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Structural similarity analysis of midfacial fractures – a feasibility study

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Abstract: The structural similarity index metric is used to measure the similarity between two images. The aim here was to study the feasibility of this metric to measure the structural similarity and fracture characteristics of midfacial fractures in computed tomography (CT) datasets following radiation dose reduction, iterative reconstruction (IR) and deep learning reconstruction. Zygomaticomaxillary fractures were inflicted on four human cadaver specimen and scanned with standard and low dose CT protocols. Datasets were reconstructed using varying strengths of IR and the subsequently applying the PixelShine™ deep learning algorithm as post processing. Individual small and non-dislocated fractures were selected for the data analysis. After attenuating the osseous anatomy of interest, registration was performed to superimpose the datasets and subsequently to measure by structural image quality. Changes to the fracture characteristics were measured by comparing each fracture to the mirrored contralateral anatomy. Twelve fracture locations were included in the data analysis. The most structural image quality changes occurred with radiation dose reduction (0.980036 ± 0.011904), whilst the effects of IR strength (0.995399 ± 0.001059) and the deep learning algorithm (0.999996 ± 0.000002) were small. Radiation dose reduction and IR strength tended to affect the fracture characteristics. Both the structural image quality and fracture characteristics were not affected by the use of the deep learning algorithm. In conclusion, evidence is provided for the feasibility of using the structural similarity index metric for the analysis of structural image quality and fracture characteristics.

Keywords: Maxillofacial trauma; structural similarity index; fracture visualization; computed tomography (CT); low dose; advanced model based iterative reconstruction; deep learning

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Introduction

Computed tomography (CT) is the routine imaging method of choice for the diagnosis of midfacial fractures in emergency department patients (1). Unfortunately, obtaining a good quality image involves a substantial radiation dose (1,2) However, iterative reconstruction (IR) and deep learning algorithms have been proposed to maintain image quality on reducing the radiation dose (1,3).

The IR algorithm was developed to improve the image quality of reconstructed datasets by producing image data that accurately corresponds to the measured projection data (4). The most recent full model-based IR algorithm involves both backward and forward projection. This complex algorithm uses the difference between an estimation of the raw data and the real measured data from the imaging system to reduce the image noise in successive iterations (5).

Recently, deep learning processing was proposed as a completely new strategy to optimize the image quality of reconstructed CT datasets. The proprietary PixelShine™ (PS) algorithm is a software technology based on a deep, artificial neural network (AlgoMedica Inc., Sunnyvale, California, USA). The artificial neural network is trained at pixel level and learns the relationship between baseline and low dose datasets to determine a function that improves the image quality of the dataset.

In a previous study by our research group, we found that both the advanced modeled IR and the PS algorithm substantially improve the noise related image quality of CT protocols for maxillofacial trauma (3). Although outcomes such as the signal-to-noise and contrast-to-noise ratio quantify image noise, they are not directly related to how the fracture visualization is affected. The same is true for other more complex metrics such as noise power spectrum (NPS), noise equivalent quanta (NEQ), the modulation transfer function (MTF) and detective quantum efficiency (DQE) which are related more to the image formation process of the system (6). Thus, human observer studies were designed to perform a specific task or to quantify the image quality using a confidence rating scale. However, human observer studies are prone to the natural limitations of perception and observer predictions bias. Therefore, the Structural Similarity Index Metric (SSIM) was developed for the assessment of image quality (7,8).

SSIM measures the similarity between two images. It is based on luminance differences, contrast differences and structural variations and ranges from minus one (opposite contrast) to zero (completely different) to one (completely

identical) (6). Later, a multi-scale structural similarity approach (MS-SSIM) was proposed for the assessment of radiological images. It simulates different spatial resolutions by iterative downsampling and weighting the different values of each component (9). The aim of our study was to assess the feasibility of this metric to measure the structural image quality and fracture characteristics of midfacial fractures. These measures were assessed with reduced scan radiation dose protocols, and IR and deep learning reconstruction algorithms were used for image quality optimization.

We present the following article in accordance with the MDAR checklist (available at <https://dx.doi.org/10.21037/qims-21-564>).

