

FOVEON VS BAYER: COMPARISON OF 3D RECONSTRUCTION PERFORMANCES

M. Vlachos¹, D. Skarlatos^{1*}, P. Bodin²

¹ Cyprus University of Technology, Dep. Of Civil Engineering and Geomatics, P.O. Box 50329, Limassol 3603, Cyprus -
(dimitrios.skarlatos, marinos.vlachos)@cut.ac.cy

² National School of Geographic Sciences (ENSG) - pierre.bodin@ensg.eu

Commission II

KEY WORDS: Photogrammetry, Bayer pattern, accuracy, de-mosaicking, point cloud, structure light scanner

ABSTRACT:

The main idea of this particular study was to validate if the new FOVEON technology implemented by sigma cameras can provide better overall results and outperform the traditional Bayer pattern sensor cameras regarding the radiometric information that records as well as the photogrammetric point cloud quality that can provide. Based on that, the scope of this paper is separated into two evaluations. First task is to evaluate the quality of information reconstructed during de-mosaicking step for Bayer pattern cameras by detecting potential additional colour distortion added during the de-mosaicking step, and second task is the geometric comparisons of point clouds generated by the photos by Bayer and FOVEON sensors against a reference point cloud. The first phase of the study is done using various de-mosaicking algorithms to process various artificial Bayern pattern images and then compare them with reference FOVEON images. The second phase of the study is carried on by reconstructing 3D point clouds of the same objects captured by a Bayer and a FOVEON sensor respectively and then comparing the various point clouds with a reference one, generated by a structured light hand-held scanner. The comparison is separated into two parts, where initially we evaluate five separate point clouds (RGB, Gray, Red, Green, Blue) for each camera sensor per site and then a second comparison is evaluated on colour classified RGB point cloud segments.

1. INTRODUCTION

All modern digital cameras capture radiometric information using a colour filter array named Bayer pattern. This way, each individual pixel of the sensor records only one colour channel. For a complete RGB image, the missing colour values are interpolated to neighbouring pixels through a processing step called de-mosaicking. The only exception to this trend are digital cameras using the FOVEON sensor, exclusively used by Sigma cameras. This sensor uses the physical properties of silicone to catch all colour information in each pixel of the sensor, hence there is no need for post processing to reconstruct full coloured images. The most commonly used cameras in photogrammetry are single sensor cameras using Bayer technology. They are more affordable than 3-CCD (Wootton, 2005) cameras and offer good results in terms of accuracy and precision for 3D modelling techniques. However, new Sigma's camera manages to capture as much information in Red, Green and Blue than 3-CCD cameras for a way much more affordable price.

When it comes to geosciences, quality of the final product directly relies on raw data quality, which is guaranteed by the use of high quality and well calibrated equipment. In photogrammetry that means good lighting conditions, proper setup of camera settings and raw image recording to ensure maximum detail capturing (Stamatopoulos et al., 2012).

Therefore, as FOVEON technology inherently records more information than Bayer in blue, green and red ("Foveon - Direct Image Sensors," n.d.), it is legitimate to investigate if, whether or not, this kind of technology can enhance geometric quality of final photogrammetric products and 3D point clouds in particular. In fact, given that the Bayern pattern consist of 50% green sensitive pixels, 25% red sensitive pixels and 25% blue

sensitive pixels, it is expected that the performance of the FOVEON should be considerably better on red and blue channels and marginally better in green.

This study aims to compare the capability of both technologies to build accurate and precise 3D models, against reference 3D point clouds acquired by hand held structure light scanner (Faro Freestyle 3D X). This comparison is based in a number of colourful graffiti following a colour classification and comparison of SfM & MVS point clouds generated by two different cameras against the structure light scanner. The devices used to lead this comparison are the Nikon D90 APS-C sensor with Bayer technology and the Sigma Sd1 Merrill APS-C FOVEON sensor.

The study is carried out in two phases. First one aims to evaluate the quality of information reconstructed during de-mosaicking step by detecting potential additional colour distortion added during the de-mosaicking step. The de-mosaicking phase is done using Rawpy ("RawPy" 2014), an open source python wrapper for libraw ("LibRaw" 2018) library. It offers the possibility to process some de-mosaicking algorithms like Adaptive Homogeneity-Directed method (AHD) (Hirakawa and Parks, 2005), Discrete Hartley Transform method (DHT) (N. Bracewell, 1983), Bilinear algorithm (DCB), Various number of gradient method (VNG) (Chang et al., 1999), Pixel Grouping method (PPG) (Lin, n.d.), AAHD algorithm. The original FOVEON photos are used as reference (Figure 1) for this test. Artificial Bayer pattern images are created and the Rawpy is used to de-mosaick the photos, which are then compared against the original FOVEON photos. The differences are quantified and represent a measure of missing information.

