

Article

## Estimation of the Image Interpretability of ZY-3 Sensor Corrected Panchromatic Nadir Data

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**Abstract:** Image quality is important for taking full advantage of satellite data. As a common indicator, the National Imagery Interpretability Scale (NIIRS) is widely used for image quality assessment and provides a comprehensive representation of image quality from the perspective of interpretability. The ZY-3 (Ziyuan-3) satellite is the first civil high resolution mapping satellite in China, which was established in 2012. So far, there has been no reports on adopting NIIRS as the common indicator for the quality assessment of that satellite image data. This lack of a common quality indicator results in a gap between satellite data users around the world and those in China regarding the understanding of the quality and usability of ZY-3 data. To overcome the gap, using the general image-quality equation (GIQE), this study evaluates the ZY-3 sensor-corrected (SC) panchromatic nadir (NAD) data in terms of the NIIRS. In order to solve the uncertainty resulting from the exceeding of the ground sample distance (GSD) of ZY-3 data (2.1 m) in GIQE (less than 2.03 m), eight images are used to establish the relationship between the manually obtained NIIRS and the GIQE predicted NIIRS. An adjusted GIQE is based on the relationship and verified by another five images. Our study demonstrates that the method of using adjusted GIQE for calculating NIIRS can be used for the quality assessment of ZY-3 satellite images and reveals that the NIIRS value of ZY-3 SC NAD data is about 2.79.

**Keywords:** image quality; interpretability; NIIRS; GIQE; linear regression analysis; ZY-3

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## 1. Introduction

The ZY-3 (Ziyuan-3) satellite, launched and operated by the Satellite Surveying and Mapping Application Center (SASMAC), National Administration of Surveying, Mapping and Geo-information (NASG), China, is the first of a new type of civil high resolution satellite in China. With a high spatial resolution camera (2.1-m ground sample distance (GSD) TDI-CCD panchromatic (PAN) camera and 5.8 m GSD multispectral (MUX) camera), image products are designed to meet the requirements of the following fields: surveying and mapping, monitoring land resources, land use and planning, agriculture, environmental monitoring and protection, traffic and other important areas. The satellite was launched on 19 January 2012, and to date, there have been 525,824 images, which correspond to approximately 352 TB in the archives, seamlessly covering areas within 84 degrees of north and south latitudes. With a revisit cycle of five days, the mission duration of ZY-3 is five years [1].

The ZY-3 satellite's 2.1-m panchromatic nadir data can be used in generating orthorectified images, fusing images, updating maps and performing interpretation tasks. The panchromatic band pass of ZY-3 ranges from 450 nm to 800 nm; details of the specifications of ZY-3 can be seen in Table 1. The digital number (DN) dynamic range is 10-bit quantization. The swath width of its panchromatic camera is 50 km at nadir. A technical description of the ZY-3 satellite can be found on the official website of SASMAC [1]. There are five levels of ZY-3 nadir (NAD) products provided by the NASG: sensor-corrected (SC) product, geocoded ellipsoid-corrected (GEC) product, enhanced geocoded ellipsoid-corrected (EGEC) product, geocoded terrain-corrected (GTC) product and digital orthorectification map (DOM) product. The SC product is processed with radiometric calibration and sensor calibration; it is the standard original product and has the widest usage among all of the ZY-3 products.

**Table 1.** Specifications of the ZY-3 (Ziyuan-3) satellite. GSD, ground sample distance.

Specification	Value
Orbit height	505.983 km
Orbit inclination	97.421 °
Descending node time	10:30 AM
Revisit cycle	5 days
Mission duration	5 years
Field of regard	Up to 32 ° off-nadir
Swath width	50 km at nadir
Radiometric resolution	10 bit
Panchromatic GSD	2.1 m at nadir
Panchromatic band	450 nm to 800 nm

The image quality of satellite data is of vital importance in various applications, and it can also provide prior knowledge for decision-making and subsequent satellite design [2]. For the panchromatic data of ZY-3, its image quality directly affects the benefit of its usage.

The image quality of the satellite images can be expressed by many technical parameters, such as the GSD, the modulation transfer function (MTF) and the signal-to-noise ratio (SNR) [3–5]. However, these parameters cannot completely describe the quality of the image [6], and they are used primarily

in technical fields and by technical people, such as satellite designers, satellite manufacturers, optical engineers or electric engineers; as a result, the parameters provide quality information and feedback for these technical workers. However, these parameters are unfamiliar and complicated to satellite image product users, which, according to a survey of the U.S. Geological Survey (USGS), may come from different fields, such as agriculture, environmental sciences and management, land use/land cover surveying and even education [7]. In this case, an indicator that can present a comprehensive evaluation of the overall image quality can assist image users to understand the usability of the image according to different tasks [8]. For this reason, image interpretability based on the National Imagery Interpretability Scale (NIIRS) has been proposed as a measure of image quality [9].

Developed by the Imagery Resolution Assessments and Reporting Standards (IRARS) Committee, USA, the NIIRS is the subjective assessment standard of image quality connecting with task requirements [10]. From the perspective of the users, the NIIRS is probably the best measure of assessing the quality of panchromatic images in terms of interpretability [11]. Higher interpretability indicates that more ground objects can be detected or identified, and interpretability in an intuitive and quantitative sense determines which task can be performed with the imagery [12].

Although the NIIRS rating can be estimated by human operators, developers require a tool that will estimate or predict NIIRS performance prior to actually building and testing a new sensor system [9]. The general image-quality equation (GIQE) was developed to provide such predictions. The GIQE was developed under the auspices of the IRARS Committee. The GIQE estimates the NIIRS as a function of a predicted images scale, sharpness, resolution and the signal-to-noise ratio [9,13].

Currently, the research on calculating image interpretability using the GIQE is concentrated on satellite data with very high resolution (VHR), such as QuickBird, IKONOS and Komspat-2 [6,14–16]. Ryan *et al.* [14] discussed the relationship between the NIIRS value and the other parameters of IKONOS imagery. Kim *et al.* [6] predicted the NIIRS value of IKONOS and QuickBird satellite imagery. For QuickBird images, predicted NIIRS values are provided within the metadata. In addition, the IKONOS NIIRS values are also published for the users. The NIIRS provides a reference of the quality and usability of remote sensing images for researchers and users around the world.

In China, only recently have high resolution (HR) satellites been used. The satellite imagery users outside of China are not yet familiar with these HR satellite products and have rarely used them, not to mention the products of the ZY-3 satellite, which was launched in 2012. There have been studies regarding the statistical characteristics, SNR and MTF of ZY-3 data [17,18]; however, no research is available on using GIQE for evaluating the NIIRS of ZY-3 data, which would be useful for users to assess image quality. For this reason, the NIIRS can be used as a common indicator of image quality for users around the world to determine the quality of ZY-3 data.

There is one question remaining in the study of the NIIRS estimation of ZY-3 imagery using the GIQE. According to the study of Leachtenauer *et al.* [12], the revised GIQE is validated over the conditioned range listed in Table 2. The accuracy and validity of the GIQE outside of these bounds is unknown. The range of the GSD is from 0.0762 m (3 inches) to 2.032 m (80 inches); however, the published GSD of the ZY-3 SC NAD data is 2.1 m [1], *i.e.*, slightly higher than 2.032 m. Therefore, whether the interpretability of ZY-3 images can still be estimated with the GIQE should be investigated.