Methods

Human cadaver specimen

A selection of four fresh frozen human cadaver specimens were obtained from the anatomy section of the Department of Neurosciences of the University Medical Center Groningen (University of Groningen, Groningen, the Netherlands), in accordance with our institute's regulations. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

Artificial infliction of fractures

Unilateral zygomaticomaxillary fractures were inflicted on each specimen using an experimental design whereby a free-falling mass was used to simulate a blunt facial trauma. In this experiment, the biomechanical tolerance of the zygomaticomaxillary complex to small and non-dislocated fractures was studied. More details regarding this experimental approach were described previously by our research group (10).

Computer tomography imaging

The specimens were scanned using a third generation Siemens SOMATOM Force Computed Tomography (CT) scanner (Siemens Healthcare AG, Erlangen, Germany). Each specimen was scanned using both a standard maxillofacial trauma protocol (ref mAs 50) and a reduced radiation dose reduced protocol where the reference mAs was reduced to the lowest limit (ref mAs 20). The raw data of both protocols were reconstructed using a model-based

Table 1 Siemens SOMATOM force CT acquisition and reconstruction parameters

Parameter	Value
Tube voltage	80 kV
Tube current modulation	CARE Dose4D
Quality reference mAs	50 & 20
ADMIRE strength	1, 3 & 5
Field of view	220.0 mm
Collimation	192×0.6 mm
Average scan length	118 mm
Slice thickness	0.6 mm
Position increment	0.4 mm
Grayscale depth	12 bit
Pitch	0.6
Rotation time	0.5 s
Exposure time	0.5 s
Scan time	3.4 s
Matrix	512×512
Reconstruction kernels	Bone Hr59d
Post-processing	PixelShine™ deep learning processing (v.1.2.57)

CT, computed tomography; ADMIRE, Advanced Modelled Iterative Reconstruction.

IR algorithm with strengths of one, three and five, the latter being the highest strength possible (ADMIRE, Siemens Healthcare AG, Erlangen, Germany) and the Hr59d bone kernel. Subsequently, all the datasets were subjected to the deep learning PS algorithm (v.1.2.57) to improve the image quality further. The datasets were all exported according to the Digital Imaging and Communications in Medicine (DICOM) standard. The acquisition and reconstruction parameters are summarized in *Table 1*.

Nondisplaced fracture selection

The datasets were assessed by an experienced independent European certified head and neck radiologist (BD) to identify fractures suitable for analysis, with particular attention being paid to minimal and non-dislocated fractures. The standard maxillofacial trauma datasets (ref. mAs 50) were assessed on a medical diagnostic display.

Osseous anatomy attenuation

All the datasets were imported into a Python software application for data analysis (Python Software Foundation. Python Language Reference, version 3.6.1. Available on <http://www.python.org>). A Hounsfield unit (HU) based sigmoidal soft thresholding function was applied so that only the osseous anatomy of all the datasets was the point of focus, using the following equation:

$$\left[1 - \exp\left(\frac{HU - 1200}{200}\right)\right]^{-1} \quad [1]$$

After attenuation, the effects of radiation dose reduction, IR strength and the use of the PS algorithms on fracture visibility were assessed in two experiments (*Figure 1*).

Structural image quality measurements [1]

In the first experiment, the MS-SSIM was used to assess the effects of radiation dose reduction, IR strength and the use of the PS algorithm on structural image quality. An area of interest was isolated, consisting of a 1,283-pixel cube, using the fracture as the centre. The MS-SSIM was calculated to compare the corresponding areas of interest of the varying radiation dose protocols (ref. mAs 50 & 20), ADMIRE strengths (1,3 & 5) and after applying the PS algorithm (no & yes).

Assessment of fracture characteristics (II)

In the second experiment, the MS-SSIM was used to measure the differences in fracture characteristics by comparing each fracture location with its contralateral anatomy. Using the fracture locations as the centre, a spherical area of interest was isolated using an 8-pixel radius. Accordingly, the uninjured side of the midface was computationally mirrored and superimposed on the side where a fracture had been inflicted. As we only wanted to measure the differences imposed by the inflicted fracture, registration was used to undo misalignment due to anatomical asymmetry and imperfect positioning of the specimens within the gantry during data acquisition. First, solid registration aligned the fracture and uninjured side orthogonally, initialized with manual estimations of the registration angles and offsets. Secondly, elastic registration corrected for misalignments due to global mismatching, solid registration and anatomical asymmetry between the fractured and uninjured sides (11). The MS-SSIM was