* dimitrios.skarlatos@cut.ac.cy; phone +357 25002360; fax +357 25002806; www.photogrammetric-vision.weebly.com

The second phase consists in point cloud geometric comparisons against a reference point cloud. First a comparison of five point clouds per sensor, per site is done. There, the RGB, Gray, Red, Green and Blue point clouds derived from each sensor's imagery are compared with a structured light handheld scanner generated 3D point cloud. As a secondary comparison, segments from each sites RGB point clouds are extracted for both sensors and then are compared with the reference hand held scanner point cloud. For each test site only one segment is exported per sensor which refers to each site's dominant colour i.e. for one site the blue points are extracted for the second the green points and for the third the red points are extracted.

2. PREVIOUS WORK

Perko et al., (2002) presented various alternatives for the creation of digital colour images from an aerial or space sensor. They also explained the issue of "pan-sharpened" images using higher resolution panchromatic images and lower resolution colour images as they addressed the different coloration schemes combining high resolution black and white pixels with the lower resolution colour pixels and observed the differences between the colour images obtained from various approaches. Additionally, an experiment in their work is described to classify the errors committed in de-mosaicking Bayer pattern colour images.

Perko et al., (2005) gave an evaluation concept to assess the geometrical accuracy of resulting colour images. In that particular concept, only if no geometrical distortions were created during the de-mosaicking process, it is allowed to use the images for 3D reconstruction or arbitrary metrical photogrammetric processing.

Li et al., (2008) made a comparative study evaluating 11 selective inter-channel de-mosaicking algorithms on two different benchmark data sets. The two benchmark data sets that were used are Kodak PhotoCD images and high-quality IMAX images. During their study the authors stated that in order to reduce the risk of mismatches and thus the presence of artefacts in the images, a fuse of de-mosaicking results by different algorithms is recommended.

Stamatopoulos et al., (2012) took advantage of the camera's higher dynamic range without applying any pre-processing steps and applied various de-mosaicking algorithms to the RAW images of a Bayer sensor. Following that, the authors examined the extent of variation in the mean positional standard error of object target point coordinates. The authors concluded that all tested de-mosaicking algorithms performed better than the standard in-camera JPEG image formation.

Riutort-Mayol et al., (2012) introduced a practical, comprehensive and flexible laboratory procedure to analyse the radiometric values and the uncertainty effects due to FOVEON sensor system. Their procedure was performed on the grey level output signal using image raw units and was entirely based on statistical and experimental techniques.

Journes, (2014) proposed to study the impact of the different components of the acquisition chain on the image quality. The author also studies some of the state-of-the-art algorithms and the different quality metrics for image quality assessment. Author states that the choice of an appropriate algorithm is dependent on the frequency content which evolves all over the acquisition chain.

Fent and Meldrum, (2016) Showcased a method using the FOVEON sensor in full spectrum mode in combination with green-pass filters to reproduce the red-magenta hues of healthy vegetation in aerial images without resorting to multi-exposures or channel swaps. During their study the authors developed a colour model for a FOVEON/green-pass filter using quantum efficiencies to explain the colour effects observed.

3. DE-MOSAICKING PROTOCOL

The de-mosaicking testing procedure, wishes to quantify how much information is omitted in Bayern pattern colour images and how effective the tested algorithms are. The main concept is that the images taken from the FOVEON camera can be used as a reference, which is then converted in an artificial Bayer pattern file and then de-mosaicked with several de-mosaicking algorithms. The recreated image from the de-mosaicking process, can be compared against the FOVEON original image, to evaluate the loss of information. In technical terms, Rawpy Python module, was used for creating the artificial Bayer pattern from original FOVEON images and recreating the images using several de-mosaicking algorithms. The conversion of the FOVEON reference jpeg image I(ref) is done by replacing every pixel value of the Bayer sensor raw image with the radiometric values of FOVEON and respecting a GRBG Bayer pattern and then, de-mosaicking step is followed using the aforementioned 6 algorithms for every I(Bayer) that was derived. The de-mosaicked resulting images, I(demo), are then compared to reference image to see how different from I(ref), I(demo) is.