**Table 2.** Range of values of the general image-quality equation (GIQE).

Parameter	Minimum	Maximum
GSD	0.076 m	2.032 m
RER	0.2	1.3
Overshoot	0.9	1.9
Noise Gain	1	19
SNR	2	130

With the NIIRS criteria published by IRARS, the NIIRS can be estimated by human operators. The results of operators on one image are so close, with a standard deviation of 0.3 NIIRS [12], that this observed NIIRS (ONIIRS) value can be considered as the approximate value of the true NIIRS value for ZY-3 image interpretability. Meanwhile, the GIQE can be used to calculate the predicted/estimated NIIRS (PNIIRS) value for the same image. The relationship can be found by a regression analysis using the ONIIRS and the PNIIRS as variables. If there is an acceptable correlation between the ONIIRS and PNIIRS, it would be possible to use this regression expression to estimate the true NIIRS for a ZY-3 image, *i.e.*, the adjusted GIQE can be still used for the estimation of the interpretability of the ZY-3 images, even if its GSD is out of the range designated by IRARS [12].

Hence, as mentioned above, the estimation of the NIIRS for the ZY-3 SC NAD data is necessary and feasible. The purpose of this study is to develop a method to estimate the NIIRS of the ZY-3 images, thereby allowing image interpretability assessment of ZY-3 SC NAD data.

This paper is structured as follows. Section 2 presents the data and method of this study. In Section 3, the result of the ONIIRS and PNIIRS assessments will be presented. Next, a regression function between the ONIIRS and PNIIRS is built with the data of nine images, and subsequently, an adjusted GIQE is proposed based on the regression function, followed by verification of the adjusted GIQE using another four images. Finally, conclusions from this study are drawn in Section 4.

## 2. Data and Method

### 2.1. Data

ZY-3 SC NAD data provided by SASMAC were used in this study for estimating the ONIIRS and the PNIIRS. The SC data is the raw product that has only been radiometrically calibrated and sensor-response calibrated. There were 13 different SC-level NAD data acquired in different areas (Figure 1) or at different times; some of the metadata are listed below in Table 3. These data were acquired by the ZY-3 panchromatic camera from February 2012 to August 2013, with no clouds or a low percentage of clouds, covering the area of Wuhan of Hubei province (HB), Hangzhou of Zhejiang province (ZJ), Beihai and Baise of Guangxi province (GX), Sanya of Hainan province (HN), Baotou of Inner Mongolia province (IM), Wuhu of Anhui province (AH) and Shanghai. These data were received from the station of Miyun (MYN) in Beijing, the station of Sanya (SAY) in Hainan province and the station of Kashgar (KAS) in Xinjiang province, respectively. In addition, QuickBird panchromatic level 2A imagery, which provided the NIIRS value in its metadata, was used in the study in order to recognize the possible bias of human-obtained NIIRS and the true NIIRS.

Figure 1. The distribution of ZY-3 NAD data used in this study.

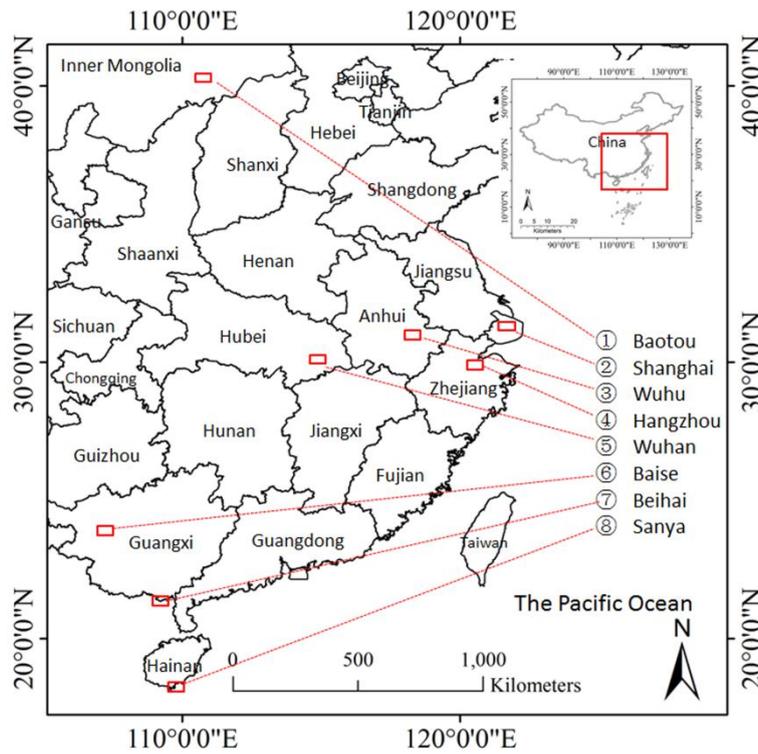


Table 3. Overview of the test datasets. SC, sensor-corrected; MYN, Miyun; SAY, Sanya; KAS, Kashgar.

Location	Scene ID	Date	Track Number	Longitude/Latitude(°)	Sensor	Level	Receiving Station	Cloud Percentage
Wuhu, AH	44506	02/19/2012	624	118.6/31.5	PAN	SC	MYN	5
Wuhan, HB	119315	04/22/2012	1582	114.4/30.7	PAN	SC	MYN	9
Beihai, GX	218553	05/16/2012	1947	109.3/21.5	PAN	SC	SAY	5
Shanghai	164323	05/28/2012	2137	121.4/31.4	PAN	SC	MYN	10
Wuhan, HB	195065	06/20/2012	2479	114.2/30.7	PAN	SC	MYN	10
Beihai, GX	177804	07/09/2012	2775	109.3/21.1	PAN	SC	MYN	3
Baotou, IM	283812	09/15/2012	3809	109.8/40.6	PAN	SC	MYN	7
Sanya, HN	314777	09/21/2012	3902	109.7/18.4	PAN	SC	KAS	38
Baise, GX	324669	09/30/2012	4037	106.7/23.9	PAN	SC	MYN	0
Hangzhou, ZJ	434442	12/30/2012	5413	120.1/30.3	PAN	SC	SAY	3
Wuhan, HB	742360	08/07/2013	8767	113.9/30.7	PAN	SC	KAS	0
Wuhan, HB	748199	08/12/2013	8843	114.4/30.7	PAN	SC	KAS	1
Wuhan, HB	748200	08/12/2013	8843	114.3/30.3	PAN	SC	KAS	3

All of the image data contained different types of areas, such as urban areas, suburban areas and mountainous areas, where abundant ground objects are found in the image, e.g., buildings, fields, water bodies, grass, forest, ground vehicles and some other geographic or artificial features, all of which were related to interpreting features or elements in the National Imagery Interpretability Scale (NIIRS) criteria and were useful in predicting the NIIRS by human operators or using the GIQE.