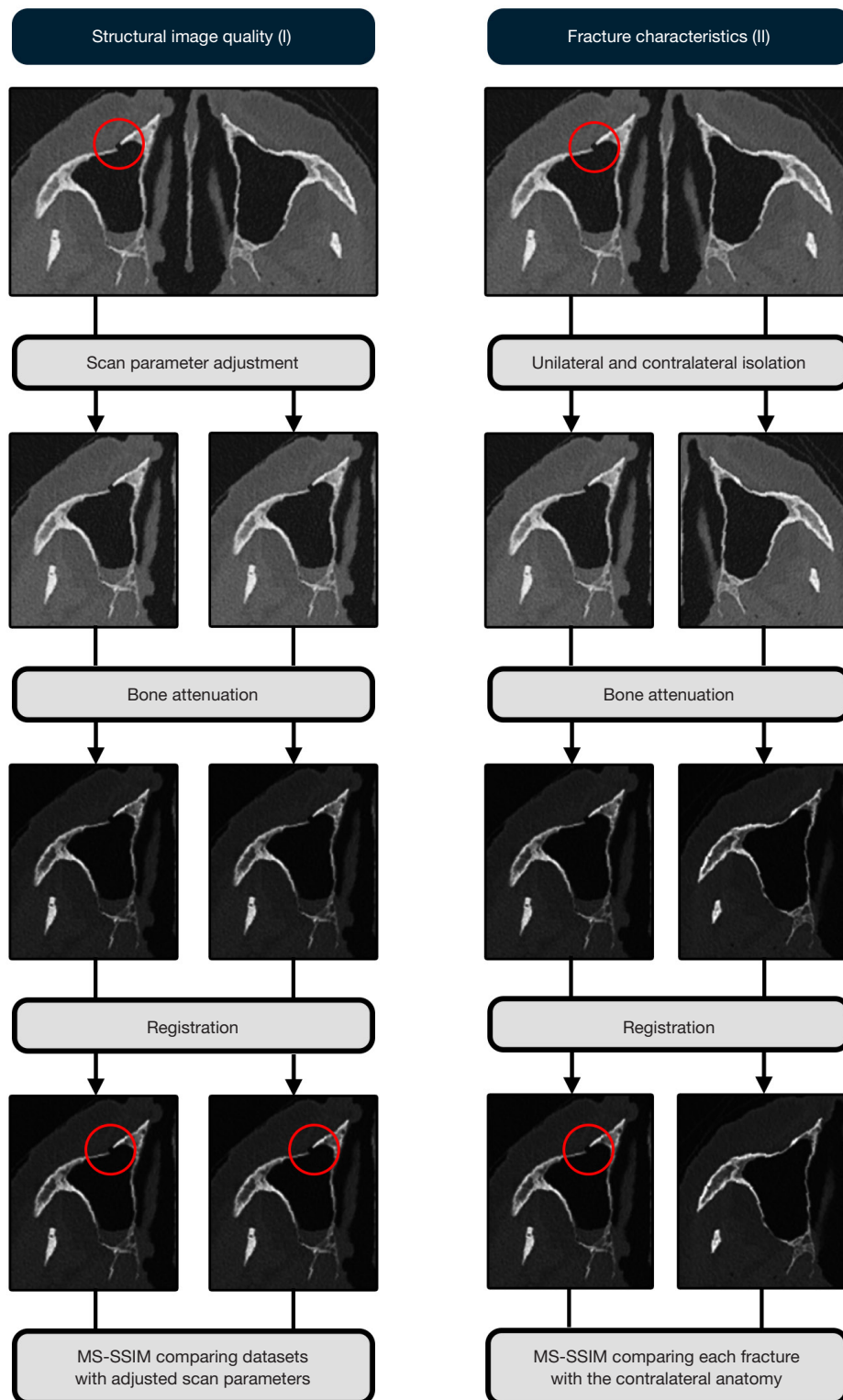


Figure 1 Study workflow for multi-scale structural similarity (MS-SSIM) index measurements (the fracture is an exaggerated example).

Table 2 MS-SSIM structural image quality outcome

Parameter	Reference	Comparison	MS SSIM (mean \pm SD)
Radiation dose protocol ¹	Ref. mAs 50	Ref. mAs 20	0.980036 \pm 0.011904
ADMIRE strength ²	1	3	0.999249 \pm 0.000170
	1	5	0.995399 \pm 0.001059
	3	5	0.998303 \pm 0.000401
PS use ³	No	Yes	0.999996 \pm 0.000002

¹, Only the ref. mAs 50 radiation dose was adjusted to ref. mAs 20 and the ADMIRE was kept at strength 1 and PixelShine was not used.