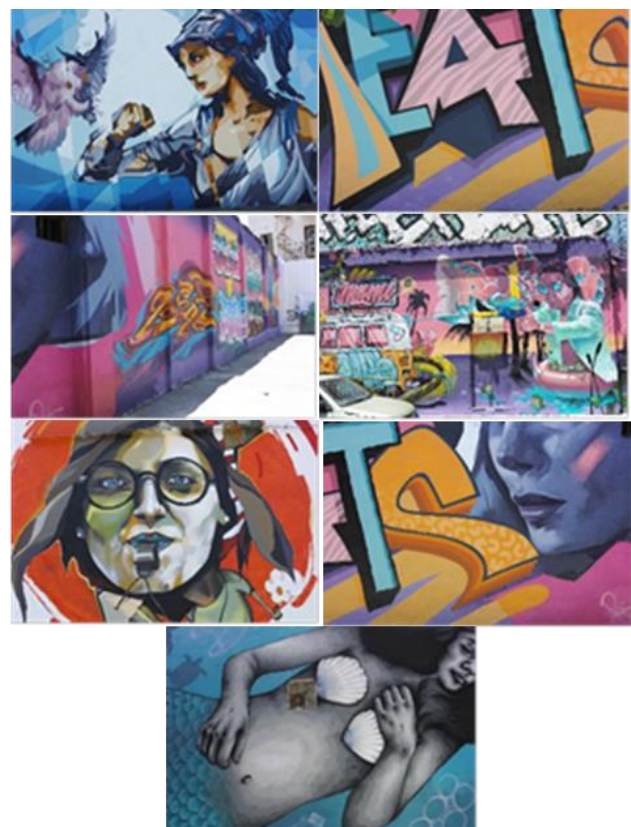


Figure 1: Array of images containing I(ref)s used for the testing (From left to right, top to bottom): Athena, aZoomed, pinkGraf, Miami, redWhistle, sLips, Sirena.

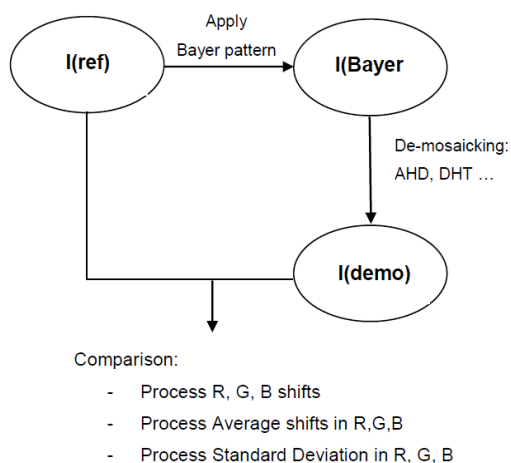


Figure 2: De-mosaicking test pipeline

To evaluate how false, reconstructed colour is, for each channel, a comparison of the radiometric values of Red, Green and Blue channels in I(ref) and I(demo) is needed. Each channel in I(demo) is subtracted from the I(ref) and produces a Red, a Green and a Blue map of differences in float values. These maps are exported as float files and are analysed to extract statistics (Table 2&3, Figures 8&9) which will quantify false colours.



Figure 3: From Top to bottom: Athena, Red, Green and Blue difference maps for AHD algorithm

Figure 3 shows Red, Green and Blue map of differences for Athena I(ref). To illustrate, Green map is obtained by subtracting Green value from de-mosaicked Athena to Green value from reference Athena. This computation is done for Red, Green and Blue channel and for the 7 different references (Table2, Figure 8). In total 21 difference maps for every de-mosaicking algorithm are produced.

4. POINT CLOUDS COMPARISON PROTOCOL

The second phase of our study was to evaluate which camera delivers the best datasets to produce accurate 3D point clouds, in correlation with each colour channel. Three different testing sites were chosen for the testing each one giving a large emphasis to each colour channel. As reference, point clouds captured by a structured light hand-held scanner sensor, which provides an accuracy better of 1mm at 1m object-scanner distance were used. The capture protocol was designed so that to have a digital image GSD of approx. 2.25 mm and a corresponding accuracy on the camera lens axis of approx. 4.5 mm. To achieve that, distance from the objects should have been 11m for the FOVEON and 10m for the Bayer as their physical pixel sizes differ slightly (Table 1), and it was necessary to have exactly similar GSD, to be able to compare their geometric accuracies. With the structure light scanner accuracy being 4 times better that the photogrammetric one, the structured light point served as reference. Both the hand-held scanner and photogrammetric point clouds were georeferenced using local Ground Control Point networks in each site separately.

