

Article

Explaining Global Trends in Cattle Population Changes between 1961 and 2020 Directly Affecting Methane Emissions

Katarzyna Kozicka ¹, Jan Žukovskis ² and Elżbieta Wójcik-Gront ^{1,*}

¹ Department of Biometry, Institute of Agriculture, Warsaw University of Life Sciences–SGGW, Nowoursynowska 159, 02-776 Warsaw, Poland; katarzyna_kozicka@sggw.edu.pl

² Department of Business and Rural Development Management, Vytautas Magnus University, 53361 Kaunas, Lithuania; jan.zukovskis@vdu.lt

* Correspondence: elzbieta_wojcik_gront@sggw.edu.pl

Abstract: Methane (CH₄) emissions from agricultural sources contribute significantly to the total anthropogenic greenhouse gas emissions, which cause climate change. According to the guidelines of the International Panel on Climate Change (IPCC) for calculating greenhouse gas emissions, agriculture is responsible for approximately 10% of total CH₄ emissions from anthropogenic sources. CH₄ is primarily emitted from livestock farming, particularly from cattle production during enteric fermentation and from manure. This article describes the results of multivariate statistical analyses carried out on data collected from 1961 to 2020 for thirty countries with the largest cattle populations. The study evaluated the trends in temporal changes in cattle populations and identified groups of countries with similar patterns during the study period. The global cattle population was highly correlated with CH₄ emissions from the enteric fermentation of cattle and their manure. The countries experiencing the largest increase in cattle population were primarily developing countries located in South America, Africa and Southeastern Asia. The cattle population in these countries showed a strong correlation with the human population. On the other hand, the countries where the cattle population remained stable during the study period were mainly highly developed countries. The correlations between most of the examined variables associated with cattle production and the cattle population in these countries were inconsistent and relatively weak. In the near future, further increase in the cattle population and the associated CH₄ emissions are expected, mainly in developing countries with high population growth.

Keywords: enteric fermentation; agriculture; long-term changes; multivariate relationships



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1. Introduction

One of the main greenhouse gases (GHGs) that contribute to global warming and climate change, alongside carbon dioxide and nitrous oxide, is methane (CH₄) [1]. Despite being present in the atmosphere in smaller quantities than carbon dioxide, CH₄ has a 100-year global warming potential 25 times greater than carbon dioxide due to its higher ability to absorb infrared radiation [2,3]. CH₄ is released from various sources, including landfills, waste management, energy production from coal, oil, and natural gas mining and processing [4]. It is also associated with agricultural practices. The concentration of CH₄ in the atmosphere has increased 2.5 times since pre-industrial times, primarily due to the intensive use of fossil fuels and the growth of ruminant farming, landfills, and rice fields, in line with the expansion of the human population [3,5]. Agricultural sector emissions account for approximately 25% of total global anthropogenic emissions, with direct emissions from agriculture estimated to constitute about 10–12% of total global GHG emissions in 2010 [6,7]. Additional indirect emissions result from deforestation, energy use, and the production of animal feed [8]. Livestock, particularly ruminants such as cattle, contribute the majority of direct agricultural emissions [9,10]. Therefore, reducing livestock emissions is crucial for achieving ambitious global mitigation targets [11,12].

The International Panel on Climate Change (IPCC) provides guidelines for estimating livestock emissions [2,13]. Animals are typically categorized by species because the type of digestive system significantly influences CH₄ emissions. Ruminant species such as cattle are the main source of CH₄ emissions due to their intensive food fermentation [14]. CH₄ emissions from manure management are usually lower than those from enteric fermentation [15]. Under anaerobic conditions, manure decomposition leads to substantial CH₄ production [2].

As estimated emissions are directly proportional to the cattle population (emission = emission factor $\times$ number of cattle) [14], the countries with the highest cattle population are the primary contributors to methane emissions from agricultural sources. The main regions for cattle production are South and North America, as well as Southeastern Asia. Cattle production varies in intensity and efficiency across different regions [16]. Developing countries often have lower productivity in terms of milk and beef, resulting in higher CH₄ emissions per unit of milk or beef compared to developed countries. However, developing countries may have lower CH₄ emissions per head of cattle due to less intensive production, including poorer nutrition. For example, the annual milk yield per cow in the US is approximately six times higher than in India or Pakistan [17]. GHG output (kg of CO₂ equivalents per kg of milk) ranges from 1.3 for developed countries like the USA to 7.4 for central African countries. The same level of milk or beef production can be achieved with a lower cattle population and higher production efficiency or with a higher cattle population and lower production intensity. The growing world population necessitates increased food production, including milk and beef, which can be achieved by increasing the cattle population or improving efficiency. Despite the gradual shift towards plant-based diets, the global demand for milk and beef continues to rise. Therefore, it is crucial to maintain sustainable cattle production [18]. Milk and beef production are closely connected, with dairy-beef accounting for 45% of global beef production, depending on the region [9]. The specific conditions of cattle production in different regions lead to varying changes in the cattle population, influenced by production intensity and the demand for milk and beef. On-farm practices aimed at CH₄ mitigation are more likely to focus on reducing emissions per unit of milk or meat rather than individual animal emissions [19]. Mitigation strategies that do not hinder production while effectively reducing CH₄ emissions in cattle are necessary. In practice, sustainable cattle production should be economically viable, ensuring high efficiency and low emissions per unit of production [20]. Previous studies have demonstrated that increased livestock production contributes to higher CH₄ emissions unless effective strategies to mitigate GHG emissions in livestock systems are implemented [21].

The primary objective of this study is to analyze worldwide trends in cattle populations. Methane emissions and cattle population trends are closely interconnected, as the size and management of cattle populations directly impact methane emissions from the livestock sector. An increasing cattle population generally leads to higher methane emissions. As more cattle are raised for meat and dairy production, the overall methane output from enteric fermentation and manure management tends to rise. Thus, the investigation goals are to identify countries that exhibit similar trends in cattle populations over the past 60 years (1961–2020) and examine the factors associated with these trends. It is expected that countries will fall into categories of growing, stable or declining cattle populations, influenced by various factors such as economic conditions, government policies, environmental concerns and shifts in consumer preferences. Furthermore, this research aims to identify specific variables related to cattle breeding that can effectively characterize the selected groups of countries. The study also includes a comprehensive analysis of CH₄ emissions specifically attributed to livestock through enteric fermentation and manure management.