## 2.2. Method Description

There were several steps for estimating the NIIRS of the ZY-3 SC NAD data. First, the ONIIRS is evaluated by human operators. Second, the PNIIRS is calculated through the required parameters using the GIQE. Third, a regression analysis is performed for attaining the expression of the ONIIRS and PNIIRS. Then, we validate the regression expression using correlation coefficients to determine whether this regression expression could be used for estimating the NIIRS of ZY-3 SC NAD data.

### 2.2.1. Observed NIIRS Assessment

In the experiment, every individual image of these 13 ZY-3 NAD images and one QuickBird image from different areas or different time periods was divided into 4 sub-images. Six participants who were experienced in image processing and analysis interpreted the sub-images according to the Additional NIIRS Criteria [9], in which a number of additional criteria (agricultural, cultural and natural) appeared [19]. In the experiment, the sub-images of ZY-3 and QuickBird were mixed together for a blind test. After the participants interpreted the image by checking the corresponding or similar feature described in the Additional NIIRS Criteria; they provided a decimal score for every sub-image for higher precision, and the final NIIRS level for one image was determined by taking an average of the ONIIRS values estimated for the 4 sub-images from the 6 participants. To achieve an objective result, the test was performed with a sufficient amount of time, with the participants taking 4 working days to complete this interpretation task.

To recognize the possible bias, from the result of QuickBird image, we calculate the error between the ONIIRS and the NIIRS provided in the metadata of the QB image. For the purpose of decreasing the subjective error, the average error is considered as the bias between the ONIIRS and the true NIIRS in our experiment. With the help of this bias, we can revise the ONIIRS for ZY-3 imagery, through which the subsequent analysis is carried out. The revised ONIIRS was regarded as the true NIIRS value for the image in this study.

### 2.2.2. NIIRS Estimation Using the GIQE

We used the GIQE to predict the NIIRS level of the image as the PNIIRS. The early published GIQE 3.0 version used GSD, RER and SNR to calculate the NIIRS for an optical system. Subsequently, validation and modification was performed on the GIQE 3.0 version to obtain the GIQE version 4.0. GIQE version 4.0 is described by Equation (1):

$$NIIRS = 10.251 - a \times \lg GSD + b \times \lg RER - 0.656 \times H - 0.334 \times \left( \frac{G}{SNR} \right) \quad (1)$$

In the above equation, *GSD* is the ground sample distance, in units of inches; *RER* denotes relative edge response; *H* is the overshoot; *G* is the noise gain calculated from an optional sharpening kernel; and *a* and *b* are parameters. If *RER* is greater than 0.9, then *a* is equal to 3.32 and *b* is equal to 1.559; otherwise, *a* equals 3.16 and *b* equals 2.817. This method makes it possible to predict the NIIRS values based on the imaging parameters [12].

The GSD was calculated as the geometric mean of the ground sampled distance based on a projection of the pixel pitch distance to the ground [12,20]. The GSD was computed in inches in both

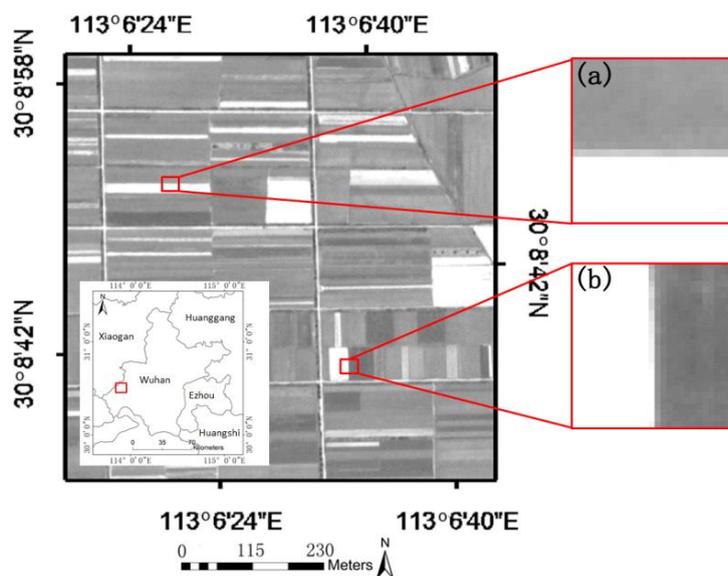
the X and Y dimensions (the along-scan and cross-scan directions, respectively), and we calculated their geometric mean. The GSD is given by Equation (2).

$$GSD = \frac{\left(\frac{PP}{FL}\right) \times SL}{\cos(LA)} \tag{2}$$

In Equation (2), *PP* is the pixel pitch, *FL* is the focal length, *SL* is the slant range and *LA* is the look angle; the details can found be in [12].

The RER of the image can be measured by analyzing the slopes of the edge profiles within the image, and this value represents the MTF characteristics of the image. To calculate the RER of the image, we first selected the edge manually, where the intensities of the DN value of both sides of the edge were changing rapidly. Figure 2 shows the example of the edge determined manually for edge response (ER) generation; Figure 2a is an edge from the horizontal direction, and Figure 1b presents an edge in the vertical direction. Next, the values of the pixels along the perpendicular lines are interpolated with cubic splines, the edge profile of the row or column perpendicular to the edge provided was calculated. This procedure was repeated for each perpendicular line. Finally, all of the estimated edge profiles were used to obtain one averaged function that could represent the empirical edge spread function (ESF) (Figure 3a) [12]. From this average ESF of the edge, the differential of the normalized ER in +0.5 and -0.5 was calculated in the vertical and horizontal direction, respectively, as the ER<sub>x</sub> and ER<sub>y</sub>. The RER is calculated as a geometric mean of ER<sub>x</sub> and ER<sub>y</sub> [21].

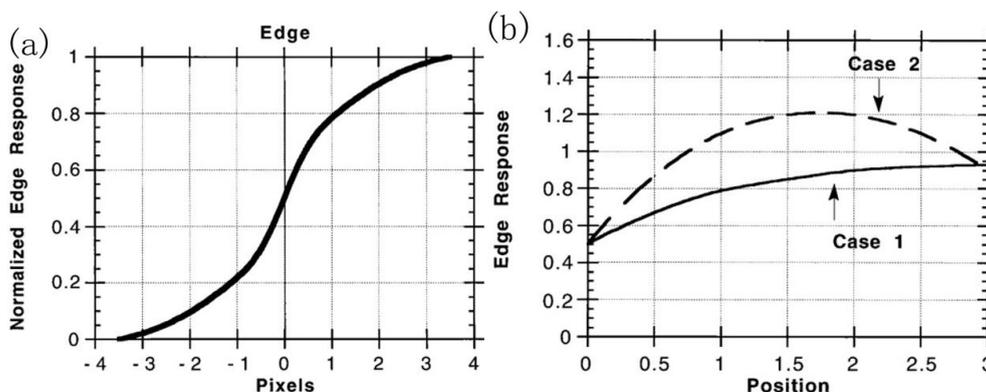
**Figure 2.** Two samples in the image of the agriculture area of Wuhan, HB, for edge extraction. The two edges are the common borders between different fields. (a) Edge extracted for edge response (ER)<sub>x</sub>. (b) Edge extracted for ER<sub>y</sub>.