	SIGMA SD1	NIKON D90
Focal lens	24mm	24mm
Effective pixels	15M 4700x3200	12.2M 4288x2848
Pixel pitch	5.0µm	5.5µm
Effective area	23.5mm x 15.7mm	23.60mm x 15.80mm

Table 1: Camera Specifications

The base between captures has been computed for an overlap of 80% and resulting datasets are composed of three photos where the working section is totally visible (Figure 4). Nine Ground control points have been measured with total station to build a common local coordinate system in every test site separately. The photogrammetric pipeline was executed using Agisoft Photoscan. From RGB Bayer and FOVEON datasets red, green and blue radiometric values were extracted to build a Red, a Green and a Blue dataset for each camera. Grey-scale datasets were also computed for each camera, which resulted in 10 blocks in total (RGB, R, G, B and grey-scale for each camera) for each site. According to software's reports, the effective acquisition distance is 10.80m for FOVEON camera and 10m for Bayer Camera, which resulted to a GSD of 2.24mm for FOVEON camera and 2.26mm for Bayer camera, as obtained from bundle adjustment results.

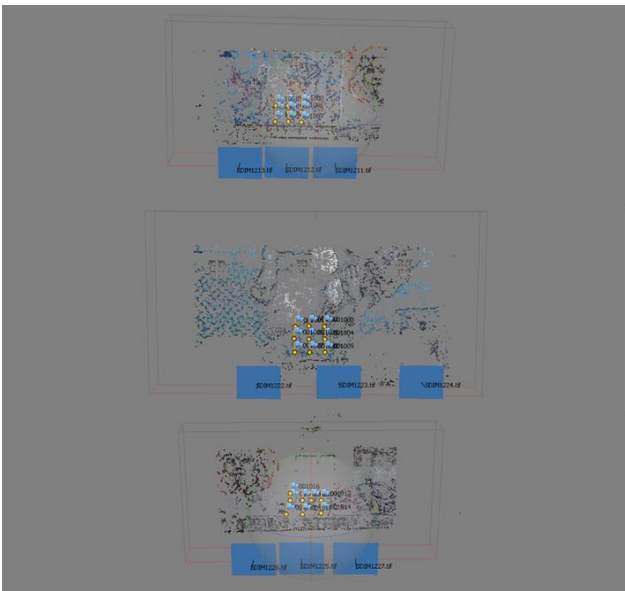


Figure 4: Capture plan of images as is shown from the UI of Agisoft Photoscan



Figure 5: Bayer & FOVEON RGB Point Clouds

After the point clouds were generated, they were compared against the reference laser scanner point cloud in CloudCompare. Cloud to cloud distances in the axis parallel to the camera lens axis (Y axis) are computed from the reference and then from each comparison the computed mean distance, standard deviation and RMS error were evaluated.

5. RESULTS

5.1 De-Mosaicking Results

The final results of the de-mosaicking algorithms are demonstrated below in Figure 6. For the three sites all algorithms produce I(demo) images with a slight red tone across the images. To eliminate that and obtain I(demo) images with the same colours as the I(ref), a white balance and a brightness adjustment

was applied to try reproducing the colour from the reference image. Unfortunately, after that step the I(demo) images appeared still to have a red tone as it can be seen below in Figure 6. An assumption is that FOVEON channel brightness is affected by the silicon layers. Hence the Blue is more bright and red is less bright, as an average. During the post-processing within the FOVEON camera, this is balanced somehow. Unfortunately, in the six de-mosaicking algorithms that were used, this phenomenon is not taken into account, and that's the reason we are ending up with de-mosaicked images with red tone. Assuming the above to correct that phenomenon during the de-mosaicking, the parameters for white balance and brightness must be adjusted accordingly in order to reproduce an I(demo) image visually identical to the I(ref).



Figure 6: Athena, redWhistle and mermaid graffiti. From top to bottom: The final I(ref), I(demo)AAHD, I(demo)AHD, I(demo)PPG, I(demo)DCB, I(demo)DHT and I(demo)VNG.

It is obvious that there is no visual changes or differences with the naked eye between the I(demo) images produced from the 6 different de-mosaicking algorithms. Additionally, this is validated below with the evaluation of the statistical analysis for every algorithm as the average differences between I(ref) and I(demo) are almost identical between the 6 algorithms (Figure 8).

Table 2 shows the Mean of the average differences in radiometric values for Red, Green and Blue channels between I(ref) and I(demo) for seven different images which have been de-mosaicked with the 6 different algorithms. The results below represent the average from all 6 algorithms per image. The resulting differences to the reference highly depend on the de-

mosaicked images. This is linked to the way white balance and brightness adjustment are processed. White balance and brightness don't represent the capture conditions in which I(refs) have been shot. The results shown in Table 2 & 3 should be interpreted carefully as in the derived images' white balance and brightness correction parameters may carry heterogeneous errors from one image to another. The analytic results for every algorithm's performance are shown in Figures 8 & 9 in the Appendix section.