2. Materials and Methods

Data from The Food and Agriculture Organization Corporate Statistical Database (FAOSTAT) [22] spanning the period from 1961 to 2020 were utilized to examine shifts in the global cattle population. The analysis focused on 30 countries that had existed since 1961 and possessed a cattle population of at least 10 million heads. These 30 countries include: Argentina (ARG), Australia (AUS), Bangladesh (BGD), Bolivia (BOL), Brazil (BRA), Burkina Faso (BFA), Canada (CAN), Chad (TCD), China (CHN), Colombia (COL), France (FRA), Germany (DEU), India (IND), Indonesia (IDN), Kenya (KEN), Mali (MLI), Mexico (MEX), Myanmar (MMR), New Zealand (NZL), Niger (NER), Nigeria (NGA), Pakistan (PAK), Paraguay (PRY), South Africa (ZAF), Turkey (TUR), Uganda (UGA), United Republic of Tanzania (TZA), the United States of America (USA), Uruguay (URY), Venezuela (VEN) (Figure 1). The combined cattle population of these countries accounted for over 70% of the global cattle population in 2020 [22]. Therefore, the trends observed in these analyzed countries will significantly influence the overall trends in the global cattle population.

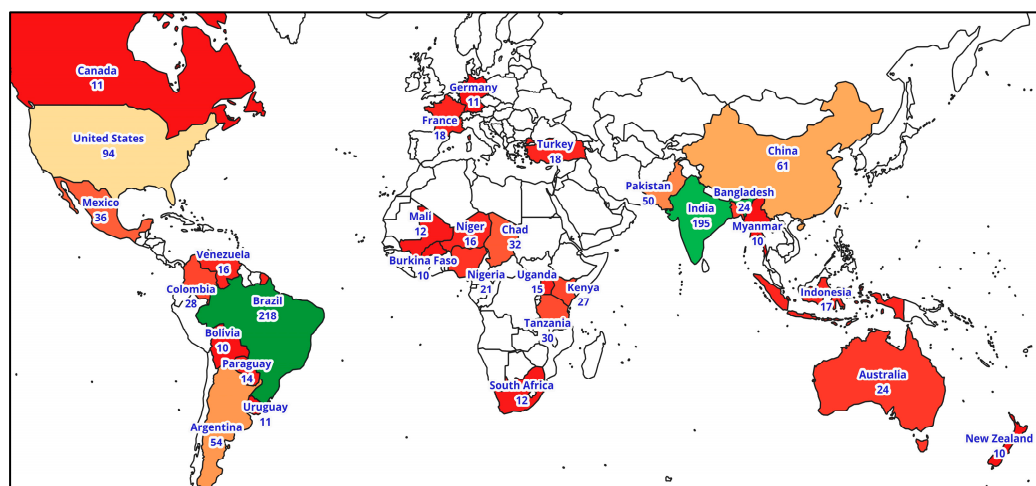


Figure 1. Countries selected for the analyses and their cattle population in millions of heads in 2020.

The data analyzed included the following variables related to the cattle population: size of the cattle population (CT), agricultural land (AL), farm machinery (FM), GDP per capita (GDP), land under permanent meadows and pastures (LMP), beef consumption per capita (MBC), total meat consumption per capita (MTC) including fish and seafood, milk consumption per capita (MC), milk yield per animal (MYA), rural population percent (RPP), total population (TP) and two ratios based on cattle population, cattle/agricultural land (CT/AL) and cattle/total population (CT/TP). The data also included CH₄ emissions from cattle enteric fermentation and manure management (CH₄).

To compare trends in the size of a country's cattle population, an increment (I) was used instead of absolute numbers of animals. The increment is calculated using the formula:

$$I = \frac{y_{i+1} - y_i}{y_i}$$

where i represents the decade number, starting from the first decade of the analysis (1961–1970) denoted by y_i . The last decade is 2011–2020.

The data on CH₄ emissions from enteric fermentation and manure management were obtained from the FAOSTAT database. The calculations were performed using the Tier 1 method, separately for dairy cattle and non-dairy cattle [2,22]. The Tier 1 method, as outlined in the 2006 IPCC guidelines, is a simplified approach for estimating CH₄ emissions from enteric fermentation and manure management. It provides a basic methodology that can be applied at the country level, taking into account factors such as livestock population, feed intake, CH₄ conversion rates and regional characteristics like climate region or temper-

ature. The FAOSTAT database provides CH₄ emission data from enteric fermentation and manure management by country, covering the period from 1961 to 2020. CH₄ emissions from enteric fermentation are a significant component of the overall GHG emissions from the agricultural sector. The emissions factors (EFs) values for enteric fermentation depend on the livestock type (dairy cattle and non-dairy cattle) and regional grouping specified in IPCC guidelines, Table 10.11 [2]. The EF values for manure management assigned to each country depend on the region and the country average annual temperature. The EF values applied for cattle were taken from IPCC Table 10.14 [2]. The methane emission factors from enteric fermentation and manure management used are presented in Table 1.

Table 1. The methane emission factors from enteric fermentation and manure management used in the study.

	CH ₄ Emission Factor [kg head ^{−1} per year]		Countries Using Presented Value
	Enteric Fermentation	Manure Management	
dairy	46	1 (2 TUR)	BFA, KEN, MLI, NER, NGA, TCD, TUR, TZA, UGA, ZAF—Countries of Africa and Middle East
non-dairy	31	1	
dairy	58	5	BGD, IND, PAK Countries of Asia
non-dairy	27	2	
dairy	68	9 CHN, 27 IDN, 23 MMR	CHN, IDN, MMR Countries of Asia
non-dairy	47	1	
dairy	72	1 (2 VEN)	ARG, BOL, BRA, COL, MEX, PRY, URY, VEN Latin America and Caribbean
non-dairy	56	1	
dairy	90	23 NZL, 29 AUS	AUS, NZL Countries of Oceania
non-dairy	60	1 NZL, 2 AUS	
dairy	117	21 DEU, 22 FRA	DEU, FRA Countries of Europe
non-dairy	57	6 DEU, 7 FRA	
dairy	128	48	CAN, USA Countries of North America
non-dairy	53	1	

Cattle production varies across regions of the world due to various factors such as climate, geography, cultural practices, and economic conditions. The IPCC methodology recognizes that different regions exhibit distinct characteristics in cattle production, which in turn influence the values of emission factors used. In Africa extensive grazing systems are common, with cattle often raised in open pasturelands. Many cattle breeds are adapted to withstand heat and tropical diseases. In Asia diverse cattle production systems exist, including intensive, semi-intensive and extensive systems. In Europe cattle production systems vary from intensive indoor systems to extensive grazing on pasturelands. Dairy farming is a significant focus, with specialized dairy breeds and high milk yields. In Latin America and Caribbean extensive grazing systems and large-scale ranching are common, especially in countries like Brazil and Argentina. In North America cattle production involves a mix of intensive feeding and extensive grazing systems. Dairy farming is also significant, particularly in the United States and Canada. In Oceania, cattle production revolves around extensive grazing systems on large pasturelands. Dairy farming is significant in countries like Australia and New Zealand.