The GSD and RER are the dominant terms in the equation of GIQE, while the overshoot term, *H*, the *G* term and the SNR have a much smaller impact [12]. The overshoot-height term, *H*, models the RER overshoot that is due to the modular transfer function compensation (MTFC). The overshoot height is measured over the range of 1.0 to 3.0 pixels from the edge in 0.25-pixel increments. If the

edge is monotonically increasing, it is defined as the value at 1.25 pixels from the edge. Otherwise, it is the maximum value in the pixel range of +1 to +3, as shown in Figure 3b [12]. As for the GSD and the RER, H is calculated on both the X and Y directions.

**Figure 3.** (a) The measurement of RER. (b) The measurement of H. The edge in Case 1 is monotonically increasing, the H is defined as the value at 1.25 pixels from the edge. Case 2 is not monotonically increasing, the H is the maximum value in the pixel range of +1 to +3 [12].



G is the noise gain resulting from the edge sharpening and is the geometric mean of the coefficients. Because the ZY-3 SC NAD data are not processed by MTFc, the value of G is 1. The SNR has the least effect on the NIIRS value (GSD 72%, RER 20% and SNR less than 1%) [22]; in this case, although there are methods of analyzing the SNR from the image, we use the laboratory method for its simplicity and convenience. In this method, the ratio of the mean and the standard deviation was calculated from a homogeneous area in the image, and this ratio was used as the approximate value of the SNR [23]. The calculation of the SNR is given by Equations (3)–(5):

$$\overline{DN} = \frac{1}{N} \sum_{i=1}^N DN_i \tag{3}$$

$$\sigma = \sqrt{\frac{1}{(N-1)} \sum_{i=1}^N (DN_i - \overline{DN})^2} \tag{4}$$

$$SNR = \frac{\overline{DN}}{\sigma} \tag{5}$$

$\overline{DN}$  is the mean value of the DN value in the homogeneous area,  $\sigma$  is the standard deviation of the DN value of this area, and SNR is estimated as the ratio of  $\overline{DN}$  and  $\sigma$ .

Using the values of GSD, RER, H, G and SNR, the PNIIRS can be estimated with GIQE, and its result could be used in the experiment to analyze the relationship between the PNIIRS and the ONIIRS, with the latter value considered as the true value of the NIIRS of the image.

### 2.2.3. Linear Regression Analysis

Linear regression analysis was further used to investigate the relationships between ONIIRS and PNIIRS values. The PNIIRS values were calculated using the GIQE from eight of the study images

and then were related to the ONIIRS values using a linear regression analysis. In the regression analysis, the Shapiro-Wilk test was used to examine the normal distribution of the ONIIRS and the PNIIRS, which were the two variables in the analysis. R and R-squared were used to investigate the correlation between the ONIIRS and PNIIRS. The F-test was used to estimate the significance of the relationship between the ONIIRS and PNIIRS. The analysis involved determining whether the relationship between the ONIIRS and PNIIRS was significant at the 0.05 level. We then validated the PNIIRS and ONIIRS of another five images, examining their NIIRS predicted from the adjusted GIQE with the ONIIRS. If a high correlation and similarity still existed in the result, the adjusted GIQE equation was further proven to be applicable for the estimation of the NIIRS of ZY-3 SC NAD data.

All of the statistical analyses were performed using Microsoft® Excel® 2010 and IBM® SPSS® Statistics 20. A significance level of 0.05 was established in the analysis. If the result fulfills the requirements, then we can use regression expression as the prediction expression for estimating the NIIRS of the ZY-3 SC NAD data.

### 3. Results and Discussion

#### 3.1. Results of Observed NIIRS Assessment

Subjectivity is the main source of error in image interpretation by human operators, because different individuals may differ in their interpretation of an area of the image, such as different familiarity and experience for the scenes or geographical features. In addition, fatigue may affect the accuracy of the interpretation. In this case, the result that is represented by the NIIRS value may be different. To avoid subjective error, we provided a fixed and stable working environment for the participants. For example, the brightness of the environment was stable, and we set up the same contrast and brightness of the display used by the participants and maintained them for the entire experiment period. There are 56 sub-images in total for the operators to interpret during the experiment; to avoid fatiguing the operators in this exhausting work and to ensure the accuracy of the interpreted result, a period of four working days was provided to the operators as a sufficient amount of time for completing the assessment.

The ONIIRS assessment results are presented in Table 4. In this article, the NIIRS values are set to two decimal places, and the other values without specific states are set to three decimal places to enable higher accuracy and better comparison. As shown in Table 4, the mean value of the ONIIRS was 2.94. Except for the image of Hangzhou, which had a 0.325 standard deviation of the ONIIRS value, all other images had a standard deviation lower than 0.3, and the mean value of their standard deviation was 0.255, which represents an acceptable accuracy for the assessment. The reason why the image in Hangzhou had such a high standard deviation is that the acquisition date of the image was 30 December 2012 (winter), when it had just snowed, so some of the ground areas were covered by snow; in addition, although the cloud percentage of the image was only 3%, the clouds were all aggregated on the left bottom of the image. As a result, the difference of the ONIIRS value between the sub-images of the left bottom and other sub-images of the Hangzhou image is significant, which results in a higher standard deviation.

**Table 4.** Results of the observed NIIRS (ONIIRS).

Date	Location	Scene ID	ONIIRS (SD)	Revised ONIIRS
02/19/2012	Wuhu, AH	44506	2.98 (0.284)	2.81
04/22/2012	Wuhan, HB	119315	3.09 (0.237)	2.92
05/16/2012	Beihai, GX	177804	2.98 (0.267)	2.81
05/28/2012	Shanghai	164323	2.93 (0.245)	2.76
06/20/2012	Wuhan, HB	195065	2.98 (0.285)	2.81
07/09/2012	Beihai, GX	218553	2.95 (0.257)	2.78
09/15/2012	Baotou, IM	283812	2.81 (0.292)	2.64
09/21/2012	Sanya, HN	314777	2.84 (0.272)	2.67
09/30/2012	Baise, GX	324669	2.92 (0.174)	2.75
12/30/2012	Hangzhou, ZJ	434442	2.81 (0.325)	2.64
08/07/2013	Wuhan, HB	742360	2.78 (0.215)	2.61
08/12/2013	Wuhan, HB	748199	3.10 (0.247)	2.93
08/12/2013	Wuhan, HB	748200	3.08 (0.212)	2.91
	Mean		2.94 (0.255)	2.77
	SD		0.111	0.109

We compared the ONIIRS result of the QuickBird image (4.567 NIIRS on average, for estimating the bias; we set this value three decimal places) with the NIIRS provided in its metadata (4.400 NIIRS), and the bias of ONIIRS and the true NIIRS was attained as 0.167. Therefore, the ONIIRS of the ZY-3 NAD data was revised by the bias, and the final results of the revised ONIIRS are listed in Table 4. The mean of the revised ONIIRS was 2.77, a bit lower than the original ONIIRS value of 2.94. In the subsequent study, we used the revised ONIIRS as the final ONIIRS for analysis.

