated valuation of ecosystem services. *Ecosyst. Serv.* 2016, 22, 238–249. [CrossRef]
- 26. Browne, D.; Ryan, L. Comparative analysis of evaluation techniques for transport policies. *Environ. Impact Assess. Rev.* **2011**, *31*, 226–233. [CrossRef]
- 27. European Commission. *Energy* 2020: *A Strategy for Competitive, Sustainable and Secure Energy*; European Commission: Brussels, Belgium, 2010; pp. 1–21.

Energies **2019**, 12, 3212 25 of 27

28. European Commission. *Making the Internal Energy Market Work*; European Commission: Brussels, Belgium, 2012; pp. 1–23.

- 29. Helm, D. The European framework for energy and climate policies. Energy Policy 2014, 64, 29–35. [CrossRef]
- 30. Maltby, T. European Union energy policy integration: A case of European Commission policy entrepreneurship and increasing supranationalism. *Energy Policy* **2013**, *55*, 435–444. [CrossRef]
- 31. Schmid, E.; Knopf, B. Quantifying the long-term economic benefits of European electricity system integration. *Energy Policy* **2015**, *87*, 260–269. [CrossRef]
- 32. Ringler, P.; Keles, D.; Fichtner, W. How to benefit from a common European electricity market design. *Energy Policy* **2017**, *101*, 629–643. [CrossRef]
- 33. Jamasb, T.; Pollitt, M. Electricity market reform in the European Union: Review of progress toward liberalization & integration. *Energy J.* **2005**, *26*, 11–41.
- 34. Kappf, L.; Pelkmans, J. *Interconnector Investment for a Well-functioning Internal Market: What EU Regime of Regulatory Incentives?* College of Europe: Bruges, Belgium, 2010; pp. 1–41.
- 35. Zakeri, B.; Price, J.; Zeyringer, M.; Keppo, I.; Mathiesen, B.V.; Syri, S. The direct interconnection of the UK and Nordic power market—Impact on social welfare and renewable energy integration. *Energy* **2018**, *162*, 1193–1204. [CrossRef]
- 36. Gugler, K.; Haxhimusa, A.; Liebensteiner, M. Integration of European Electricity Markets: Evidence from Spot Prices. *Energy J.* **2018**, 39. [CrossRef]
- 37. Keppler, J.H.; Meunier, W.; Coquentin, A. Determining Optimal Interconnection Capacity on the Basis of Hourly Demand and Supply Functions of Electricity. *Energy J.* **2018**, *39*. [CrossRef]
- 38. Mezősi, A.; Pató, Z.; Szabó, L. Assessment of the EU 10% interconnection target in the context of CO<sub>2</sub> mitigation. *Clim. Policy* **2016**, *16*, 658–672. [CrossRef]
- 39. Parail, V. *Properties of Electricity Prices and the Drivers of Interconnector Revenue*; Electricity Policy Research Group: Cambridge, UK, 2010; pp. 1–73.
- 40. Dutton, J.; Lockwood, M. Ideas, institutions and interests in the politics of cross-border electricity interconnection: Greenlink, Britain and Ireland. *Energy Policy* **2017**, *105*, 375–385. [CrossRef]
- 41. Edmunds, R.K.; Cockerill, T.T.; Foxon, T.J.; Ingham, D.B.; Pourkashanian, M. Technical benefits of energy storage and electricity interconnections in future British power systems. *Energy* **2014**, *70*, 577–587. [CrossRef]
- 42. Fürsch, M.; Hagspiel, S.; Jägemann, C.; Nagl, S.; Lindenberger, D.; Tröster, E. The role of grid extensions in a cost-efficient transformation of the European electricity system until 2050. *Appl. Energy* **2013**, *104*, 642–652. [CrossRef]
- 43. Neuhoff, K.; Barquin, J.; Bialek, J.W.; Boyd, R.; Dent, C.J.; Echavarren, F.; Grau, T.; von Hirschhausen, C.; Hobbs, B.F.; Kunz, F. Renewable electric energy integration: Quantifying the value of design of markets for international transmission capacity. *Energy Econ.* **2013**, *40*, 760–772. [CrossRef]
- 44. Lynch, M.Á.; Tol, R.S.J.; O'Malley, M.J. Optimal interconnection and renewable targets for north-west Europe. *Energy Policy* **2012**, *51*, 605–617. [CrossRef]
- 45. Brancucci Martínez-Anido, C.; Vandenbergh, M.; de Vries, L.; Alecu, C.; Purvins, A.; Fulli, G.; Huld, T. Medium-term demand for European cross-border electricity transmission capacity. *Energy Policy* **2013**, *61*, 207–222. [CrossRef]
- 46. Boie, I.; Fernandes, C.; Frías, P.; Klobasa, M. Efficient strategies for the integration of renewable energy into future energy infrastructures in Europe—An analysis based on transnational modeling and case studies for nine European regions. *Energy Policy* **2014**, *67*, 170–185. [CrossRef]
- 47. Tenggren, S.; Wangel, J.; Nilsson, M.; Nykvist, B. Transmission transitions: barriers, drivers, and institutional governance implications of Nordic transmission grid development. *Energy Res. Soc. Sci.* **2016**, *19*, 148–157. [CrossRef]
- 48. Hirst, D. *Carbon Price Floor (CPF) and the Price Support Mechanism*; House of Commons Library: London, UK, 2018: pp. 1–20.
- 49. Supponen, M. Factors that Influence the Targets and Criteria for Electricity Interconnector Investments; European University Institute: Florence, Italy, 2012; pp. 1–43.
- 50. General Secretariat of the Council. *Political Declaration Setting out the Framework for the Future Relationship between the European Union and the United Kingdom*; Council of the European Union: Brussels, Belgium, 22 November 2018.
- 51. Energy, UK. Brexit & the Future EU-UK Energy Relationship; Energy UK: London, UK, 2018; pp. 1–13.