The analysis of obtaining groups with homogeneous countries in terms of the cattle change trend was performed using cluster analysis. This approach facilitates the identification of essential features based on the population trend analysis of each group. Ward's method, which is based on a variance approach, was applied in the cluster analysis as it is considered very effective [23]. The square of the Euclidean distance was used to calculate

the multivariate distance between objects, giving more weight to objects that are farther apart. Correlation coefficients were used to evaluate the relationships between selected variables and the cattle population or the cattle population per agricultural land or per human population. Regression analysis was also performed to evaluate the temporal trends of the cattle population, as well as the cattle population per agricultural land or per human population. Additionally, principal component analysis (PCA) was employed to assess the multivariate differences between the studied countries and the relationships between variables included in the study. The results of PCA were presented graphically as a biplot. The analyses were conducted using Statistica 13 (Tibco Software Inc., Palo Alto, CA, USA). The significance level for all the tests was set at 0.05.

3. Results

3.1. Temporal Trends in Cattle Population in Period 1961–2020

In 1961, the global cattle population was approximately 942 million. In 2020, it had reached around 1523 million heads. When plotting the changes in the number of cattle over time, the average annual increase is approximately 8.3 million and can be well described by a linear function ($R^2 = 0.95$) (Figure 2a). The increase in the cattle population correlates with the rise in cattle density, represented as the number of cattle heads per 1000 ha of agricultural land. However, the growth rate of the cattle population (about 62% during the study period) surpassed the increase in the cattle-to-agricultural land ratio (about 51%), as depicted in Figure 2. The number of cattle heads per 1000 people exhibited a linear decrease ($R^2 = 0.98$) during the study period, declining from 307 to 194 (a reduction of approximately 37%). The downward trend in recent years has been slower. CH_4 emission associated with cattle production, including both enteric fermentation and manure, were strongly correlated with the cattle population (Figure 2b). The average yearly global increase in CH_4 emissions during the study period amounted to 0.34 million tons. The relationship between cattle population and methane emission from cattle worldwide was nearly linear (Figure 2c). Based on the regression analysis, it was determined that an increase in the cattle population by one head results in an average annual increase in CH_4 emissions of 42.7 kg.

Temporal trends of the cattle population and its ratio per agricultural land or per number of people varied significantly among different countries. To evaluate these changes from 1961 to 2020, the means for decades (1961–1970 and 2011–2020) were calculated. Decade means were used because the values for individual years were highly variable in some countries, such as Germany, where recent data exhibited significant year-to-year variability. The changes between the first decade (1961–1970) and the last decade (2011–2020) are presented in Table 2. The highest increase in the cattle population was observed in Chad (488% higher cattle population in the last decade compared to the first decade). Bolivia and Burkina Faso also experienced increases of over 300% in their cattle population, while Brazil, Niger, Paraguay and Uganda saw increases in the range of 200–300%. Most of the studied countries exhibited an increase in their cattle population, with only three countries experiencing a decrease: Germany (−33%), France (−9%) and the USA (−14%). The area of agricultural land remained relatively stable over time, and the ratio of cattle population to agricultural area was generally higher in the last decade (1961–1970) compared to the first decade (2011–2020).

The number of cattle per 1000 people decreased in most countries, with the strongest decreases observed in Bangladesh (−67%), India (−63%), Turkey (−61%) and South Africa (−66%). Only four countries showed an increase in the cattle population-to-number of people ratio: Burkina Faso (39%), Bolivia (60%), Brazil (10%) and Chad (64%).

To identify groups of countries with similar patterns of cattle population changes, a cluster analysis was conducted. The analysis used mean increments calculated for subsequent decades (1961–1970, . . . , 2011–2020) based on the formula presented in the Material and Methods section. Since data for the first decade (1961–1970) did not have associated data for the previous decade (1951–1960), five variables were used for the analysis, with the first variable representing the decade 1971–1980 and the last variable

representing the decade 2011–2020. The cluster analysis identified four groups of countries, as shown on the dendrogram in Figure 3. The patterns of changes in cattle population over time are presented in Figure 4, and the groups of countries are displayed on the map in Figure 5, along with the percentage change in cattle population between 1961–1970 and 2011–2020. The first group of countries consisted of four countries from central Africa, Burkina Faso, Mali, Niger, Uganda, and one country from south Asia—Pakistan. These countries experienced a high increase in cattle population, particularly in the last two decades (2001–2020). A similar pattern of cattle population changes was also observed in Chad, which was atypical due to the highest increase in cattle population throughout the entire study period, especially in the decade 1991–2000 (approximately 220%). The second group of countries included Bolivia, Brazil, Venezuela, Paraguay, Mexico (South America and the southern part of North America), Indonesia, Myanmar (Southeastern Asia), Kenya, Nigeria and Tanzania (Central Africa). These countries exhibited a relatively stable increase in the cattle population throughout the study period, with slightly higher increases in the first half compared to the second half.