Mean of Average Differences	MEAN OF RED	MEAN OF GREEN	MEAN OF BLUE
AZoomed	26.432	14.934	14.169
Athena	1.964	14.646	34.358
sLips	30.208	4.498	-0.002
redWhistle	27.926	-4.153	-10.407
miami	28.574	21.002	26.959
Sirena	-4.025	4.782	15.434
pinkGraf	20.209	-2.985	21.034
MEAN	18.756	7.532	14.506

Table 2: Mean of Average Differences in radiometric values for Red, Green, Blue of the 6 De-mosaicking algorithms for every dataset.

Mean StDev	MEAN OF RED	MEAN OF GREEN	MEAN OF BLUE
AZoomed	26.080	22.024	24.287
Athena	21.858	25.319	32.589
sLips	23.183	17.863	18.989
redWhistle	27.897	25.554	24.332
miami	31.691	29.087	31.506
Sirena	20.336	18.364	21.581
pinkGraf	27.818	21.803	29.234
MEAN	25.552	22.859	26.074

Table 3: Mean of Standard Deviations in Red, Green, Blue of the 6 De-mosaicking algorithms for every dataset.

Taking into consideration the "MEAN" row in Tables 2 and 3, it can be noticed that de-mosaicking results are coherent with the Bayer pattern theory. As a Bayer sensor gathers twice more green information than blue or red information, the de-mosaicking interpolation step should reconstruct green missing values better than red and blue ones. As is shown in Figure 3 where the blue difference map is brighter and red is less bright on average, the assumption that FOVEON channel brightness is affected by the silicon layers is verified by the "MEAN" row in Table 2.

The standard deviation calculation statistically eliminates, previously mentioned, white balance and brightness heterogeneous effects. The standard deviation computed varies among the images and the values for different algorithms are comparable between images. Additionally, based on Figure 9 of the Appendix section, values are really close between them and it cannot be decided whether an algorithm is better than another with the applied methodology. Even from the Standard deviation section, Green channel seems to be better reconstructed than Blue and Red, which is coherent with Bayer filter theory.

5.2 Point Cloud Comparisons Results

Because of the red tone that the de-mosaicked images appeared to have, for the 3D photogrammetric point cloud generation and comparison, the raw camera files converted into 16-bit tiff files,

were used in point cloud generation, instead of the de-mosaicking algorithms data sets.

For the point cloud comparisons two methodologies were followed. The first methodology was the comparison of five different point clouds in their entirety (RGB, Grayscale, Red, Green and Blue) per camera system with the reference. The second comparison was made on the RGB point clouds by classification of points based on the radiometric attributes of specific segments and then comparing those segments with the reference point cloud following a similar procedure and analysis.

5.2.1 Cloud to Cloud Comparisons

Athena (Blue dominant)	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON RGB	-0.323	1.989	2.015
Bayer RGB	-0.649	3.714	3.770
FOVEON Gray	-0.287	2.215	2.234
Bayer Gray	-1.059	5.065	5.175
FOVEON B	-0.624	2.313	2.396
Bayer B	1.167	5.810	5.926
FOVEON G	-0.129	2.195	2.199
Bayer G	-0.363	5.453	5.465
FOVEON R	-0.510	2.438	2.491
Bayer R	3.936	6.567	7.656

Table 4: Mean Distance, Standard Deviation and RMS error of every point cloud from the reference point cloud for Athena graffiti.

Mermaid (Green Dominant)	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON RGB	-0.317	2.900	2.917
Bayer RGB	0.376	3.102	3.125
FOVEON Gray	-0.189	3.033	3.039
Bayer Gray	-0.219	4.514	4.519
FOVEON B	-0.511	3.251	3.291
Bayer B	1.489	4.629	4.863
FOVEON G	-0.183	3.130	3.135
Bayer G	0.179	4.238	4.242
FOVEON R	-0.339	3.450	3.467
Bayer R	-2.555	10.225	10.539

Table 5: Mean Distance, Standard Deviation and RMS error of every point cloud from the reference point cloud for Mermaid graffiti.

Random Red (Red Dominant)	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON RGB	-0.287	2.840	2.855
Bayer RGB	-0.753	2.841	2.939
FOVEON Gray	-0.115	3.264	3.266
Bayer Gray	-1.167	3.944	4.113
FOVEON B	-0.550	3.361	3.406
Bayer B	-0.096	3.200	3.201
FOVEON G	-0.073	3.509	3.510
Bayer G	-0.867	4.158	4.247
FOVEON R	-0.260	3.580	3.589
Bayer R	-3.579	8.317	9.054

Table 6: Mean Distance, Standard Deviation and RMS error of every point cloud from the reference point cloud for Random Red graffiti.