### 3.2. Results of NIIRS Predicted Using the GIQE

In the experiment, we calculated the GSD, RER, H, G and SNR for every image, with the results listed in Table 5. The GSD in the X dimension had an average value of 2.083, while the average value of the GSD in the Y dimension was 2.089. It was found that the GSD in the Y dimension (GSD<sub>y</sub>) was 0.3% greater than that in the X dimension (GSD<sub>x</sub>). The geometric mean of GSD<sub>x</sub> and GSD<sub>y</sub> was 2.086 on average. The GSD had a maximum value of 2.062 and a minimum value of 2.122. The results of GSD are shown in Figure 4a. With a standard deviation of 0.013, this GSD value is acceptable. Compared with the GSD required by the original design objective of the ZY-3 satellite (2.1 m), the GSD of the image data in practice (2.086 m) is better than expected.

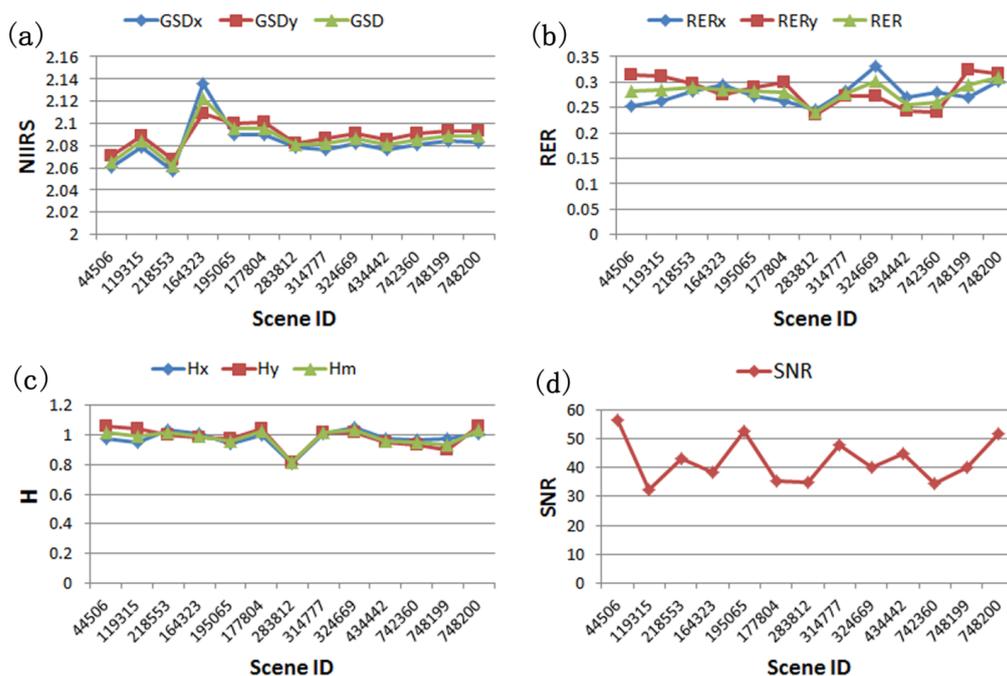
The results of the calculation of the RER indicate that (Figure 4b) the average value of the ER<sub>x</sub> and the ER<sub>y</sub> are 0.277 and 0.284, respectively, with the latter value being 2.52% greater than the former value. The higher ER<sub>y</sub> value means better sharpness of the edge in the along-scan direction. The average value of the RER is 0.280, with a standard deviation of 0.019. Therefore, according to GIQE 4.0, the RER value is less than 0.9, so *a* is set as 3.16 and *b* is set as 2.817 (Equation (1)). In the experiment, we found that the edge used to calculate the ER for the ESF can be found from the triangular roof of the building, a high reflective building and its shadow, the border of a road or runway, the common border of the two different fields (Figure 2) and the edge between the port and water surface.

**Table 5.** Results of the predicted NIIRS (PNIIRS) assessment.

Date	Location	Scene ID	GSD (m)	RER	H	SNR *	PNIIRS
02/19/2012	Wuhu, AH	44506	2.066	0.282	1.018	56	1.99
04/22/2012	Wuhan, HB	119315	2.084	0.286	0.997	32	2.01
05/16/2012	Beihai, GX	177804	2.062	0.289	1.015	43	2.02
05/28/2012	Shanghai	164323	2.122	0.285	0.997	38	1.98
06/20/2012	Wuhan, HB	195065	2.095	0.281	0.959	52	2.01
07/09/2012	Beihai, GX	218553	2.096	0.281	1.024	35	1.96
09/15/2012	Baotou, IM	283812	2.080	0.240	0.814	35	1.92
09/21/2012	Sanya, HN	314777	2.081	0.278	1.014	48	1.97
09/30/2012	Baise, GX	324669	2.086	0.301	1.037	40	2.04
12/30/2012	Hangzhou, ZJ	434442	2.081	0.256	0.962	45	1.90
08/07/2013	Wuhan, HB	742360	2.086	0.260	0.951	35	1.92
08/12/2013	Wuhan, HB	748199	2.088	0.295	0.938	40	2.08
08/12/2013	Wuhan, HB	748200	2.088	0.309	1.036	52	2.08
	Min		2.062	0.240	0.814	32	1.90
	Max		2.122	0.309	1.037	56	2.08
	Mean		2.086	0.280	0.982	42	1.99
	SD		0.015	0.019	0.060	7.698	0.058

\* The value of SNR is in the form of integer.

**Figure 4.** Results of the GIQE parameters. (a) The calculation results of GSD. (b) The chart of the results of RER. (c) The values of H. (d) The results of SNR.



From the ESF calculated for the ER, we estimated the H value of the image according to the monotonicity of the ESF in the range of +1 to +3 pixels. In the experiment, we found that for the ZY-3 image, the curves of the ESF in both the X dimension and the Y dimension were not monotonically increasing, so the maximum value in the range of +1 to +3 was considered as the Hx or Hy value. H is

also the geometric mean of  $H_x$  and  $H_y$ . The values of  $H_x$  and  $H_y$  in the images were all very close, as shown in Figure 4c. The mean of the  $H$  calculated from the images is 0.982.

Because the ZY-3 SC NAD data are processed without MTFC, the value of  $G$  is one. Using the laboratory method to estimate the SNR, we obtained a value of approximately 42 on average. This result is only a probable estimation, because different choices of the homogeneous areas in the same image may result in some different value, and the calculation result of the laboratory method itself has been shown to be slightly greater than the true value [23]. However, the SNR only had a slight effect on the NIIRS (less than 1%), so we used this method for the study. The result is shown in Figure 4d.

With the aid of these parameters, the PNIIRS was calculated for these images; the results are listed in Table 5. A mean value of 1.99 for the PNIIRS was obtained, with a standard deviation of 0.058. The lowest PNIIRS is 1.90 in the Hangzhou area; in contrast, the highest PNIIRS is 2.08 in the middle area of Wuhan. We can compare these two images. The GSD of the image in the middle of Wuhan (Scene ID 748199) has lower resolution (higher GSD), which is 2.088 m in GSD and is 0.3% higher than that of the image of Hangzhou (2.081 m). However, the former image has the RER with a value of 0.295; conversely the RER value of the image of Hangzhou is just 0.256, which is 15% lower. As a result, the image of Wuhan has a 9% higher PNIIRS than that of Hangzhou. Because GSD and RER take up more than 90% of the effect of NIIRS calculation, the values of  $H$ , SNR and  $G$  are not discussed in this comparison. Therefore, we can find in the comparison that higher GSD cannot guarantee a higher NIIRS of the image, because the sharpness of the edge is also important for image interpretability.