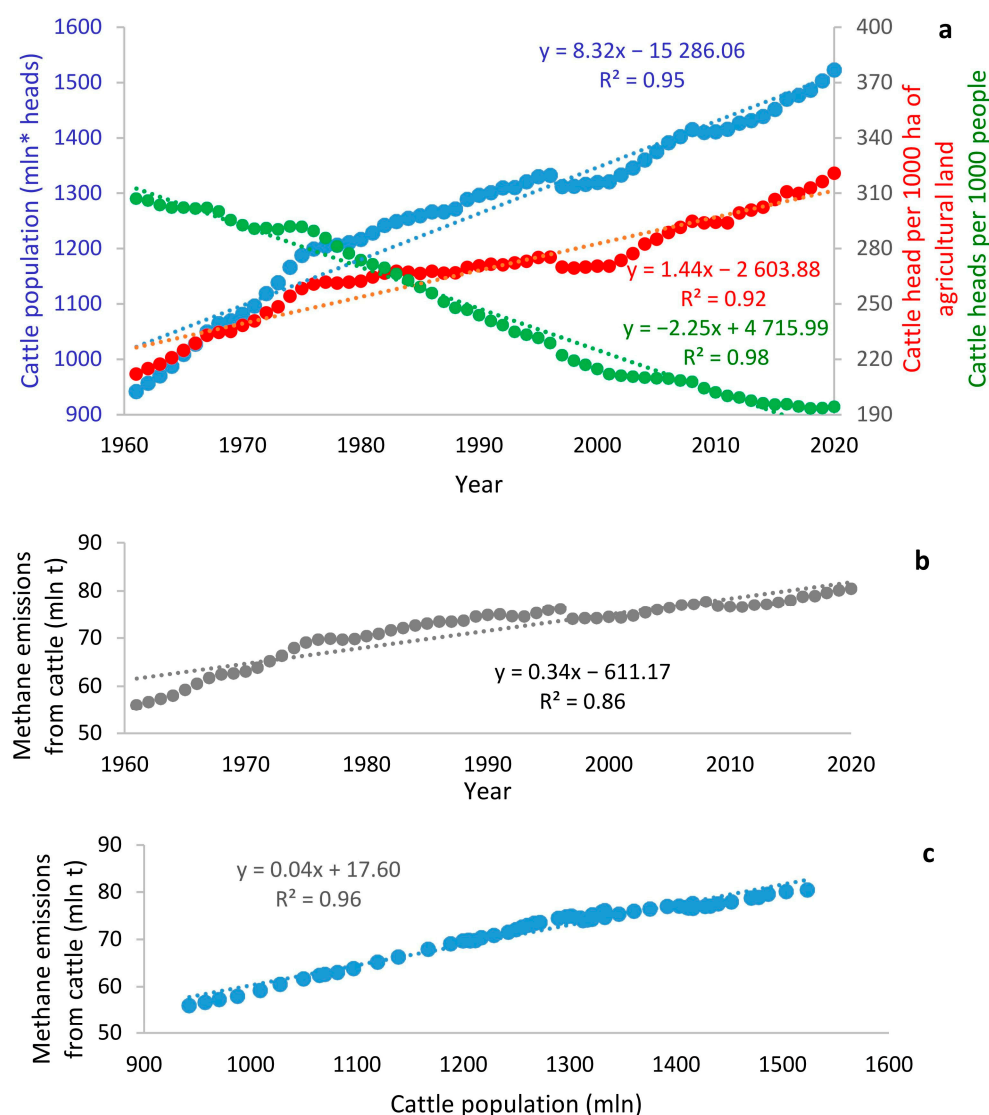


Figure 2. The cattle population (in blue) for the entire world from 1961 to 2020 along with the cattle rate per 1000 ha of agricultural land (in red) and per 1000 people (in green) (a), the global CH_4 emission from cattle (in grey), which includes both emissions from enteric fermentation and manure management (b) and relationship between the world cattle population and methane emission from cattle (c). * mln—million.

Table 2. The mean cattle population for studied countries during the periods 1961–1970 and 2011–2020, as well as the corresponding changes. The color of the background indicates differences between countries in the values shown in the columns.

Country	Cattle (mln * heads)			Cattle Density (heads/1000 ha of Agricultural Land)			Cattle Heads per 1000 People		
	1961–1970	2011–2020	Change **	1961–1970	2011–2020	Change	1961–1970	2011–2020	Change
Argentina	46.3	52.4	13%	351	451	29%	2497	1347	−46%
Australia	19.0	26.9	41%	39	73	88%	2082	1331	−36%
Burkina Faso	2.2	9.4	326%	269	773	188%	494	684	39%
Bangladesh	23.0	23.7	3%	2392	2505	5%	521	171	−67%
Bolivia	2.1	9.1	345%	68	243	255%	614	982	60%
Brazil	65.2	214.4	229%	376	911	142%	1052	1155	10%
Canada	11.5	11.8	2%	182	203	12%	738	367	−50%
Chad	4.4	25.8	488%	92	516	462%	1607	2632	64%
China	52.6	63.1	20%	147	120	−19%	88	49	−45%
Colombia	17.5	24.5	40%	415	532	28%	1310	585	−55%
Germany	18.4	12.3	−33%	948	735	−22%	257	151	−41%
France	20.7	18.9	−9%	614	657	7%	477	313	−34%
Indonesia	6.7	15.7	135%	174	264	52%	87	69	−20%
India	175.9	190.7	8%	993	1063	7%	445	167	−63%
Kenya	7.6	19.9	162%	301	718	139%	1162	560	−52%
Mexico	19.6	33.7	72%	200	343	71%	630	323	−49%
Mali	4.5	10.7	137%	143	260	82%	909	824	−9%
Myanmar	6.1	14.0	128%	575	1091	90%	316	294	−7%
Niger	4.0	12.6	216%	126	274	118%	1339	919	−31%
Nigeria	7.4	19.9	169%	128	290	127%	183	143	−22%
New Zealand	7.4	10.1	37%	467	941	102%	3484	2480	−29%
Pakistan	14.4	42.2	194%	394	1159	194%	352	245	−30%
Paraguay	4.4	13.7	212%	404	820	103%	2623	2517	−4%
Turkey	13.1	14.7	12%	350	385	10%	554	216	−61%
Tanzania	9.2	26.3	185%	343	683	99%	1067	674	−37%
Uganda	3.6	13.8	279%	377	960	155%	558	501	−10%
Uruguay	8.6	11.6	35%	539	814	51%	3649	3506	−4%
United States of America	106.9	92.0	−14%	244	227	−7%	668	311	−53%
Venezuela	7.3	16.2	122%	373	755	102%	1100	615	−44%
South Africa	11.7	13.3	14%	120	138	15%	807	272	−66%

* mln—million, ** Relative change between two periods, 2011–2020 and 1961–1970 (reference period).

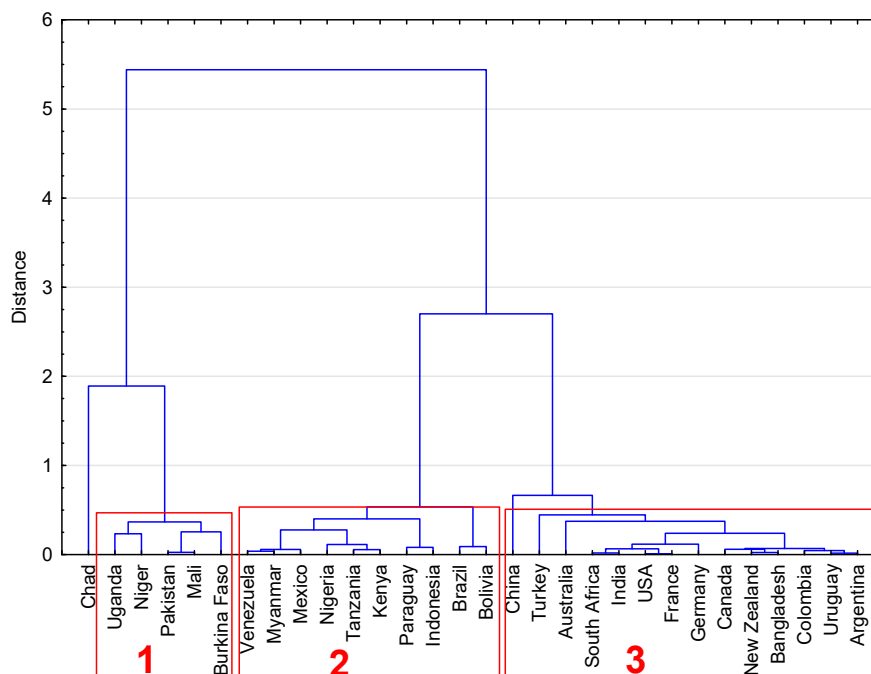


Figure 3. Cluster analysis of trends in cattle population for countries around the world based on relative values of changes for subsequent decades (1961–1970, . . . , 2011–2020).