As it is shown in Tables 4, 5 and 6, for both sensors and the C2C comparisons, the mean distances from the reference for the RGB and grayscale point clouds are as expected, i.e FOVEON has the smaller mean distance from the reference in every single comparison. In the individual colour comparison, the mean distance between the FOVEON PCs and the reference ones are -0.624mm (Athena Blue) to -0.073mm (Random Red Green) which is much smaller (on absolute values) than the GSD of the acquired images. Regarding the Standard deviation and the RMSE of FOVEON the values are varying from 2.4mm to 3.6mm. As for the Bayer sensors the standard deviation and RMS error are overall worse than FOVEON varying from 3.2mm to 10.2mm and 3.2mm to 10.5mm respectively. Regarding the Bayer sensor the mean distances, StDev and RMS error from the reference PC are not similar to all channels. The Green PC has the smallest values regarding the individual colour comparisons in the Athena and Mermaid Graffiti where the blue PC has the smallest values and thus the better quality in the case of the Random Red Graffiti for both sensors.

Additionally, as we see from the comparisons, the Bayer Red point clouds for all the graffiti were by far the most erroneous. That could be explained by the fact that while blue and green channels have specific wavelengths, that is not the case for the red. In some sensors, information about the Red Edge channel might be recorded within the red channel, where in other sensors only the strict red band width is recorded. This difference might affect the point cloud extraction procedure. The bandwidth that both sensors record the radiometric information in the Red channel is unknown to this study thus we can only make a hypothesis for the above statements. Further tests could prove the aforementioned but that is not the focus of this study.

Based on the above comparison, FOVEON appears to be a superior sensor than Bayer regarding the photogrammetric point cloud generation with less noise as the standard deviation and RMS error values show in most of the comparisons. That is what was initially expected, since FOVEON records in all pixels original radiometric information of Red, Green and Blue, instead of interpolating two channels in every pixel.

5.2.2 Cloud Segmentation and classification

In this phase, a classification of the RGB point clouds based on specific point selection was made. More specifically, a manual selection was made on specific regions to extract only points with radiometric attributes similar to the dominant band of each site. After the point classification a comparison was carried on against the reference point cloud provided by the hand-held laser scanner.

Athena (Blue)	# of Points	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON	15992	-0.047	1.062	1.063
Bayer	39646	-0.373	2.751	2.776
Mermaid (Green)	# of Points	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON	83298	-0.225	4.073	4.079
Bayer	155086	-0.038	3.909	3.909
Random Red (Red)	# of Points	C2C Mean Dist. (mm)	StDev (mm)	RMS (C2C)
FOVEON	135754	0.335	2.150	2.176
Bayer	90748	-0.512	2.505	2.557

Table 7: Mean Distance, Standard Deviation and RMS error of Red, Green and Blue classified points for every respective site.

In the case of Athena and Random Red graffiti it is clear that Blue and Red classified points of both sensors have the better statistics where the Green classified points in mermaid are worse. In the case of mermaid graffiti, the mean distance, standard deviation and RMS error for Green classified points are worse because of the lack of distinct green points. Additionally, Bayer sensor selected points have better statistics than FOVEON's in that particular scenario.

Even though in this comparison FOVEON outperforms Bayer for Red and Blue, we cannot consider this comparison as most reliable, since the two sensors record different radiometric information for the same objects. Because of that, a proper classification based on the radiometry of the reconstructed objects, cannot be made as different areas and points with different radiometric properties will be selected respectively. This is shown in the case of the blue classified points of the Athena graffiti where many of the points selected by the classifier were closer to red as well as many points in the Random Red graffiti are classified as red but in reality, are white (Figure 7). For the case of Mermaid, the number of green classified points for FOVEON is quite large in comparison with Athena potentially due to the lack of distinct green points as blue and green are almost equally dominant.

