Improving PSNR-based Quality Metrics Performance for Point Cloud Geometry

Instituições Associadas



universidade de aveiro



altice

NOKIA

UNIVERSIDADE BEIRA INTERIOR

U.PORTO

ISCTE 🕼 IUL Istituto Universitàrio de Lisboa

Alireza Javaheri

Catarina Brites Fernando Pereira João Ascenso

IEEE 27th International Conference on Image Processing (ICIP) 25th-28th October 2020, Abu-Dhabi, UAE



Outline

- 1. Introduction
- 2. PSNR-based Geometry Quality Metrics
- 3. Intrinsic Resolution PSNR-based Quality Metrics
- 4. Resolution Adaptive PSNR-based Quality Metrics
- 5. Performance Evaluation
- 6. Conclusions





1. Introduction





Context and Motivation

- Efficient compression is necessary to deal with large amount of data for point cloud (PC) in most applications
- Quality assessment is fundamental to evaluate the performance of the compression schemes



- Motivation
 - PC intrinsic characteristics play an important role on the final perceived quality
 - MPEG D1 and D2 PSNR do not consider the content type, number of points and distribution of the points (sparse vs. dense)





Objective and Contibutions

The objective of this paper is to propose new quality metrics that measure the level of geometry degradation of PCs with different characteristics, e.g. content type, number of points and distribution of the points (sparse vs. dense) to achieve higher performance

Contributions

- Propose and evaluate several geometry quality PSNR-based metrics that exploit the *intrinsic characteristics* of a PC
- Propose and evaluate a novel geometry quality PSNR-based metric that exploits the way that PCs are typically rendered





2. PSNR-Based Geometry Quality Metrics





MPEG PC Geometry Quality Metrics – D1 PSNR

• Point-to-point Euclidean distance is measured between each point in *PC A* and its nearest neighbor in *PC B*

- Directed Mean Squared Error (MSE) between all points to their nearest neighbor is computed in two directions:
 - Original (A) to degraded (B)
 - Degraded (B) to original (A)
- **D1** is the <u>maximum</u> of two directed distances:
- **D1 PSNR** is computed using D1 distance:

 $d_{A,B}^{Po2Po} = \|\vec{e}(i,j)\|_{2}^{2}$ $d_{A,B}^{MSE} = \frac{1}{N_{A}} \sum_{\forall a_{i} \in A} d_{A,B}^{Po2Po}(i)$ $d_{B,A}^{MSE} = \frac{1}{N_{B}} \sum_{\forall b \in D} d_{B,A}^{Po2Po}(i)$

 $D1 = max (d_{A,B}^{MSE}, d_{B,A}^{MSE})$

 $D1 PSNR = 10 \log_{10}(\frac{p_s^2}{D1})$





MPEG PC Geometry Quality Metrics – D2 PSNR

• Point-to-plane distance is computed as projection of the point-to-point Euclidean distance vector onto the normal vector at reference point $d_{AB}^{Po2Po} = \|\vec{e}(i, i)\|_{2}^{2}$

$$d_{A,B}^{Po2Pl} = \|\hat{e}(i,j)\|_{2}^{2} = (\vec{e}(i,j).\vec{n}_{j})^{2}$$



- Directed MSE between all points to their nearest neighbor is computed in two directions:
 - Original (A) to degraded (B)
 - Degraded (B) to original (A)
- D2 is <u>maximum</u> of two <u>projected</u> directed distances:
- **D2 PSNR** is computed on D2 distance:









MPEG PC Geometry Quality Metrics – PSNR Peak

- Non-voxelized
 - The largest diagonal (LD) distance of the PC bounding box.

$$p_{s} = LD = \left\| (x_{max}, y_{max}, z_{max}) - (x_{min}, y_{min}, z_{min}) \right\|_{2}$$

- Voxelized
 - Point coordinates are bounded between zero and a constant integer related to the PC precision (*b*)
 - Coordinate peak $p_c = 2^b 1$
 - Signal peak

$$p_s = ||(p_c, p_c, p_c) - (0, 0, 0)||_2 = \sqrt{3}p_c$$





3. Intrinsic Resolution PSNR-Based Quality Metrics (I-PSNR)





Intrinsic Resolution PSNR-based Quality Metric

Intrinsic Resolution

Measure of distance between points within a PC

Intrinsic Resolution Estimation

- Maximum nearest neighbour¹ (MNN)
 - Sensitive to holes and local sparse areas
- Average nearest neighbor (ANN) proposed
- Average of K nearest neighbors (ANN_k) proposed

Quality Metric Design

Intrinsic resolution is used as PSNR peak



$$I - PSNR = 10 \log_{10} \left(\frac{p_s^2}{d^{MSE}} \right)$$

4. Resolution Adaptive PSNR-Based Quality Metrics (RA-PSNR)





Rendering Resolution

- PCs are rendered as images or videos from one or more viewpoints before being shown on a 2D display
- PCs are always evaluated by the users after rendering:
 - The final perceived quality also strongly depends on the rendering process
- Rendering resolution may vary for different parts of the PC because of the orientation of the PC surfaces in the 3D world relative to the observer viewing location