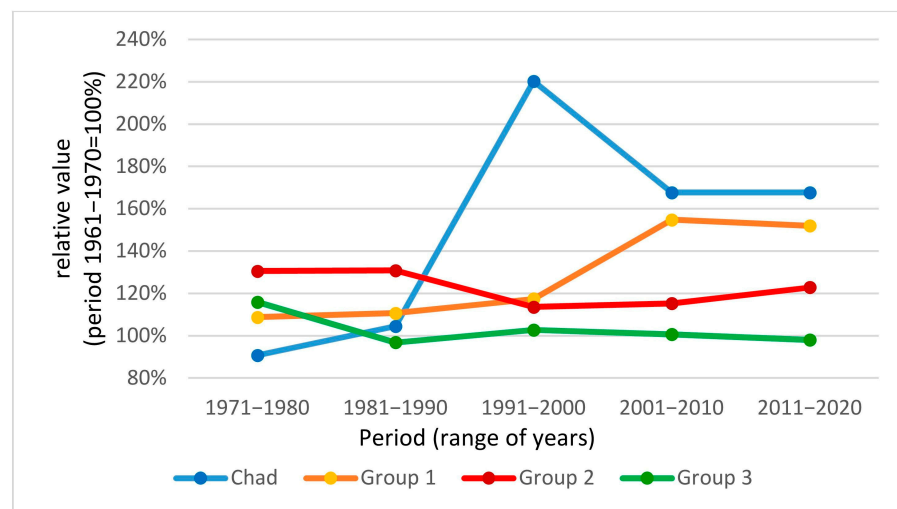


Figure 4. Means values of relative changes in cattle population for subsequent decades (1971–1980, ... , 2011–2020) compared to the previous decade for groups distinguished by cluster analysis (group of countries are presented in Figure 3).

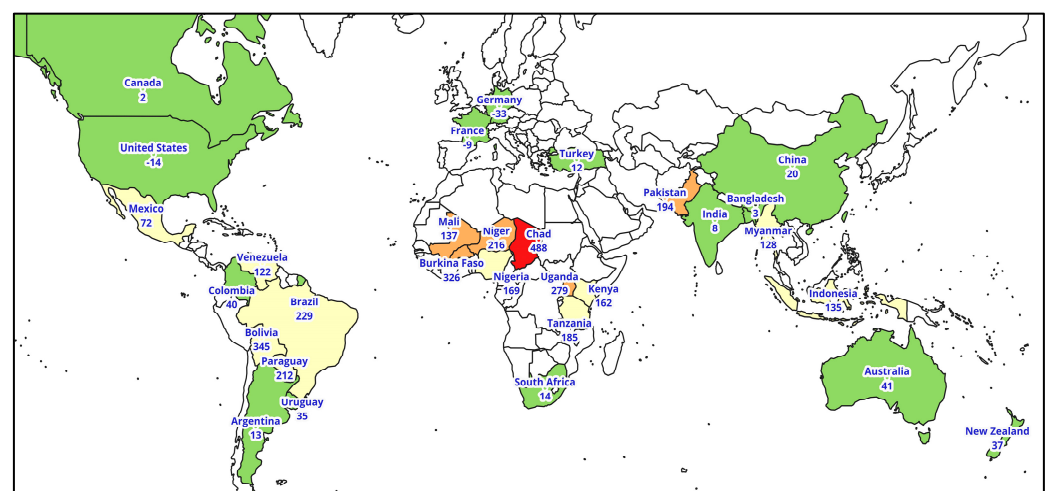


Figure 5. Maps presenting the distinguished groups of countries based on cluster analysis (Figure 3) in different colors: orange for group 1, where a strong increase in cattle population was observed; yellow for group 2, where a strong increase was observed in the first half of the study period followed by a slight increase; green for group 3, where the cattle population remained quite stable along the study period; and Chad in red, indicating a very strong increase. The values next to the country names represent the percentage change in cattle population between 1961–1970 and 2011–2020.

The third group of countries was the largest and included the following countries: Argentina, Uruguay, Colombia (South America), Canada, USA (North America), Germany, France (Europe), Bangladesh, China, India, Turkey (Asia), South Africa, New Zealand and Australia. These countries exhibited a significant increase in cattle population at the beginning of the study period, but from 1981 to 2020, the cattle population remained relatively stable or slightly decreased. The countries in this group are located in different regions of the world, with many of them being highly developed countries.

3.2. Relationship between Cattle Population and Other Variables

To evaluate the relationship between cattle population in each country and various variables that characterize agricultural production, food consumption and economic conditions, a correlation analysis was conducted using yearly data from 1961 to 2020. The results of the correlation analysis are presented in Table 3.

Table 3. Correlation coefficients between cattle population and other variables from 1961 to 2020, categorized by country groups based on cluster analysis. Positive correlations are indicated by red cells, while negative correlations are marked in blue.

