







Learning-based Point Cloud Decoding With Independent and Scalable Reduced Complexity

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IEEE International Conference on Image Processing, Abu Dhabi, UAE | October 27-30, 2024





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- 3. Proposed Reduced Complexity PC Decoding Solutions
- 4. Performance Assessment
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What is a Point Cloud?

- * A point cloud (PC) is a 3D visual representation model consisting of an unordered set of 3D points, characterized by their coordinates (x, y, z).
- * Target applications include virtual/augmented reality for entertainment and communication, autonomous driving, and among others.





Why Point Cloud Coding is a Must?

- * PCs may need millions of points to offer accurate representations, realism, and immersion, requiring very large amounts of data.
- * Efficient PC coding solutions are essential for practical storage and transmission.





JPEG PCC Standard

- * JPEG recently finalized the JPEG Pleno Learning-based Point Cloud Coding (JPEG PCC) standard.
- * JPEG PCC offers improved Rate-Distortion (RD) performance, notably outperforming the MPEG PC coding standards, particularly for geometry coding of solid static PCs.





Learning-based Coding Challenge: Model Complexity

Domain 🔵 Language 😑 Vision

 DL-based codecs have become state-of-the-art in many fields, including image and PC coding.

 Nevertheless, their high model complexity presents challenges for decoding in resource-constrained devices, such as smartphones.



Image source: https://www.worldbank.org/



Propose two novel reduced complexity decoding solutions, one of them based on the adoption of scalability principles, to decode the same JPEG PCC compliant bitstream with minimal impact on RD performance.







JPEG PCC Standard



JPEG PCC Standard Architecture: Geometry-Only





JPEG PCC DL-based Coding Model

Autoencoder Variational Autoencoder s:2 s:2 s:2 s:2 s:2 ŝ ŝ f:128 28 Scaling 28 28 28 f:32 f:64 f:1 — Input Block ÷ ÷ k:3×3×3 ReLU ReLU ReLU ReLU k:3×3×3 ReLU \mathfrak{C} k:3×3×3 Conv k: $3 \times 3 \times 3$ Conv k:3×3×3 IRB IRB IRB $\overline{\times}$ $3 \times 3 \times 3$ Rounding → Rounding \times Ľ. Ľ. Conv Conv Conv Conv Conv RE ≪. RE 1V ıÔ Side Info s:2 s:2 s:2 Bitstream s:2 s:2 ŝ Bitstream s:1 92 28 28 Conv k:3×3×3 f:256 f:64 32 <u>f:</u> Decoded Block f:1 f:1 ÷ ÷ Sigmoid ReLU ReLU ReLU ReLU k:1×1×] ReLU k:3×3×3 k:3×3×3 k:3×3×3 \mathcal{O} \mathbf{c} IRB IRB IRB ў Х $\times 3\times$ **K**.... RD RD X k:3 k:3 Conv Conv Conv Conv] Conv Conv Scaling⁻¹



JPEG PCC DL-based Coding Model: Bitstream

Autoencoder Variational Autoencoder s:2 s:2 s:2 s:2 s:2 ŝ S Scaling 28 28 28 $\frac{28}{28}$ 28 N N N f:64 32 fil fil f:1 -Input Block ÷ ÷ \sim ReLU k:3×3×3 ReLU DALI ReLU ReLU k:3×3×3 Conv k:3×3×3 \mathbf{c} IRB IRB IRB X X Rounding Bitstream is generated by k:3× Х $\stackrel{(\times)}{\approx}$ Conv k: coding the entropy Ň latent Conv Conv Conv onv RE K..... representations conditioned by Gaussian parameters produced IV by hyper-synthesis transform. s:2 s:2 Bitstream s:2 s:2 ŝ Bitstream s:1 S 256 28 28 92 f:64 \sim Decoded Block 3 ÷ ÷ ÷ ÷ ÷ Sigmoid ÷ k:1×1×1 $3 \times 3 \times 3$ $3 \times 3 \times 3$ ReLU ReLU ReLU ReLU ReLU k:3×3×3 k:3×3×3 \mathcal{O} IRB IRB IRB $\overset{\circ}{\times}$ RD RD **K**..... ∞ **RE: Range Encoder** X X <u>k:3</u> \mathfrak{C} Ň <u>.</u> M Conv Conv Conv Conv] Conv Conv Conv **RD: Range Decoder** Scaling⁻¹