Rendering Resolution Estimation Procedure

- 1. Define a local neighborhood around point a_i , which includes the k nearest neighboring points
- 2. Define a local plane tangent at point a_i . This plane is perpendicular to the normal vector \vec{n}_i and represents the PC surface at this point
- 3. Project all distance vectors $(\vec{d}_{i,j})$ between point a_i and its jth nearest neighbour (planar distance) $\overrightarrow{PD_i} = Proj_{plane}(\vec{d}_{i,j}) = \vec{d}_{i,j} - Proj_{\vec{n}_i}(\vec{d}_{i,j})$
- 4. Estimate the rendering resolution as the average (planar) distance between point a_i and their k local neighbors on the tangent plane

$$APD_{k} = \frac{1}{N_{O}} \sum_{i \in O} \left(\frac{1}{k} \sum_{j=1}^{k} \left\| \overrightarrow{PD}_{i,j} \right\|_{2}^{2} \right)$$

ÉCNICO



Normal vector \vec{n}_i



RA-PSNR PC Quality Metric Design

RA-PSNR estimates the PC intrinsic quality by exploiting the <u>intrinsic</u> or <u>rendering</u> resolution and compensates changes in the <u>precision</u>

• Density coefficient μ corresponds to the normalization of the intrinsic/rendering resolution according to the PC precision

$$\iota = \frac{p_c}{r}$$

• RA-PSNR scales MSEs according to μ as well as p

$$RA - PSNR = 10 \log_{10} \frac{3p_c^2}{\mu d^{MSE}} = 10 \log_{10} \frac{3rp_c}{d^{MSE}}$$





5. Performance Evaluation





IST Rendered Point Cloud (IRPC) Quality Assessment Dataset

- Content: 6 PCs from MPEG voxelized PCs
- Codecs and Rates:
 - Octree-based codec from PCL
 - MPEG G-PCC (TriSoup)
 - MPEG V-PCC codec



PC Name	No. Points	Precision	Category		
Egyptian Mask	272,684	12 bit	Inanimate Objects		
Frog	3,614,251	12 bit	Inanimate Objects		
Facade9	1,596,085	12 bit	Facades & Buildings		
House without roof	4,848,745	12 bit	Facades & Buildings		
Loot	805,285	10 bit	People		
Longdress	857,966	10 bit	People		

- 3 distinguishable qualities using the rates defined in MPEG CTC
- **RPoint rendering:** PCs are rendered using a point representation with uniform color and shading
- **Subjects:** 20, 18 common between sessions
- Methodology: Double Stimulus Impairment Scale (DSIS)
- Camera Path: Video sequences are created by rotating the object

instituto de telecomunicaçã



Correlation Performance of the Proposed Metrics

- I-PSNR outperforms MPEG PSNRs
- I-PSNR using proposed intrinsic resolution estimations is performing better than I-PSNR using MNN

Metric	Variant	Туре	PCL		G-PCC		V-PCC		All	
			PLCC	SROCC	PLCC	SROCC	PLCC	SROCC	PLC	SROCC
DCND	Р	Po2Po	87.0	73.9	86.9	87.4	53.1	62.0	67.3	64.7
		Po2Pl	89.6	80.9	83.4	85.6	51.4	49.6	70.3	65.6
FSINK	LD	Po2Po	83.3	82.3	86.0	89.3	48.9	54.1	70.4	68.6
		Po2Pl	86.7	85.9	75.6	71.9	59.9	58.9	71,4	67.2
	MNN	Po2Po	68.7	65.0	40.8	33.3	45.6	14.3	49.7	42.2
		Po2Pl	69.6	66.7	44.1	39.7	49.8	25.4	52.1	43.4
LDCND	ANN	Po2Po	92.2	89.0	79.3	76.3	66.6	61.0	64.7	52.5
I-PSINK		Po2Pl	92.3	84.5	86.6	79.0	76.2	69.1	66.4	55.6
RA-PSNR	ANNk	Po2Po	88.8	87.0	78.1	68.4	70.8	62.9	66.5	59.2
		Po2Pl	90.8	87.5	75.3	68.2	79.6	74.0	67.4	62.6
	ANN	Po2Po	92.8	86.8	84.9	82.5	49.2	45.6	68.9	64.0
		Po2Pl	93.4	86.7	88.5	85.8	67.9	63.0	71.0	67.1
	ANNk	Po2Po	94.1	86.2	93.6	94.6	68.5	62.5	74.1	71.1
		Po2Pl	95.1	91.3	94.0	94.0	77.1	73.7	74.9	72.5
	APDk	Po2Po	94.1	85.2	93.7	94.8	59.9	59.7	74.0	71.4
		Po2Pl	95.2	91.7	94.0	94.0	77.9	72.3	75.6	73.8