Energies **2019**, 12, 3212 26 of 27

52. OECD. OECD Economic Surveys: United Kingdom 2017; Organisation for Economic Co-operation and Development: Paris, France, 2017. [CrossRef]

- 53. Office for National Statistics. Who Does the UK Trade with? Explore the UK's Trade Relationships through Interactive Maps, Charts and Analysis. Available online: https://www.ons.gov.uk/businessindustryandtrade/internationaltrade/articles/whodoestheuktradewith/2017-02-21 (accessed on 20 May 2018).
- 54. Chang, W.W. Brexit and its economic consequences. World Econ. 2018, 41, 2349–2373. [CrossRef]
- 55. Ebell, M.; Warren, J. The long-term economic impact of leaving the EU. *Natl. Inst. Econ. Rev.* **2016**, 236, 121–138. [CrossRef]
- 56. Dhingra, S.; Ottaviano, G.; Sampson, T. A hitch-hiker's guide to post-Brexit trade negotiations: Options and principles. *Oxf. Rev. Econ. Policy* **2017**, *33*, 22–30. [CrossRef]
- 57. Wilson, I.A.G.; Barbour, E.; Ketelaer, T.; Kuckshinrichs, W. An analysis of storage revenues from the time-shifting of electrical energy in Germany and Great Britain from 2010 to 2016. *J. Energy Storage* **2018**, 17, 446–456. [CrossRef]
- 58. Pye, S.; Mathieu, C.; Deane, P. The energy sector implications of Brexit. Hot Energy Top. 2017, 20, 1–9.
- 59. Nationalgrid ESO. *Future Energy Scenarios*; National Grid Electricity System Operator: London, UK, 2019; pp. 1–166.
- 60. DBEIS. *Digest of UK Energy Statistics (DUKES): Electricity;* Department for Business, Energy & Industrial Strategy: London, UK, 2017; pp. 1–41.
- 61. Diakoulaki, D.; Karangelis, F. Multi-criteria decision analysis and cost–benefit analysis of alternative scenarios for the power generation sector in Greece. *Renew. Sustain. Energy Rev.* **2007**, *11*, 716–727. [CrossRef]
- 62. Georgopoulou, E.; Lalas, D.; Papagiannakis, L. A multicriteria decision aid approach for energy planning problems: The case of renewable energy option. *Eur. J. Oper. Res.* **1997**, *103*, 38–54. [CrossRef]
- 63. Vögele, S.; Rübbelke, D.; Mayer, P.; Kuckshinrichs, W. Germany's "No" to carbon capture and storage: Just a question of lacking acceptance? *Appl. Energy* **2018**, 214, 205–218. [CrossRef]
- 64. Mytilinou, V.; Lozano-Minguez, E.; Kolios, A. A framework for the selection of optimum offshore wind farm locations for deployment. *Energies* **2018**, *11*, 1855. [CrossRef]
- 65. Shen, J.; Zhang, X.; Wang, J.; Cao, R.; Wang, S.; Zhang, J. Optimal operation of interprovincial hydropower system including Xiluodu and local plants in multiple recipient regions. *Energies* **2019**, *12*, 144. [CrossRef]
- 66. Özcan, E.; Yumuşak, R.; Eren, T. Risk Based Maintenance in the Hydroelectric Power Plants. *Energies* **2019**, *12*, 1502. [CrossRef]
- 67. Mateo, J.R.S.C. Multi Criteria Analysis in the Renewable Energy Industry; Springer: London, UK, 2012.
- 68. Wang, J.-J.; Jing, Y.-Y.; Zhang, C.-F.; Zhao, J.-H. Review on multi-criteria decision analysis aid in sustainable energy decision-making. *Renew. Sustain. Energy Rev.* **2009**, *13*, 2263–2278. [CrossRef]
- 69. Hwang, L.; Koon, Y. Multiple Attribute Decision Making: Methods and Applications; Springer: New York, NY, USA, 1981.
- 70. Behzadian, M.; Otaghsara, S.K.; Yazdani, M.; Ignatius, J. A state-of the-art survey of TOPSIS applications. *Expert Syst. Appl.* **2012**, 39, 13051–13069. [CrossRef]
- 71. Dodgson, J.S.; Spackman, M.; Pearman, A.; Phillips, L.D. *Multi-Criteria Analysis: a Manual*; Department for Communities and Local Government: London, UK, 2009; pp. 1–168.
- 72. Doukas, H.; Karakosta, C.; Psarras, J. Computing with words to assess the sustainability of renewable energy options. *Expert Syst. Appl.* **2010**, *37*, 5491–5497. [CrossRef]
- 73. Choudhary, D.; Shankar, R. An STEEP-fuzzy AHP-TOPSIS framework for evaluation and selection of thermal power plant location: A case study from India. *Energy* **2012**, 42, 510–521. [CrossRef]
- 74. Guo, S.; Zhao, H. Optimal site selection of electric vehicle charging station by using fuzzy TOPSIS based on sustainability perspective. *Appl. Energy* **2015**, *158*, 390–402. [CrossRef]
- 75. Erbaş, M.; Kabak, M.; Özceylan, E.; Çetinkaya, C. Optimal siting of electric vehicle charging stations: A GIS-based fuzzy Multi-Criteria Decision Analysis. *Energy* **2018**, *163*, 1017–1031. [CrossRef]
- 76. Büyüközkan, G.; Güleryüz, S. Evaluation of Renewable Energy Resources in Turkey using an integrated MCDM approach with linguistic interval fuzzy preference relations. *Energy* **2017**, *123*, 149–163. [CrossRef]
- 77. Lin, M.-C.; Wang, C.-C.; Chen, M.-S.; Chang, C.A. Using AHP and TOPSIS approaches in customer-driven product design process. *Comput. Ind.* **2008**, *59*, 17–31. [CrossRef]
- 78. Dağdeviren, M.; Yavuz, S.; Kılınç, N. Weapon selection using the AHP and TOPSIS methods under fuzzy environment. *Expert Syst. Appl.* **2009**, *36*, 8143–8151. [CrossRef]