JPEG PCC Decoding Model Complexity

Autoencoder Variational Autoencoder s:2 s:2 s:2 S: I S: s:2 s:2 f:128 f:128 Conv k:1×1×1 f:128 Scaling 28 28 À f:64 f:32 F. f:1 Input Block ReLU Rounding $\stackrel{\times}{\sim}$ k:3×3×3 ReLU ReLU ReLU ReLU Conv k:3×3×3 Conv k:3×3×3 Conv k:3×3×3 IRB IRB IRB Rounding k:3× Conv Conv Conv] RE <------RE 1V ЧĊ Side Info s:2 s:2 s:2 Bitstream s:2 s:2 S Bitstream S 256 f:192 128 28 f:64 f:32 Decoded Block نت f:1 Sigmoid ReLU ReLU ReLU k:3×3×3 ReLU k:3×3×3 k:3×3×3 ReLU k:3×3×3 \mathbf{c} IRB IRB IRB $\frac{3}{2}$ RD <.... RD $\overline{\times}$ k:3× <u>k:</u> onv Conv Conv Conv Conv Vuo Scaling⁻¹

★ The decoding model complexity in terms of number of parameters is 2783089 (~2.8M)!

 ★ Devices with limited computational resources may struggle to deal with this level of complexity.



Proposed Reduced Complexity PC Decoding Solutions





Proposed Reduced Complexity Decoding Solutions





Independent Complexity JPEG PC Decoding Solution

- * Independent Complexity JPEG PC Decoding (IC-JPCD) provides multiple independent reduced complexity decoder models.
- * Independent reduced complexity is achieved by reducing a given percentage of the parameters associated with the filters in each layer of the JPEG PCC synthesis transform.
- * Each independent model offers a trade-off between the complexity and RD performance.





Scalable Complexity JPEG PC Decoding Solution

- * Scalable Complexity JPEG PC Decoding (SC-JPCD) uses one single scalable decoder model, designed with a hierarchical structure of interdependent layers over a base layer.
- * Scalability is achieved by progressively learning a set of additional parameters of the synthesis transform for each level, while reusing the parameters of the previous complexity level.
- * Consumes less memory compared to the previous independent approach as it requires only one full model for all decoding complexity levels.









Designing Varying Complexity PC Decoders: A Three Phases Process

The designing of the two strategies for the two reduced complexity decoding solutions, independent and scalable, includes three phases:





Regularization: Objective and Process



- * The model parameters vary in their impact on RD performance, with some having more influence than others.
- * The objective of the regularization step is to determine the filters in each layer that contribute the least to the RD performance by minimizing the norm of the parameters associated with those filters.
- * The initial JPEG PCC models are trained using a new RD loss function integrating an L1 regularization term.

* Loss = Distortion +
$$\lambda$$
 * Coding Rate + ρ * $\sum |w|$
RD Loss function L1 regularization



Pruning: Objective and Process



- * The objective of the pruning is to remove the least important filters from each layer of synthesis transform.
- * Global structured pruning is adopted, which means that individual parameters are not pruned, but rather the entire filters over all the layers of synthesis transform.
- * Consequently, each layer may have a different number of pruned filters depending on their impact on the RD performance.





IRB2 –

Filters are gathered from each layer of synthesis transform.









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Pruning Strategies: Full Freedom vs Layer Restriction

Full Freedom:

Filters are eliminated without any constraints, based on the sorted list of filters.

Filter Impact Pruning

Consequence:

Some layers will be entirely removed.



Full Decoder Model vs Synthesis Transform Model





Training: Objective and Process



- * The objective is to train the new pruned architecture (per target rate) to achieve the best possible RD performance.
- * For each **independent** decoder layer:
 - Encoder part is initialized with JPEC PCC encoder model which is kept frozen.
 - Then, the pruned synthesis transform architecture is randomly initialized and trained.
- * For each <u>scalable</u> decoder layer:
 - For scalable decoder base layer, the training approach is the same as for the independent solution.
 - All the synthesis transform filters from the previous scalable layer are initialized and kept frozen.
 - Then, the new filters (depending on the target percentage) are trained with random initialization.