### 3.3. Linear Regression Analysis

After we estimated the ONIIRS and PNIIRS of the images, we found that there is a significant difference between the PNIIRS and ONIIRS. The value of the PNIIRS calculated using the GIQE is obviously lower than that of the ONIIRS, as found in Table 6. The reason for this difference may be that the GSD of the ZY-3 data exceeds the range of the GSD in the GIQE study.

**Table 6.** Comparison between the ONIIRS and PNIIRS.

Date	Location	Scene ID	PNIIRS	ONIIRS
02/19/2012	Wuhu, AH	44506	1.99	2.81
04/22/2012	Wuhan, HB	119315	2.01	2.92
05/16/2012	Beihai, GX	177804	2.02	2.81
05/28/2012	Shanghai	164323	1.98	2.76
06/20/2012	Wuhan, HB	195065	2.01	2.81
07/09/2012	Beihai, GX	218553	1.96	2.78
09/15/2012	Baotou, IM	283812	1.92	2.64
09/21/2012	Sanya, HN	314777	1.97	2.67

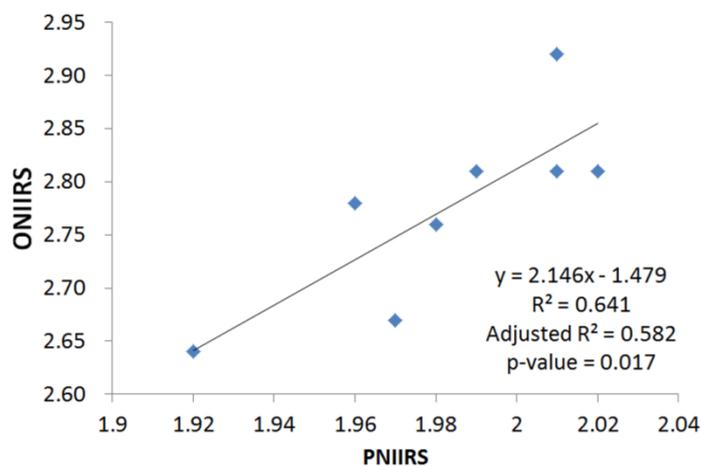
A linear regression analysis was used to examine the relationship between the ONIIRS and PNIIRS. The results of the ONIIRS and PNIIRS in the Shapiro–Wilk test are 0.489 and 0.536, respectively, which are all greater than 0.05, so a normal distribution can be assumed in both of these two variables.

A regression analysis indicates that the value of the ONIIRS and the value of the PNIIRS can effectively predict one another, as shown in Figure 5. The regression expression is given by:

$$y = 2.146 \times x - 1.479 \tag{6}$$

In the expression,  $y$  refers to the ONIIRS (which is considered as the true value of the NIIRS), and  $x$  refers to the PNIIRS. We discovered that the relationship has a high correlation coefficient, with a value of R-squared of 0.641, *i.e.*, much greater than 0.5. This value of R-squared means that 64.1% of the variation of ONIIRS values can be reflected by the PNIIRS. The adjusted R-squared is 0.582, and we found a significant correlation between these two variables, as the  $p$ -value is 0.017, which is less than the significance level of 0.05.

**Figure 5.** Linear regression analysis. The expression is  $y = 2.146 \times x - 1.479$ , and a high correlation can be found in the R-squared value of 0.641; the adjusted R-squared is 0.582, and the  $p$ -value is 0.017.



The linear regression analysis results indicate that a highly significant relationship exists between the ONIIRS and the PNIIRS. In addition, with the help of the regression expression, in spite of the fact that the GSD is outside the proper range, the GIQE can still be adjusted to estimate the true NIIRS value of the ZY-3 data using Equation (7):

$$NIIRS_{ZY3} = 2.146 \times NIIRS - 1.479 \tag{7}$$

$NIIRS_{ZY3}$  is the adjusted NIIRS value that is suitable for ZY-3 SC NAD data, and the  $NIIRS$  is calculated using GIQE 4.0.

The validation of the equation was performed using five other images, with the results presented in Table 7; the values of the ONIIRS and  $NIIRS_{ZY3}$  calculated using the Equation (7) are very close: the mean of the error is 0.054 with a standard deviation of 0.069, which represents the very high similarity of the results.

The values of the ONIIRS, PNIIRS and  $NIIRS_{ZY3}$  of all thirteen ZY-3 SC-level NAD data are shown in Table 8 and Figure 6. From the results of the estimation, we find that the  $NIIRS_{ZY3}$  values calculated using the adjusted GIQE of the study images are nearly equal and are related with the corresponding values of the ONIIRS. The average value of  $NIIRS_{ZY3}$  is 2.79, which, to some extent, is nearly the same

as that of the ONIIRS (2.77). The standard deviation of NIIRS<sub>ZY3</sub> is 0.12, *i.e.*, lower than the value of 0.3 in the GIQE. As a result, the approximate NIIRS of the ZY-3 SC NAD data is 2.79.

**Table 7.** Result of the validation.

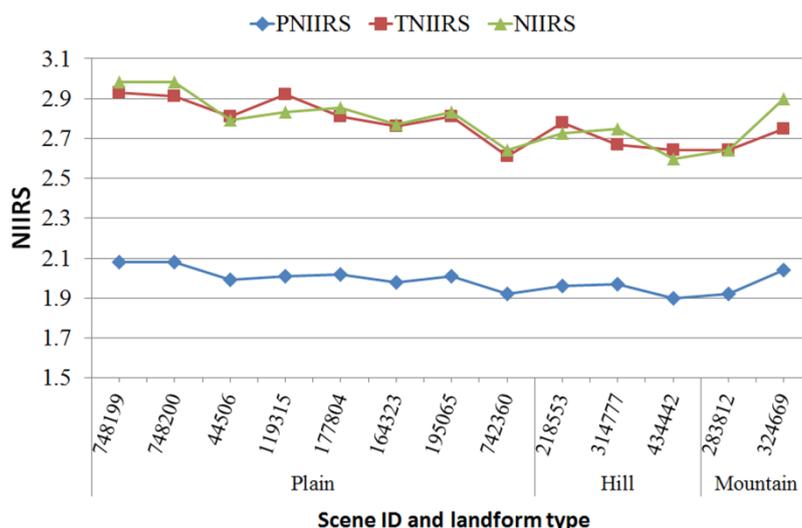
Date	Location	Scene ID	ONIIRS	NIIRSZY3	ERROR
2012/9/30	Baise, GX	324669	2.75	2.90	0.149
2012/12/30	Hangzhou, ZJ	434442	2.64	2.60	−0.042
2013/8/7	Wuhan, HB	742360	2.61	2.64	0.031
2013/8/12	Wuhan, HB	748199	2.93	2.98	0.055
2013/8/12	Wuhan, HB	748200	2.91	2.98	0.075
	Mean		2.82	2.77	0.054
	SD		0.188	0.148	0.069