Country	Group	AL ¹	FM	GDP	LMP	MBC	MTC	MC	MYA	RPP	TP	CH ₄
Chad	0	0.98 *	−0.13	0.75 *	0.00	0.85 *	0.89 *	−0.72 *	−0.80 *	−0.75 *	0.98 *	1.00 *
Burkina Faso	1	0.98 *	0.44 *	0.98 *	0.00	0.76 *	0.85 *	0.03	−0.70 *	−0.95 *	0.98 *	1.00 *
Mali	1	0.85 *	−0.24	0.87 *	0.82 *	0.43 *	0.28 *	0.31 *	−0.84 *	−0.86 *	0.95 *	1.00 *
Niger	1	0.91 *	−0.02	−0.29 *	0.88 *	−0.33 *	−0.49 *	−0.41 *	0.80 *	−0.53 *	0.92 *	1.00 *
Pakistan	1	0.02	0.92 *	0.91 *	0.00	0.97 *	0.87 *	0.06	0.92 *	−0.89 *	0.96 *	1.00 *
Bolivia	2	0.94 *	0.50 *	0.83 *	0.81 *	0.92 *	0.95 *	0.79 *	0.94 *	−0.97 *	0.98 *	1.00 *
Brazil	2	0.78 *	0.93 *	0.96 *	0.66 *	0.97 *	0.96 *	0.95 *	0.79 *	−0.99 *	0.99 *	1.00 *
Indonesia	2	0.92 *	0.94 *	0.96 *	−0.74 *	0.82 *	0.96 *	0.25	0.91 *	−0.94 *	0.94 *	1.00 *
Kenya	2	0.87 *	0.81 *	0.88 *	0.00	−0.36 *	−0.18	0.37 *	0.74 *	−0.88 *	0.92 *	0.98 *
Mexico	2	0.47 *	0.92 *	0.94 *	0.04	0.80 *	0.87 *	0.61 *	0.78 *	−0.94 *	0.88 *	1.00 *
Myanmar	2	0.73 *	0.79 *	0.74 *	−0.24	0.57 *	0.78 *	0.65 *	0.90 *	−0.78 *	0.90 *	0.85 *
Nigeria	2	0.71 *	0.64 *	0.57 *	0.23	−0.62 *	0.50 *	−0.36 *	0.17	−0.97 *	0.95 *	1.00 *
Paraguay	2	0.97 *	0.77 *	0.96 *	0.82 *	−0.68 *	−0.45 *	0.72 *	0.80 *	−0.96 *	0.98 *	1.00 *
Tanzania	2	0.94 *	−0.46 *	0.97 *	0.82 *	−0.19	−0.50 *	0.15	0.96 *	−0.86 *	0.97 *	1.00 *
Uganda	2	0.87 *	0.69 *	0.96 *	0.93 *	−0.38 *	0.33 *	0.79 *	0.53 *	−0.44 *	0.93 *	1.00 *
Venezuela	2	0.87 *	0.93 *	0.57 *	0.93 *	0.15	0.75 *	−0.15	0.05	−0.98 *	0.95 *	1.00 *
Argentina	3	−0.34 *	0.05	0.32 *	−0.39 *	−0.10	0.11	0.12	0.15	−0.42 *	0.31 *	1.00 *
Australia	3	−0.28 *	0.50 *	0.49 *	−0.28	0.14	0.59 *	−0.56 *	0.43 *	−0.56 *	0.46 *	0.99 *
Bangladesh	3	0.27 *	−0.04	−0.05	0.00	0.34 *	0.01	−0.09	0.45 *	0.17	−0.13	0.94 *
Canada	3	0.10	−0.10	−0.53 *	0.03	0.04	0.51 *	−0.05	0.25	−0.26 *	0.25	0.40 *
China	3	0.71 *	0.38 *	0.10	0.71 *	0.47 *	0.47 *	0.25	0.23	−0.24	0.56 *	0.99 *
Colombia	3	0.44 *	0.39 *	0.67 *	0.61 *	−0.38 *	0.53 *	0.39 *	0.36 *	−0.85 *	0.72 *	0.99 *
France	3	0.52 *	0.77 *	−0.54 *	0.71 *	0.80 *	0.04	0.53 *	−0.69 *	0.31 *	−0.54 *	0.95 *
Germany	3	0.79 *	0.97 *	−0.91 *	0.78 *	0.94 *	0.13	−0.05	−0.88 *	0.69 *	−0.59 *	0.99 *
India	3	0.81 *	0.33 *	0.32 *	−0.72 *	−0.16	0.51 *	0.69 *	0.43 *	−0.63 *	0.54 *	0.85 *
New Zealand	3	−0.80 *	0.81 *	0.93 *	−0.70 *	−0.57 *	−0.04	−0.51 *	0.81 *	−0.77 *	0.85 *	0.96 *
South Africa	3	0.21	−0.57 *	0.30 *	0.07	−0.35 *	0.42 *	−0.63 *	0.54 *	−0.61 *	0.64 *	1.00 *
Turkey	3	−0.65 *	−0.18	0.12	−0.32 *	0.30 *	0.10	0.75 *	0.08	0.12	−0.07	0.98 *
Uruguay	3	−0.75 *	0.19	0.74 *	−0.64 *	−0.73 *	−0.68 *	0.09	0.71 *	−0.80 *	0.82 *	1.00 *
USA	3	0.67 *	0.24	−0.75 *	−0.05	0.92 *	−0.58 *	0.27 *	−0.75 *	0.60 *	−0.72 *	0.94 *

* Significant correlations at 0.05 probability level. ¹ Abbreviations used in the table: agricultural land (AL), farm machinery (FM), gross domestic product per capita (GDP), land under perm. meadows and pastures (LMP), meat beef consumption per capita (MBC), meat total (incl. fish and seafood) consumption per capita (MTC), milk consumption per capita (MC), milk yield per animal (MYA), rural population percent (RPP), total population (TP), methane emission from cattle enteric fermentation and manure management (CH₄).

For all countries from the first and second groups, as well as Chad, a very strong positive correlation was observed between the cattle population and the human population. The correlation coefficients ranged from 0.85 to 0.99, indicating that the increase in cattle population in these countries was almost linearly associated with the growth of the human population. Moreover, cattle population showed a strong positive correlation with GDP per capita and the area of agricultural land while exhibiting a negative correlation with the percentage of rural population. These significant correlations were observed for most countries from the first and second groups, although not for all of them. Other correlations within the first and second groups were less consistent. For example, an increase in cattle population was associated with an increase in milk yield per animal, but only for approximately two-thirds of the countries in these groups.

The correlations within the third group of countries were not consistent, as both negative and positive correlations with cattle populations were observed for all variables. Most of these correlations were weaker compared to those observed in countries belonging to the first and second groups.

For Canada, the correlation coefficient between changes in cattle population and changes in CH₄ emissions from cattle is positive and statistically significant, albeit lower compared to other countries. This is because the population of dairy cattle in Canada has been declining over the study period, while the number of non-dairy cattle has increased or fluctuated. Dairy cattle tend to have higher CH₄ emission factors compared to beef cattle (Table 1), which explains the weaker correlation between cattle population and CH₄ emission from cattle during the study period.

In addition to calculating correlations for each country, correlations were also calculated across all countries based on the means for the period 2011–2020. These correlations encompassed all the studied variables. In addition, two ratios: cattle-to-agricultural land and cattle-to-total human population, were included in the analysis. The results showed that the cattle population was significantly correlated only with the area of agricultural land and the total human population. These correlations were positive, indicating that larger agricultural areas are necessary to support a larger population of cattle, and a larger human population may require more animal-based food. The ratio of cattle-to-agricultural land was found to be significantly correlated with both the area of agricultural land and the area of land under permanent meadows and pastures. The correlation was negative, suggesting that countries with larger agricultural areas, including meadows and pastures, tend to have a lower cattle density per unit area. Additionally, the ratio of cattle-to-total-human population exhibited a significant correlation with beef consumption per capita. The correlation was positive, indicating that countries with higher beef consumption tend to have a higher cattle population per 1000 people. However, there was no significant correlation found with milk consumption. These relationships, as presented in Table 4, are also visualized in the form of a PCA biplot in Figure 6.

Table 4. The correlation coefficients between all studied variables in all countries based on the means for 2011–2020.