Lower Complexity Non-Compliant Decoders







Performance

Assessment





Decoding Solutions Under Comparison

* Anchors:

- JPEG PCC VM v3
- MPEG G-PCC Octree v20
- MPEG V-PCC Intra v20
- ***** Independent Decoding Solution
 - IC-JPCD: 30% (Layer Restriction, Full Freedom)
 - IC-JPCD: 20% (Layer Restriction, Full Freedom)
 - IC-JPCD: 10% (Layer Restriction, Full Freedom)
- ***** Scalable Decoding Solution
 - SC-JPCD: 30% (Layer Restriction)
 - SC-JPCD: 20% (Layer Restriction)
 - SC-JPCD: 10% (Layer Restriction)



RD Performance: JPEG PCC vs MPEG PCC



RD Performance: Independent (IC-JPCD) Solution



RD Performance: Independent (IC-JPCD) Solution





JPEG PCC vs Independent Reduced Complexity Non-Compliant Decoders (avg. over all test dataset)

	Druming	PSNR D1		PSNR D2		Nº of Model P	Deceding		
Solution	Strategy	BD-Rate	BD-PSNR	BD-Rate	BD-PSNR	Synthesis Transform	Full Decoder	All Models	Time (GPU)
IC-JPCD: 10%	Full	4.8%	-0.14	5.3%	-0.18	-23.51%	-9.51%	+240.22%	-10.6%
IC-JPCD: 20%		6.8%	-0.19	7.8%	-0.25	-50.15%	-20.32%		-12.93%
IC-JPCD: 30%	i i cedoiii	14.3%	-0.39	14.1%	-0.45	-74.18%	-30.05%		-15.16%
IC-JPCD: 10%	Layer	4.1%	-0.12	5.8%	-0.18	-24.80%	-10.06%	+240.85%	-7.75%
IC-JPCD: 20%		6.9%	-0.18	7.0%	-0.22	-49.67%	-20.13%		-9.91%
IC-JPCD: 30%		13.1%	-0.34	12.5%	-0.39	-71.47%	-28.96%		-11.88%



JPEG PCC vs Independent Reduced Complexity Non-Compliant Decoders (avg. over all test dataset)

			Smaller penalty in RD performance.			Lower complexity reduction.		Lower of time red	decoding duction.
DSNR D1						Nº of Model ramete		and Memory Siz	
Solution	Pruning Strategy	BD-Rate	BD-PSNR	BD-Rate	BD-PSNR	Synthesis Transform	Full Deco	der All Models	Decoding Time (GPU)
IC-JPCD: 10%		4.8%	-0.14	5.3%	-0.18	-23.51%	-9.51%		-10.6%
IC-JPCD: 20%	Full Freedom	6.8%	-0.19	7.8%	-0.25	-50.15%	-20.32% -30.05%	+240.22%	-12.93%
IC-JPCD: 30%	riccuom	14.3%	-0.39	14.1%	-0.45	-74.18%		0	-15.16%
IC-JPCD: 10%	_	4.1%	-0.12	5.8%	-0.18	-24.80%	-10.06%	6	-7.75%
IC-JPCD: 20%	Layer Restriction	6.9%	-0.18	7.0%	-0.22	-49.67%	-20.13%	+240.85%	-9.91%
IC-JPCD: 30%	Reserveron	13.1%	-0.34	12.5%	-0.39	-71.47%	-28.96%	0	-11.88%
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RD Performance: Scalable (SC-JPCD) Solution



RD Performance: IC-JPCD vs SC-JPCD Solution





JPEG PCC vs Scalable Reduced Complexity Non-Compliant Decoders (avg. over all test dataset)