**Table 8.** Results of the PNIIRS, ONIIRS and NIIRS<sub>ZY3</sub>.

Date	Location	Scene ID	PNIIRS	ONIIRS	NIIRSZY3
2012/2/19	Wuhu, AH	44506	1.99	2.81	2.79
2012/4/22	Wuhan, HB	119315	2.01	2.92	2.83
2012/5/16	Beihai, GX	177804	2.02	2.81	2.86
2012/5/28	Shanghai	164323	1.98	2.76	2.77
2012/6/20	Wuhan, HB	195065	2.01	2.81	2.83
2012/7/9	Beihai, GX	218553	1.96	2.78	2.73
2012/9/15	Baotou, IM	283812	1.92	2.64	2.64
2012/9/21	Sanya, HN	314777	1.97	2.67	2.75
2012/9/30	Baise, GX	324669	2.04	2.75	2.90
2012/12/30	Hangzhou, ZJ	434442	1.90	2.64	2.60
2013/8/7	Wuhan, HB	742360	1.92	2.61	2.64
2013/8/12	Wuhan, HB	748199	2.08	2.93	2.98
2013/8/12	Wuhan, HB	748200	2.08	2.91	2.98
	Mean		1.99	2.77	2.79
	SD		0.06	0.11	0.12

Considering that the landform type may affect the interpretability of the image, we grouped these 13 images into three types: the plain area, the hill area and the mountain area. Eight images are from the plain area; three images are from the hill area, and only two images are from the mountain area. The average NIIRS of ZY-3 SC NAD data in the plain area is 2.84, which is the highest in all landform types. The mountain area has an average NIIRS of 2.77, and that of the hill area is 2.69, which is the lowest. The reason for this result may be that the image of the plain area has the least distortion and the most abundant ground objects with obvious spatial structures or characteristics (*i.e.*, buildings, roads and fields), which help the image to attain the best interpretability. Compared with the hill area, the greater portion of the vegetation in the mountain area is like a background for other objects, which increases the sharpness of the edge of the object. Therefore, though there is comparatively more distortion in the image of the mountain area, it still has a higher NIIRS than the hill area. The results are in Figure 6.

**Figure 6.** Results of the ONIIRS, PNIIRS and NIIRS<sub>ZY3</sub>.



### 3.4. NIIRS Level of the ZY-3 Data

The NIIRS is a 10-level scale defined by interpretation tasks or criteria [9]. Once the NIIRS of the image is known, the available interpretation tasks or related applications can be estimated according to the NIIRS Criteria. Conversely, the tasks or applications are assigned by determining which type of image can fulfill the requirements of the users. In addition to the original NIIRS, which is mainly focused on military targets, IRARS has developed and published several other scales and products of potential interest to the civil remote-sensing community [9,24,25]. In this study, we used the adjusted GIQE (based on GIQE 4.0 [26]) to estimate the NIIRS of ZY-3 SC NAD data, with a result of 2.79. According to the Additional NIIRS Criteria in the Civil NIIRS Reference Guide [9], some of the tasks for which ZY-3 SC NAD data can be applied are listed below (Table 9). The images in Figure 7 show some other criteria that can be applied to ZY-3 SC NAD data. These tasks mainly are concentrated in NIIRS Level 1 and Level 2, such as “Distinguish between runways and taxiways at a large airfield (NIIRS 1)” and “Detect large buildings (e.g., station, hospitals, factories) (NIIRS 2)” (Figure 7a) in the Civil NIIRS Criteria; “Identify long-lot land ownership patterns along major waterways (NIIRS 1.7)” and “Detect windbreaks (i.e., rows of trees) between fields (NIIRS 2.6)” (Figure 7c) in the Agricultural Criteria; “Detect multi-lane divided highways (e.g., greater than four lanes (NIIRS 1.9)” and “Identify large buildings as multi-wing (NIIRS 2.7)” in the Cultural Criteria; “Detect large coastal sand beaches (NIIRS 1.8)” and “Detect utility towers in forested regions on the basis of a sequence of circular clearings (NIIRS 2.7)” (Figure 7b) in the Natural Criteria. Briefly, ZY-3 SC NAD data are useful for most of the common interpretation tasks for civil, agricultural and natural applications, among others.

According to the result, we compare the NIIRS of ZY-3 SC NAD data with some other satellite data. QuickBird and IKONOS are famous sub-meter VHR satellites in the world, with a GSD of 0.61 m and 0.82 m, respectively [27,28]. SPOT-5 is also a well-known satellite worldwide, with its 2.5-m GSD [29], and it has been considered as the representative of HR satellites for quite a long time. From the official product guide of QuickBird, we can find that its potential NIIRS is 4.9 [27]. The NIIRS of IKONOS is estimated to be 4.5, which can be found in the literature of Ryan [14]. SPOT-5

has the same problem as ZY-3 in calculating NIIRS using GIQE, in that their GSD (2.5 m) exceed the scope of that in GIQE 4.0 (2.03 m). Therefore, we use the adjusted GIQE in the study for achieving a rough estimation of NIIRS of SPOT-5, and the result is 2.43. As is stated in the manuscript, the NIIRS of ZY-3 is 2.79.