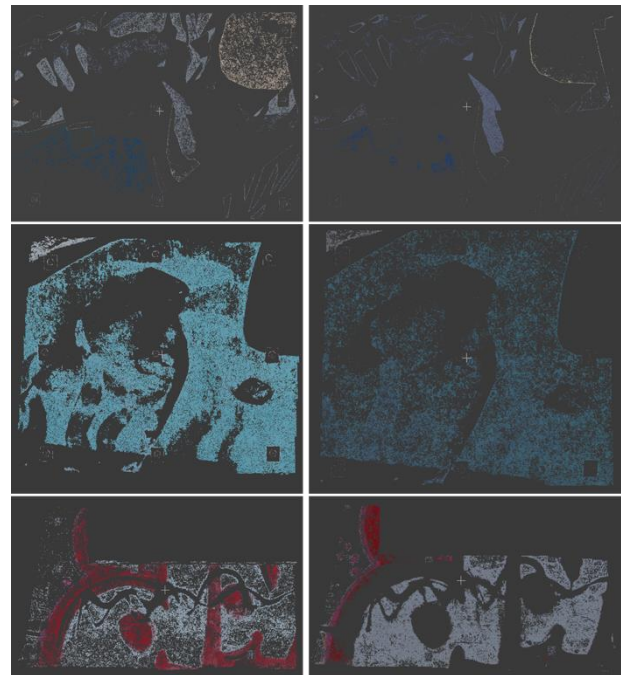


Figure 7: "Blue", "Green" and "Red" Classified Points from the Athena, Mermaid and Random Red graffiti point clouds of Bayer (Left) and FOVEON (Right) sensors.

6. CONCLUSIONS AND FUTURE WORK

This paper describes the difficulty to reconstruct missing colour values of a Bayer pattern image and tries to evaluate and compare overall quality of 3D point clouds obtained with datasets resulting from different sensor colour capturing technologies. Additionally, our test using 6 selected de-mosaicking algorithms showcased the difficulties of true colour reconstruction from a Bayer pattern raw file.

FOVEON technology is simpler in terms of post-process. It captures more information and less processing is needed for the final image, but more time is needed for recording data from

sensor to card. One of the main limitations is its use to low light conditions as high ISO shots produce a lot of noise. It still needs white balance and brightness adjustment to deliver sharp and properly coloured images. On the other hand, Bayer technology needs additional process steps to reconstruct images like the demosaicking step plus a white balance and brightness adjustment are necessary.

In a terrestrial photogrammetric context, both technologies do well. FOVEON technology's performance regarding the accuracy and noise is superior based on the results of this study. By evaluating the single channel point clouds, the difference in the radiometric information between the two technologies become more noticeable as Bayer's resulting geometric quality is on average inferior to FOVEON's.

As it is easier, and less challenging to test and compare these technologies in a terrestrial photogrammetric context, a future study can be made to compare the performance of Bayer and FOVEON technologies in aerial or underwater imaging. In order to perform the aforementioned tests, the heavy cameras, need to be fitted on an octacopter, which is feasible. For the underwater environment though, although Nikon D90 has several commercial underwater housings, Sigma lacks such accessory.

REFERENCES

Chang, E., Cheung, S., Pan, D.Y., 1999. Color filter array recovery using a threshold-based variable number of gradients, in: Sampat, N., Yeh, T. (Eds.), . International Society for Optics and Photonics, pp. 36–43. <https://doi.org/10.1117/12.342861>

Fent, L., Meldrum, A., 2016. A Foveon Sensor/Green-Pass Filter Technique for Direct Exposure of Traditional False Color Images. *J. Imaging* 2, 14. <https://doi.org/10.3390/jimaging2020014>

Foveon - Direct Image Sensors [WWW Document], n.d. URL <http://www.foveon.com/article.php?a=67> (accessed 11.23.18).

Hirakawa, K., Parks, T.W., 2005. Adaptive Homogeneity-Directed Demosaicing Algorithm. *IEEE Trans. Image Process.* 14, 360–369.

Journes, F., 2014. A study of image quality assesment and color image reconstruction for mono-sensor camera. <https://doi.org/10.1109/ICASSP.2002.1004620>

Li, X., Gunturk, B., Zhang, L., 2008. Image demosaicing: a systematic survey 68221J. <https://doi.org/10.1117/12.766768>

LibRaw | raw image decoder [WWW Document], 2018. URL <https://www.libraw.org/> (accessed 11.23.18).

Lin, C.-K., n.d. Demosaic - Chuan-kai Lin [WWW Document]. URL <https://sites.google.com/site/chklin/demosaic> (accessed 11.23.18).

N. Bracewell, R., 1983. Discrete Hartley Transform. *JOSA* 73, 1832–1835. <https://doi.org/10.1364/JOSA.73.001832>

Perko, R., Leberl, F., Gruber, M., 2002. Color in Photogrammetric Remote Sensing. *Int. Arch. Photogramm. Remote Sens. Spat. Inf. Sci.* 34, 59–64.

Perko, R., Philipp, F., Bauer, J., Klaus, A., 2005. Geometrical Accuracy of Bayer Pattern Images. Evaluation. RawPy [WWW Document], 2014. URL <https://letmaik.github.io/rawpy/api/rawpy.RawPy.html> (accessed 11.23.18).