Smaller per performan				nalty in RD ce.		Lower con reduction.	nplexity	Lower time re	decoding eduction.
	Pruning Strategy	PSN	PSNR D1 PSNR D2 Nº of Model rameters and Memo		nd Memory Siz	Deceding			
Solution		BD-Rate	BD-PSNR	BD-Rate	BD-PSNR	Synthesis Transform	Full Decode	All Models	Time (GPU)
SC-JPCD: 0%		9.5%	-0.26	9.4%	-0.30	0%	0%		-6.67%
SC-JPCD: 10%	Layer	10.9%	-0.30	11%	-0.36	-24.80%	-10.06%	0%	-7.54%
SC-JPCD: 20%	Restriction	9.9%	-0.28	10.0%	-0.33	-49.67%	-20.13%	0%	-9.35%
SC-JPCD: 30%		13.1%	-0.34	12.5%	-0.39	-71.47%	-28.96%		-11.88%



The price of having a scalable complexity decoder.

		PSNR D1		PSNR D2		Nº of Model Parameters and Memory Size			Deceding
Solution	Strategy	BD-Rate	BD-PSNR	BD-Rate	BD-PSNR	Synthesis Transform	Full Decoder	All Models	Time (GPU)
SC-JPCD: 0%		9.5%	-0.26	9.4%	-0.30	0%	0%	0%	-6.67%
SC-JPCD: 10%	Layer	10.9%	-0.30	11%	-0.36	-24.80%	-10.06%		-7.54%
SC-JPCD: 20%	Restriction	9.9%	-0.28	10.0%	-0.33	-49.67%	-20.13%		-9.35%
SC-JPCD: 30%		13.1%	-0.34	12.5%	-0.39	-71.47%	-28.96%		-11.88%



JPEG PCC vs Lower Complexity Non-Compliant Decoders (avg. over all test dataset)

	Dunning	PSN	R D1	PSNR D2		Nº of Model F	Deceding		
Solution	Strategy	BD-Rate	BD-PSNR	BD-Rate	BD-PSNR	Synthesis Transform	Full Decode	All Models	Time (GPU)
IC-JPCD: 10%		4.8%	-0.14	5.3%	-0.18	-23.51%	-9.51%		-10.6%
IC-JPCD: 20%	Full Freedom	6.8%	-0.19	7.8%	-0.25	-50.15%	-20.32%	+240.22%	-12.93%
IC-JPCD: 30%	riccuom	14.3%	-0.39	14.1%	-0.45	-74.18%	-30.05%		-15.16%
IC-JPCD: 10%	Layer	Scalab	le comple	exity deco	ding offe	rs a single	e full ^{5%}		-7.75%
IC-JPCD: 20%		model	with no n	nemorv si	ize increa	se regard	ing ^{3%}	+240.85%	-9.91%
IC-JPCD: 30%	Restriction	the no	n-adantiv	e JPEG F	PCC mod	el.	5%		-11.88%
SC-JPCD: 0%		9.5%	-0.20	9.470	-0.50	U70	U70		-6.67%
SC-JPCD: 10%	Layer	10.9%	-0.30	11%	-0.36	-24.80%	-10.06%	00/	-7.54%
SC-JPCD: 20%	Restriction	9.9%	-0.28	10.0%	-0.33	-49.67%	-20.13%	U%	-9.35%
SC-JPCD: 30%		13.1%	-0.34	12.5%	-0.39	-71.47%	-28.96%		-11.88%





Conclusions



- * The IC-JPCD and SC-JPCD decoders offer reduced independent and scalable complexity decoding solutions at the cost of a minimal RD performance penalty, while consuming the same JPEG PCC compliant bitstreams.
- * Better RD performance could be achieved by re-training the full end-to-end model (including the encoder) for the target decoding complexity, but this would violate the compliance constraint for the bitstream.
- * Scalable complexity decoders offer more versatility compared to the independent complexity decoders for the same memory size, as they can adapt based on available computational resources.







Thank you for your attention!

FCT Fundação para a Ciência e a Tecnologia

ACKNOWLEDGMENT

This work has been financially supported by the Fundação para a Ciência e a Tecnologia (FCT, Portugal) through the research project PTDC/EEI-COM/1125/2021, entitled "Deep Learning-based Point Cloud Representation".