**Table 9.** Additional NIIRS Criteria.

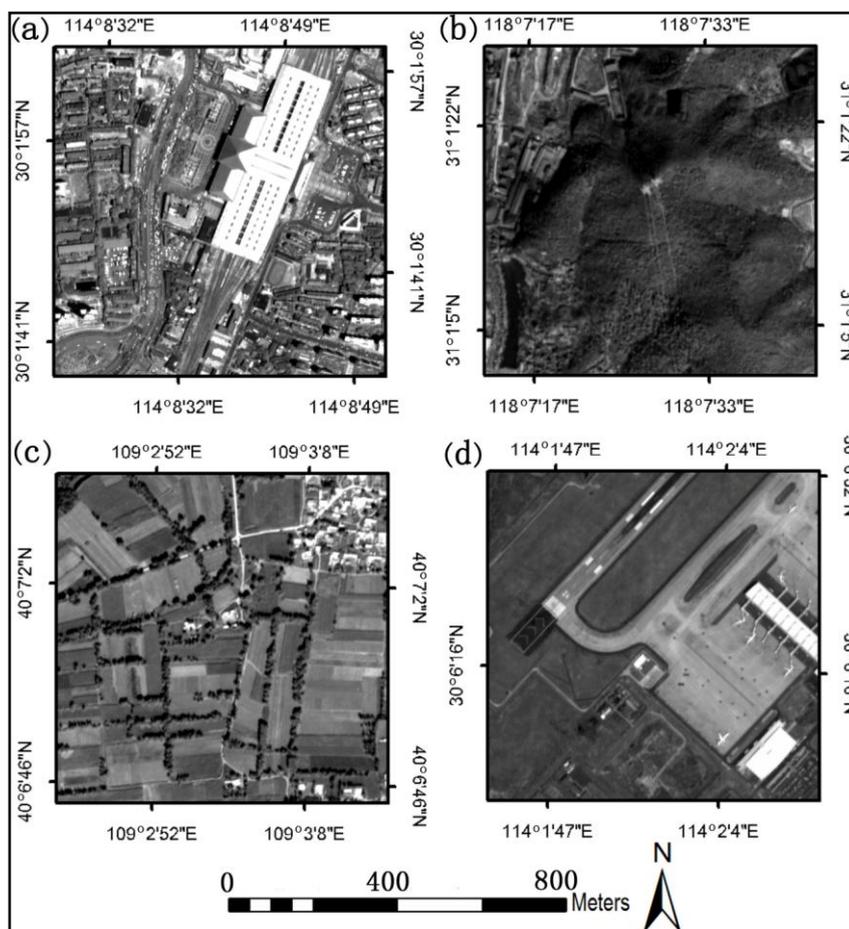
NIIRS	Civil NIIRS Criteria	Agricultural Criteria	Cultural Criteria	Natural Criteria
1	Distinguish between runways and taxiways at a large airfield.	Detect large ( <i>i.e.</i> , greater than 100 acres) slash and burn clearings in jungle areas. (NIIRS 1.4)	Detect major highway and rail bridges over water (e.g., Golden Gate, Chesapeake Bay). (NIIRS 1.7)	Detect large areas (e.g., greater than 100 acres) of timber clear cutting. (NIIRS 1.7)
	Identify large area drainage patterns by type (e.g., dendritic, trellis, radial).	Identify long-lot land ownership patterns along major waterways. (NIIRS 1.7)	Detect multi-lane divided highways (e.g., greater than four lanes). (NIIRS 1.9)	Detect large coastal sand beaches. (NIIRS 1.8)
	...	...	...	...
2	Identify large ( <i>i.e.</i> , greater than 160 acres) center-pivot irrigated fields during the growing season.	Detect forest clearings in suspected coca growing areas. (NIIRS 2.4)	Identify large buildings as multi-wing. (NIIRS 2.7)	Distinguish between islands and nests of moored ships. (NIIRS 2.2)
	Detect large buildings (e.g., hospitals, factories).	Detect windbreaks ( <i>i.e.</i> , rows of trees) between fields. (NIIRS 2.6)	Detect a dual lane/track bridge over water. (NIIRS 2.7)	Detect utility towers in forested regions on the basis of a sequence of circular clearings. (NIIRS 2.7)
	...	...	...	...

A comparison can be seen in Figure 8. The VHR satellite not only has higher resolution in its imagery, but also a higher NIIRS than the HR satellite. However, considering the wider swath width and lower cost, HR satellites, like ZY-3 and SPOT-5, are still a better choice for some certain tasks or applications.

**3.5. Uncertainties, Errors and Accuracies**

In the study, we had thirteen ZY-3 SC NAD data available for the estimation of its NIIRS, so errors may exist in the regression analysis. The accuracy of the equation of adjusted GIQE should be further discussed when there are more data. Additionally, in the experiment of ONIIRS, strictly speaking, six participants are not enough for achieving a very high accuracy, because subjective error may occur among the six participants during the experiment. We used the laboratory method in calculating the SNR, because the SNR has the least effect on predicted NIIRS, and this method is comparatively simple and convenient. The SNR derived from this method is proven to be a little greater than the true value [23]. Therefore, the result is only a rough approximation for the SNR.

**Figure 7.** ZY-3 SC nadir (NAD) images in different areas. (a) A railway station of Wuhan, HB. In the image, the large buildings of the railway station and other houses can be found, and eight tracks can be detected in the rail yard. (b) The mountainous area of Wuhu, AH. Some forest clearings can be detected in the image, and two utility towers in the forest are obvious and easily found. (c) A field in Baotou, IM, where we can find windbreaks between fields. (d) The airport of Wuhan, HB. The taxiway and runway is clearly identifiable, and some medium-sized airplanes can be detected.



**Figure 8.** Comparison of the GSD and NIIRS of different satellite imagery.

#### 4. Conclusions

This study estimated the NIIRS of ZY-3 SC NAD data using an adjusted GIQE equation derived from the analysis of the ONIIRS and the PNIIRS. In this study, we not only presented the results of the ONIIRS, but also described the characteristics of the parameters of the ZY-3 data used in the GIQE (GSD, RER, H, G and SNR) and discussed the results of the values of the PNIIRS. A linear regression analysis of the ONIIRS and PNIIRS revealed a significant relationship between these scales. As a result, with the help of the regression expression, the NIIRS of ZY-3 NAD data can be estimated using the adjusted GIQE, in spite of its exceeding the range of the GSD.

The result of the study demonstrates that, as the first civil high resolution satellite for use in China, ZY-3 has a moderate GSD of approximately 2.086 m, *i.e.*, slightly better than its design objective of 2.1 m. Although the ZY-3 SC NAD data are the original product, these data still have good quality regarding sharpness, and the NIIRS value of 2.79 for its interpretability makes it able to completely fulfill the requirements of the tasks of NIIRS 2.7 and lower levels.

The main contributions of this study are as follows:

(1) The study is the first to use the GIQE for assessing the interpretability of the image quality of ZY-3 data. We proposed the method of estimating the adjusted NIIRS of ZY-3 SC NAD data. The study provides worldwide users with the results of the interpretability of Chinese ZY-3 SC NAD data based on the NIIRS, which is a common indicator for assessing the image quality of remote sensing data. The method and results are also helpful for expanding the method of assessing the image quality of Chinese remote sensing satellite data.

(2) We discussed the availability of using the GIQE to estimate the NIIRS of satellite image data whose GSD is greater than 2.03 m and explored the use of an adjusted GIQE in estimating the NIIRS of such data. The method and the results of this study can be used as a reference for studying the interpretability of other similar satellite remote sensing data (with a GSD greater than 2.03 m) using the GIQE.

In future work, we will perform more experiments to study the relationship between the GIQE and the NIIRS for ZY-3 SC NAD data. Additionally, the parameters of the ZY-3 SC NAD data used in the GIQE will be further examined. For example, other methods will be used to improve the estimation result of the SNR. More ZY-3 SC NAD data will be used to improve the accuracy of analyzing the relationship between the ONIIRS and the PNIIRS using a regression analysis. Then, we can examine the expression used to predict the NIIRS of ZY-3 SC NAD data using the GIQE and appropriately modify it. Moreover, if possible, in addition to ZY-3 data, the data of other satellites with a GSD greater than 2.03 m (80 inches) can be used for studying the uncertainty of the GIQE, such as SPOT-5 (2.5 m), ALOS (2.5 m) and ZY-1-02C (2.36 m in the HR camera and 5 m in the panchromatic camera). Finally, we can achieve a common estimation of the NIIRS using the GIQE for most of the VHR and HR optical satellite data.

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### Author Contributions

This research was mainly performed and prepared by Lin Li and Heng Luo. Lin Li supervised the study and his comments were considered throughout the paper. Heng Luo contributed with ideas, conceived and designed the study, and wrote the paper. Haihong Zhu reviewed and edited the manuscript.

### Conflicts of Interest

The authors declare no conflict of interest.

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