PLCC and SROCC correlation performance of objective metrics after cubic fitting





Correlation Performance of the Proposed Metrics

- RA-PSNR outperforms MPEG PSNR metrics and I-PSNR significantly
- APD which considers rendering resolution performs better, especially for All codecs together

S Metric		Variant	Type	PCL		G-PCC		V-PCC		All	
Ŭ	WIEUIC	v al lalli	турс	PLCC	SROCC	PLCC	SROCC	PLCC	SROCC	PLC	SROCC
		Р	Po2Po	87.0	73.9	86.9	87.4	53.1	62.0	67.3	64.7
	DCND		Po2Pl	89.6	80.9	83.4	85.6	51.4	49.6	70.3	65.6
	FSINK	ID	Po2Po	83.3	82.3	86.0	89.3	48.9	54.1	70.4	68.6
		LD	Po2Pl	86.7	85.9	75.6	71.9	59.9	58.9	71,4	67.2
ĺ		MNN	Po2Po	68.7	65.0	40.8	33.3	45.6	14.3	49.7	42.2
		IVIININ	Po2Pl	69.6	66.7	44.1	39.7	49.8	_25.4	52.1	43.4
	I DSND	ANN	Po2Po	92.2	89.0	79.3	76.3	66.6	61.0	64.7	52.5
	I-F SINK		Po2Pl	92.3	84.5	86.6	79.0	76.2	69.1	66.4	55.6
		ANNk	Po2Po	88.8	87.0	78.1	68.4	70.8	62.9	66.5	59.2
			Po2Pl	90.8	87.5	75.3	68.2	79.6	74.0	67.4	62.6
ĺ		ANN	Po2Po	92.8	86.8	84.9	82.5	49.2	45.6	68.9	64.0
	RA-PSNR	AININ	Po2Pl	93.4	86.7	88.5	85.8	67.9	63.0	71.0	67.1
		ANNk	Po2Po	94.1	86.2	93.6	94.6	68.5	62.5	74.1	71.1
			Po2Pl	95.1	91.3	94.0	94.0	77.1	73.7	74.9	72.5
		APDk	Po2Po	94.1	85.2	93.7	94.8	59.9	59.7	74.0	71.4
			Po2Pl	95.2	91.7	94.0	94.0	77.9	72.3	75.6	73.8

PLCC and SROCC correlation performance of objective metrics after cubic fitting





Correlation Performance of the Proposed Metrics

- PCL and G-PCC are mix of 10-bit and 12bit, but V-PCC is all 10-bit
- Normalizing by Precision works as a distractor and decreases the performance

Metric	Variant	Туре	PCL		G-PCC		V-PCC		All	
			PLCC	SROCC	PLCC	SROCC	PLCC	SROCC	PLC	SROCC
	Р	Po2Po	87.0	73.9	86.9	87.4	53.1	62.0	67.3	64.7
DOND		Po2Pl	89.6	80.9	83.4	85.6	51.4	49.6	70.3	65.6
FSINK	ID	Po2Po	83.3	82.3	86.0	89.3	48.9	54.1	70.4	68.6
	LD	Po2Pl	86.7	85.9	75.6	71.9	59.9	58.9	71,4	67.2
	MNINI	Po2Po	68.7	65.0	40.8	33.3	45.6	14.3	49.7	42.2
	IVIININ	Po2Pl	69.6	66.7	44.1	39.7	49.8	25.4	52.1	43.4
I DOND	ANN	Po2Po	92.2	89.0	79.3	76.3	66.6	61.0	64.7	52.5
I-F SINK		Po2Pl	92.3	84.5	86.6	79.0	76.2	69.1	66.4	55.6
	ANNk	Po2Po	88.8	87.0	78.1	68.4	70.8	62.9	66.5	59.2
		Po2Pl	90.8	87.5	75.3	68.2	79.6	74.0	67.4	62.6
RA-PSNR	ANN	Po2Po	92.8	86.8	84.9	82.5	49.2	45.6	68.9	64.0
		Po2Pl	93.4	86.7	88.5	85.8	67.9	63.0	71.0	67.1
	ANNk	Po2Po	94.1	86.2	93.6	94.6	68.5	62.5	74.1	71.1
		Po2Pl	95.1	91.3	94.0	94.0	77.1	73.7	74.9	72.5
	APDk	Po2Po	94.1	85.2	93.7	94.8	59.9	59.7	74.0	71.4
		Po2Pl	95.2	91.7	94.0	94.0	77.9	72.3	75.6	73.8

PLCC and SROCC correlation performance of objective metrics after cubic fitting





6. Conclusions





Conclusions

- State-of-the-art PC geometry quality assessment metrics perform poorly
- Impact of the following factors on the final perceived PC quality are usually not considered:
 - Intrinsic PC characteristics
 - <u>The rendering process</u>
- In this work, the popular PSNR-based metrics are improved by including a normalization factor that accounts for changes in the intrinsic PC resolution after rendering, as well as the PC precision
- Experimental results show that the proposed metrics outperform state-of-the-art MPEG quality metrics by a significant margin





Thank you For your Attention

email: alireza.javaheri@lx.it.pt