	CT	CT/AL	CT/TP	AL	FM	GDP	LMP	MBC	MTC	MC	MYA	RPP	TP	CH ₄
Cattle population (CT)		0.13	−0.11	0.50 *	0.14	0.00	0.33	0.21	0.11	0.16	0.05	−0.09	0.57 *	0.96 *
Cattle/agricultural land (CT/AL)	0.13		0.03	−0.37	−0.07	−0.26	−0.41 *	−0.27	−0.30	−0.17	−0.34	0.27	−0.02	0.07
Cattle/total population (CT/TP)	−0.11	0.03		−0.16	−0.21	0.07	−0.06	0.46 *	0.11	0.19	−0.06	−0.25	−0.32	−0.07
Agricultural land (AL)	0.50 *	−0.37	−0.16		0.56 *	0.38 *	0.96 *	0.28	0.48 *	0.25	0.36 *	−0.19	0.65 *	0.56 *
Farm machinery (FM)	0.14	−0.07	−0.21	0.56 *		0.21	0.52 *	−0.10	0.31	0.08	0.24	−0.10	0.70 *	0.16
GDP per capita (GDP)	0.00	−0.26	0.07	0.38 *	0.21		0.40 *	0.56 *	0.79 *	0.80 *	0.88 *	−0.66 *	−0.08	0.13
Land under perm. meadows and pastures (LMP)	0.33	−0.41	−0.06	0.96 *	0.52	0.40 *		0.34	0.54 *	0.27	0.33	−0.24	0.47 *	0.43 *
Meat beef consumption per capita (MBC)	0.21	−0.27	0.46	0.28	−0.10	0.56 *	0.34		0.73 *	0.74 *	0.62 *	−0.75 *	−0.27	0.36 *
Meat total (incl. fish and seafood) consumption per capita (MTC)	0.11	−0.30	0.11	0.48 *	0.31	0.79 *	0.54 *	0.73 *		0.68 *	0.79 *	−0.74 *	−0.02	0.29
Milk consumption per capita (MC)	0.16	−0.17	0.19	0.25	0.08	0.80 *	0.27	0.74 *	0.68 *		0.78 *	−0.81 *	−0.16	0.29
Milk yield per animal (MYA)	0.05	−0.34	−0.06	0.36 *	0.24	0.88 *	0.33	0.62 *	0.79 *	0.78 *		−0.70 *	0.00	0.18
Rural population percent (RPP)	−0.09	0.27	−0.25	−0.19	−0.10	−0.66 *	−0.24	−0.75 *	−0.74 *	−0.81 *	−0.70 *		0.17	−0.23
Total population (TP)	0.57 *	−0.02	−0.32	0.65 *	0.70	−0.08	0.47 *	−0.27	−0.02	−0.16	0.00	0.17		0.46 *
Methane emission (CH ₄)	0.96 *	0.07	−0.07	0.56 *	0.16	0.13	0.43 *	0.36 *	0.29	0.29	0.18	−0.23	0.46 *	

* Significant correlations at 0.05 probability level.

Positive correlations can be observed across countries for the following variables: GDP per capita (GDP), land under perm. meadows and pastures (LMP), meat beef consumption per capita (MBC), meat total (incl. fish and seafood) consumption per capita (MTC), milk consumption per capita (MC), milk yield per animal (MYA). Conversely, these variables exhibit negative correlations with the percentage of rural population (RPP). Therefore, countries with higher GDP per capita tend to have higher meat and milk consumption per capita, higher milk yield per animal, and a lower percentage of rural population. These are the United States, Australia, Argentina, France, Canada, Brazil, Germany and New Zealand (located on the left side of the biplot in Figure 6). On the other hand, countries like Bangladesh, Uganda, Burkina Faso, Niger, Tanzania, Nigeria and Kenya (located on the right side of the biplot in Figure 6) exhibit lower meat and milk consumption per capita, lower milk yield per animal, and a higher percentage of rural population. Strong positive correlations were identified between cattle population (CP) and farm machinery (FM), methane emission attributed to cattle (CH₄), agricultural land (AL), land under perm. meadows and pastures (LMP). Notably, Brazil stands out as the country with the highest values for these variables.

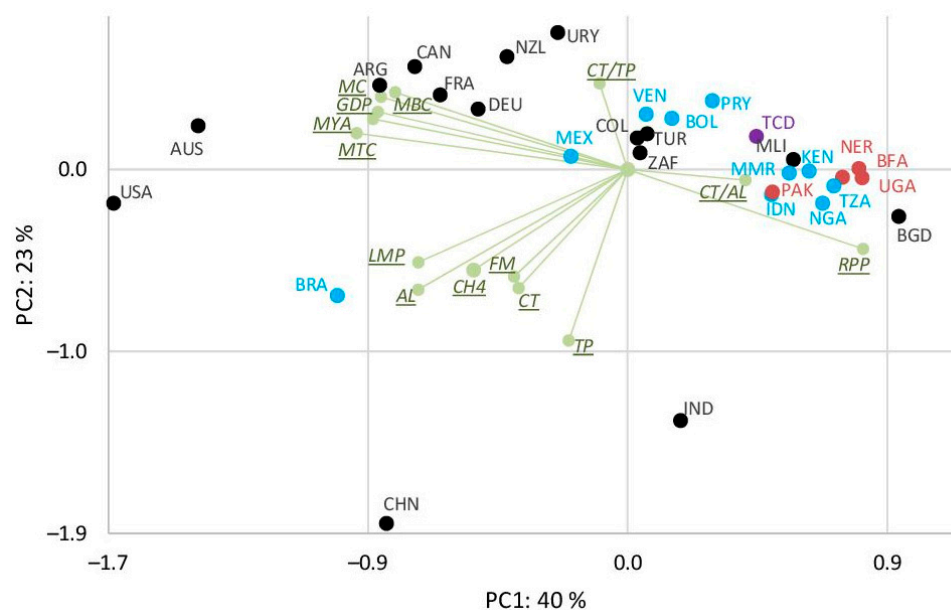


Figure 6. PCA biplot illustrating the relationships between the studied variables, as well as the multivariate differences among the countries included in the analysis, based on the means for the period 2011–2020. Abbreviations variables (marked in green and underlined): the size of the cattle population (CT) in the country and other variables which can be related to the cattle population: agricultural land (AL), farm machinery (FM), GDP per capita (GDP), land under perm. meadows and pastures (LMP), beef consumption per capita (MBC), meat total (incl. fish and seafood) consumption per capita (MTC), milk consumption per capita (MC), milk yield per animal (MYA), rural population percent (RPP), total population (TP) and two ratios based on cattle population, cattle/agricultural land (CT/AL), cattle/total population (CT/TP), methane emission connected with cattle (CH₄); countries: Argentina (ARG), Australia (AUS), Bolivia (BOL), Brazil (BRA), Burkina Faso (BFA), Canada (CAN), Chad (TCD), China (CHN), Colombia (COL), France (FRA), Germany (DEU), India (IND), Indonesia (IDN), Kenya (KEN), Mali (MLI), Mexico (MEX), Myanmar (MMR), New Zealand (NZL), Niger (NER), Nigeria (NGA), Paraguay (PRY), South Africa (ZAF), Turkey (TUR), Uganda (UGA), United Republic of Tanzania (TZA), the United States of America (USA), Uruguay (URY), Venezuela (VEN). Different colors of dots for countries indicate groups distinguished in cluster analysis (Figure 3).