², Only the ADMIRE strength was adjusted from 1 to 3 and 5 and the radiation dose was kept at ref. mAs 50 and PixelShine was not used.

³, Only the PixelShine was used and the radiation dose was kept at ref. mAs 50 and the ADMIRE strength was kept at 1. MS-SSIM, multi-scale structural similarity index metric; SD, standard deviation; Ref. mAs, milliamperere-seconds reference; ADMIRE, advanced modeled iterative reconstruction; PS, PixelShine.

calculated and compared for each combination of datasets where the radiation doses, IR strength and PS algorithm had been adjusted. The experiments are summarized in *Figure 1*.

Statistical analysis

The data were analysed with the Statistical Package for the Social Sciences (IBM Corp. Released 2015. IBM SPSS Statistics for Windows, Version 23.0. Armonk, NY, USA: IBM Corp.). The MS-SSIM outcomes were presented as means and standard deviations. The normality was tested with Q-Q plots and the Kolmogorov-Smirnov test.

Results

Fractures

Zygomaxillary fractures were successfully inflicted on all four cadavers from which a total of twelve individual fracture locations were selected as eligible for data analysis. The fracture locations included the anterior maxillary sinus (n=6), posterior maxillary sinus (n=3), lateral orbital wall (n=4), and zygomatic arch (n=1).

Structural image quality measurements

The MS-SSIM outcomes are presented in *Table 2* and *Figure 2*. The greatest decrease in structural image quality was found for radiation dose reduction (0.980036 \pm 0.011904). Regarding the IR algorithm, a decrease in structural image quality occurred with an increase in strength, whereas the highest effect was found for the 1 to 5 ADMIRE strengths

(0.995399 \pm 0.001059). The effects of the PS algorithm were minimal because the MS-SSIM outcomes were almost identical to one (0.999996 \pm 0.000002).

Fracture characteristics analysis

The MS-SSIM outcomes are presented in *Table 3* and *Figure 2*. The MS-SSIM outcomes of all the datasets were below 0.90. Regarding radiation dose reduction and the IR algorithm, the MS-SSIM tended to increase, indicating higher structural similarity when comparing the fractured side to the contralateral non-fractured side. No clear trend was observed after applying the PS algorithm.

Discussion

The concept of structural similarity measurements is based on the assumption that the human visual system is highly adapted to extracting structural information from the scene whereas a measure of structural similarity can provide a good approximation of the perceived image quality (9). In this study the MS-SSIM was used to study structural image quality and changes to midfacial fracture characteristics. Scan protocols with reduced radiation doses were assessed and the IR and PS algorithms were used to improve the image quality of the outcome measures related to structural similarity. We found that the structural image quality worsened with radiation dose reduction, while the effects of IR and PS were small. Especially the MS-SSIM of the PS algorithm was observed to be exceptionally close to one, indicating a nearly identical structural image compared to datasets where the PS had not been applied.