Energies **2019**, 12, 3212 27 of 27

79. Ghosh, D.N. Analytic hierarchy process & TOPSIS method to evaluate faculty performance in engineering education. *Dipendra Nath Ghosh et al UNIASCIT* **2011**, *1*, 63–70.

- 80. DBEIS. Historical Electricity Data: 1920 to 2017. Available online: https://www.gov.uk/government/statistical-data-sets/historical-electricity-data (accessed on 17 October 2018).
- 81. Raddant, M. *The Response of European Stock Markets to the Brexit*; Kiel Institute for the World Economy (IfW): Kiel, Germany, 2016; pp. 1–13.
- 82. Brans, J.P.; Vincke, P.; Mareschal, B. How to select and how to rank projects: The Promethee method. *Eur. J. Oper. Res.* **1986**, 24, 228–238. [CrossRef]
- 83. Behzadian, M.; Kazemzadeh, R.B.; Albadvi, A.; Aghdasi, M. PROMETHEE: A comprehensive literature review on methodologies and applications. *Eur. J. Oper. Res.* **2010**, 200, 198–215. [CrossRef]
- 84. Mareschal, B. Visual PROMETHEE 1.4 Manual; VP Solutions: Brussels, Belgium, 2013; pp. 1–192.
- 85. Brans, J.-P.; Mareschal, B. PROMETHEE methods. In *Multiple Criteria Decision Analysis: State of the Art Surveys*; Springer: New York, NY, USA, 2005; pp. 163–186.
- 86. Geldermann, J.; Lerche, N. Leitfaden zur Anwendung von Methoden der multikriteriellen Entscheidungsunterstützung: Methode Promethee; Georg-August Universität Göttingen: Göttingen, Germany, 2014; pp. 1–69.



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