Riutort-Mayol, G., Marqués-Mateu, Á., Seguí, A.E., Lerma, J.L., 2012. Grey level and noise evaluation of a foveon X3 image sensor: A statistical and experimental approach. *Sensors (Switzerland)* 12, 10339–10368. <https://doi.org/10.3390/s120810339>

Stamatopoulos, C., Fraser, C.S., Cronk, S., 2012. Accuracy Aspects of Utilizing Raw Imagery in Photogrammetric Measurement. *ISPRS - Int. Arch. Photogramm. Remote Sens. Spat. Inf. Sci.* XXXIX-B5, 387–392. <https://doi.org/10.5194/isprsarchives-XXXIX-B5-387-2012>

Wootton, C., 2005. A practical Guide to Video and Audio Compression, Audio. <https://doi.org/10.1029/2004GL022120>

APPENDIX

	AZoome	Athena	sLips	redWhis	miami	Sirena	pinkGr	Mean
AHD	26.432	1.959	30.211	27.934	28.574	-4.041	20.212	18.755
VNG	26.402	1.919	30.181	27.878	28.525	-4.021	20.185	18.724
DHT	26.453	1.993	30.221	27.931	28.533	-3.997	20.213	18.764
DCB	26.427	1.974	30.209	27.939	28.649	-4.048	20.208	18.766
PPG	26.440	1.975	30.222	27.939	28.602	-4.025	20.225	18.768
AAHD	26.440	1.963	30.203	27.936	28.562	-4.016	20.208	18.757
AHD	14.934	14.644	4.496	-4.156	20.995	4.779	-2.988	7.529
VNG	14.924	14.643	4.493	-4.162	21.016	4.779	-2.990	7.529
DHT	14.943	14.657	4.506	-4.142	21.038	4.791	-2.976	7.546
DCB	14.941	14.644	4.504	-4.151	20.959	4.785	-2.980	7.529
PPG	14.932	14.642	4.495	-4.154	21.001	4.776	-2.990	7.529
AAHD	14.930	14.648	4.495	-4.153	21.005	4.781	-2.985	7.532
AHD	14.168	34.360	-0.002	-10.404	26.981	15.440	21.035	14.511
VNG	14.175	34.363	-0.002	-10.401	26.902	15.424	21.037	14.500
DHT	14.164	34.348	-0.003	-10.415	26.925	15.427	21.022	14.495
DCB	14.173	34.368	-0.002	-10.394	27.009	15.445	21.043	14.520
PPG	14.185	34.374	0.000	-10.390	26.985	15.454	21.050	14.522
AAHD	14.151	34.339	-0.005	-10.438	26.951	15.412	21.017	14.490

Figure 8: Average Differences in radiometric values for Red, Green and Blue of the 6 De-mosaicking algorithms for every dataset.

	AZoome	Athena	sLips	redWhis	miami	Sirena	pinkGr	Mean
AHD	26.092	21.905	23.171	27.881	31.589	20.369	27.818	25.546
VNG	26.120	21.930	23.192	27.927	31.965	20.398	27.888	25.631
DHT	26.050	21.804	23.174	27.899	31.786	20.303	27.797	25.545
DCB	26.060	21.835	23.157	27.860	31.480	20.331	27.786	25.501
PPG	26.071	21.856	23.174	27.899	31.642	20.345	27.807	25.542
AAHD	26.088	21.818	23.227	27.916	31.684	20.271	27.811	25.545
AHD	22.025	25.319	17.863	25.554	29.087	18.366	21.807	22.860
VNG	22.021	25.312	17.856	25.547	29.009	18.355	21.797	22.842
DHT	22.014	25.304	17.856	25.544	28.974	18.352	21.786	22.833
DCB	22.043	25.354	17.884	25.586	29.368	18.390	21.832	22.922
PPG	22.029	25.322	17.870	25.558	29.101	18.369	21.811	22.866
AAHD	22.011	25.304	17.846	25.533	28.985	18.352	21.786	22.831
AHD	24.282	32.584	18.985	24.320	31.421	21.568	29.217	26.054
VNG	24.285	32.595	18.992	24.334	31.733	21.607	29.255	26.114
DHT	24.275	32.598	18.960	24.310	31.561	21.572	29.245	26.074
DCB	24.285	32.566	18.995	24.327	31.329	21.575	29.213	26.041
PPG	24.289	32.591	18.999	24.341	31.456	21.586	29.227	26.070
AAHD	24.303	32.598	19.002	24.359	31.533	21.579	29.248	26.089

Figure 9: Standard Deviations of the Differences in radiometric values for Red, Green and Blue of the 6 De-mosaicking algorithms for every dataset.