This result is in line with the fracture characteristics

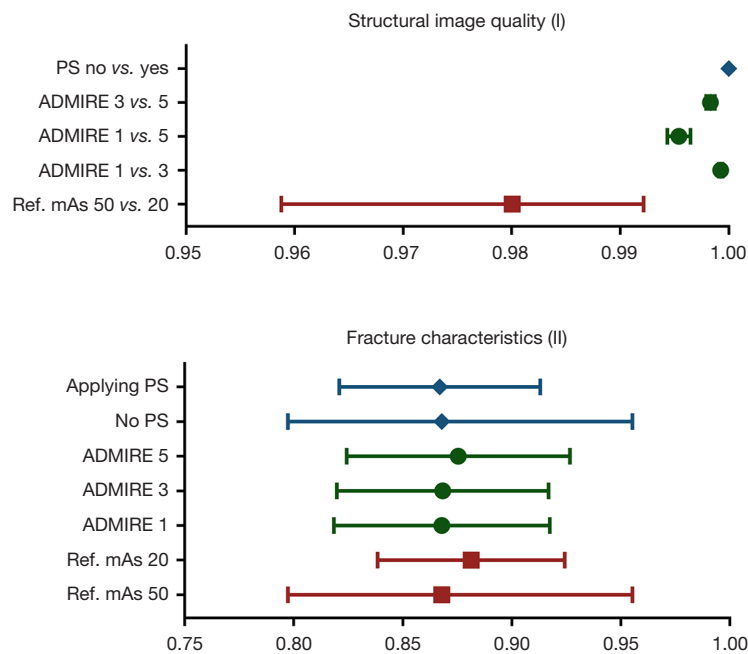


Figure 2 Multiscale structural similarity (MS-SSIM) index means and ranges for structural image quality (I) and fracture characteristics (II) analysis following radiation dose reduction (ref. mAs 50 and ref. mAs 20), Advanced Modeled Iterative Reconstruction strengths 1, 3 and 5 (ADMIRE) and PixelShine deep learning processing (no, yes).

Table 3 MS-SSIM fracture characteristics analysis outcome

Radiation dose	ADMIRE strength	PS used	MS SSIM (mean \pm SD) ¹
Ref. mAs 50	1	No	0.878369 \pm 0.043958
Ref. mAs 20	1	No	0.881375 \pm 0.042911
Ref. mAs 50	3	No	0.874766 \pm 0.047197
Ref. mAs 20	3	No	0.884703 \pm 0.046482
Ref. mAs 50	5	No	0.879656 \pm 0.037421
Ref. mAs 20	5	No	0.880935 \pm 0.037421
Ref. mAs 50	1	Yes	0.866955 \pm 0.046103
Ref. mAs 20	1	Yes	0.867904 \pm 0.049528
Ref. mAs 50	3	Yes	0.876707 \pm 0.046604
Ref. mAs 20	3	Yes	0.868314 \pm 0.048533
Ref. mAs 50	5	Yes	0.874656 \pm 0.048757
Ref. mAs 20	5	Yes	0.875457 \pm 0.051219

¹, the reference category involved the ref. mAs 50 protocol, IR strength 1 and not using the PS algorithm. MS-SSIM, Multi Scale Structural Similarity Index Metric; ADMIRE, Advanced Modeled Iterative Reconstruction; PS, PixelShine; SD, standard deviation; Ref. mAs, Milliampere-seconds Reference.

analysis where no clear trend was observed on applying the PS algorithm (*Figure 2*). This indicates that the PS algorithm denoises the dataset without affecting the fracture characteristics. Radiation dose reduction and IR strength tend to increase the MS-SSIM, indicating a greater similarity between the fracture and the non-fracture side. Although the effects were small, an MS-SSIM tendency towards ‘one’ indicates a potential smoothing or erasing of the individual fractures. Despite the above results, we found that the MS-SSIM for fracture characteristics’ analysis was lower than 0.90 for all the datasets, which might have been caused by the anatomical variance of the contralateral non-fractured side. Even though the variance was reduced to a minimum with solid and elastic registration, the differences in internal bony architecture could have also led to structural dissimilarities. One could consider unilateral analysis of the fracture side before and after inflicting the fracture. However, repositioning the specimen in the gantry can also lead to dissimilarities due to a variance in orthogonal resolution. The results of this feasibility study can be used for the conceptualization of future research to explore the effects of these scan parameters further on structural image quality and changes to fracture characteristics.

In conclusion, this feasibility human cadaver study of a structural similarity analysis of midfacial fractures provides an assessment method for structural image quality and fracture characteristics using the multi-scale structural similarity index metric. Structural image quality tends to be affected the most by radiation dose reduction, whereas the effects of IR and PS tend to be small. Fracture characteristic analyses was confined due to anatomical variances, but the data indicate that reducing the radiation dose and increasing the IR strength negatively affects the MS-SSIM, while the PS algorithm does not tend to affect this measure.

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Footnote

Reporting Checklist: The authors have completed the MDAR reporting checklist. Available at <https://dx.doi.org/10.21037/qims-21-564>

Conflicts of Interest: All authors have completed the ICMJE uniform disclosure form (available at <https://dx.doi.org/10.21037/qims-21-564>).

[org/10.21037/qims-21-564](https://dx.doi.org/10.21037/qims-21-564)). The authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). In this study, informed consent was not required because the study was conducted using human cadaver specimen. The specimens were obtained and approved according the regulations of the section anatomy of the Department of Neurosciences of the University Medical Center Groningen (University of Groningen, Groningen, the Netherlands).

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