4. Discussion

This study focused on examining the contribution of livestock systems to global warming by analyzing the emissions directly and unambiguously attributed to livestock. The analysis found a strong correlation between cattle population and CH₄ emission from cattle, with a correlation coefficient of 0.96 across countries for the last decade (2011–2020). The temporal pattern of changes in cattle population and CH₄ emissions at a global scale exhibited a similar trend.

These CH₄ emission estimations are based on Tier 1 factors, which are less detailed and may introduce biases. These factors consider regional differences in production intensity and categorize cattle into dairy and non-dairy types. To obtain more accurate emission factors, the Tier 2 method is used, which takes into account specific characteristics and activities of different livestock groups [2]. This approach considers factors like animal characteristics, diet, housing conditions, manure management practices, and other relevant parameters. By incorporating these factors, Tier 2 provides a more precise estimation of greenhouse gas emissions compared to default values. Calculating emission factors using the Tier 2 method requires detailed activity data specific to livestock categories. This data includes information on animal numbers, production parameters, feed consumption, manure management practices and other factors that influence emissions. Studies on CH₄ emissions from enteric fermentation and manure management have shown variations in

emission factor values, typically around 20%. For example, the UNFCCC (the United Nations Framework Convention on Climate Change) inventory reports indicate that the US reported an enteric fermentation CH_4 emission factor for dairy cows of $121 \text{ kg CH}_4 \text{ head}^{-1}$ in 1990 and $149 \text{ kg CH}_4 \text{ head}^{-1}$ in 2020. However, FAOSTAT uses a value of 128 for dairy cows in the US. Similar variations exist in the case of manure management. Notably, CH_4 emission rates for manure management can be significantly lower than those for enteric fermentation. The structure of cattle populations and rearing methods also influence CH_4 emissions. Although presented study did not account for this structure due to data limitations, its main objective was to demonstrate global trends in cattle population changes and their implications, such as CH_4 emissions from livestock.

Presented findings revealed different patterns of temporal changes in cattle populations among groups of countries. Two groups exhibited a strong correlation between cattle and human population growth, mainly in developing or middle-income countries located in Africa, South America, and Southeastern Asia. These countries showed a substantial increase in both cattle and human populations, along with a rise in milk and meat consumption [24,25]. However, their milk and meat consumption levels still remain lower than those in developed countries. The third group consisted primarily of highly developed countries, where cattle populations remained relatively stable, and an increase in cattle production efficiency was observed.

Over the study period (1961–2020), the global human population increased by approximately 155%, from 3.07 to 7.84 billion people, while the cattle population increased by about 62% [26]. This raises the question of whether increasing the cattle population is necessary to meet the growing food demands. It is possible to produce more beef and milk with the same cattle population by enhancing production efficiency. A notable example is the US, where the human population increased by over 100% during the same period, yet the cattle population either remained stable or slightly decreased. Such improvements in cattle production efficiency are beneficial for reducing CH_4 emissions as they decrease the emissions per unit of protein produced [16,27]. In many developing countries, CH_4 emissions per unit of production are still very high, and there is high potential for increased intensity of beef and milk production to reduce CH_4 emissions. Developing countries, especially those in Southeastern Asia and sub-Saharan Africa, still have high CH_4 emissions per unit of production, suggesting significant potential for emission intensity reduction through increased efficiency [16]. Productivity gains are particularly crucial for regions experiencing high population growth, as is the case for many developing countries.

In tropical climates, where many developing countries are located, the same livestock management practices used in developed countries may not be applicable. However, one potential approach to reduce CH_4 emissions during cattle production in tropical climates is through crossbreeding, which has the potential to improve performance [28]. A study by Haas et al. [29] demonstrated that genetic progress can reduce the intensity of CH_4 emissions (CH_4 emitted per kg of milk) by approximately 20% over the next 30 years in European conditions.

One challenge associated with improving cattle production efficiency is the negative effect of heat stress, particularly on dairy cows, leading to decreased milk production [30]. Heat stress also diminishes the efficiency of meat production [31,32]. Unfortunately, continuous climate warming exacerbates heat stress in cattle production, posing a significant obstacle to increasing production efficiency, especially in tropical climates.

Various methods can be employed to mitigate global warming by reducing CH_4 emissions in cattle production. These methods include improved grazing management, dietary modifications and nutrition for livestock, genetic improvement, better manure management [10,33]. A simple strategy for cattle producers to reduce CH_4 emissions is to adopt the practices currently used by leading producers with the lowest emission intensity. While most studies on CH_4 reduction in cattle focus on changes in enteric emissions but efforts should encompass a more comprehensive approach that includes other GHG emissions associated with cattle production [34].

Changes in livestock CH₄ emissions were primarily influenced by shifts in human population dynamics. However, in highly developed countries, emissions have been reduced through increased efficiency in cattle production. The current global challenges related to increased CH₄ emissions from cattle production are primarily concentrated in developing countries, where cattle production efficiency remains low despite growing demands for food due to population growth [35,36]. Our study, along with other research [35], has identified an ongoing increase in CH₄ emissions in regions like South Asia, tropical Africa and Brazil, driven by the expansion of cattle populations and low production efficiency. Highly developed countries still have the potential to reduce CH₄ emission from cattle production, although this potential is comparatively lower than that of developing countries, mainly due to stable human populations [37,38].

5. Conclusions

During the period from 1961 to 2020, the increase in human population was the primary driver behind the rise in cattle population in less developed countries, predominantly located in Africa and South America. Conversely, developed countries experienced relatively stable cattle populations, but notable improvements in cattle production efficiency were observed, such as higher milk yield per animal. Since methane emission is strongly correlated with cattle population, there is significant potential for mitigating CH₄ emissions from cattle production, particularly in developing countries. These regions offer favorable conditions for introducing more efficient cattle management, which can lead to higher beef and milk production while maintaining a similar cattle population.

In planning for future changes in milk and beef production, it is crucial to prioritize achieving higher production efficiency. This can be accomplished by increasing production intensity while ensuring the well-being of the animals. By focusing on both efficiency and animal welfare, it is possible to meet the growing demands for milk and beef while minimizing the environmental impact associated with methane emissions.

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