

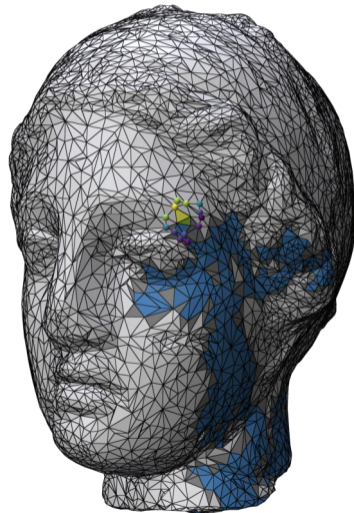
Priority-based encoding of triangle mesh connectivity for a known geometry

(and beyond)

J. Dvořák, Z. Káčereková, P. Vaněček & L. Váša

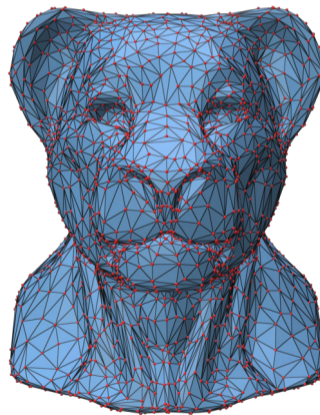
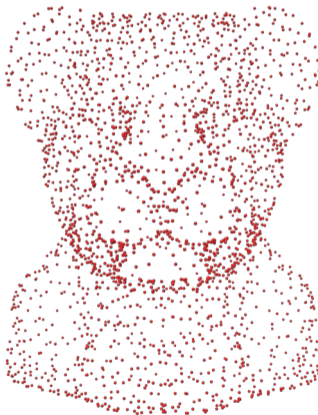
Department of Computer Science and Engineering
Faculty of Applied Sciences
University of West Bohemia
Pilsen, Czech Republic

May 19, 2023

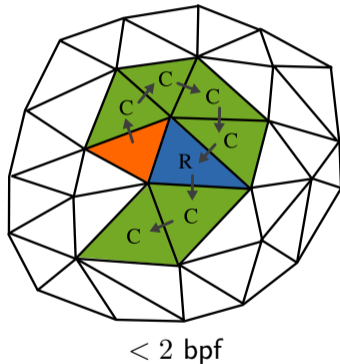


Geometry encoded first

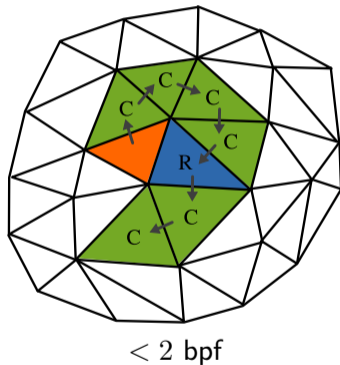
- Temporal prediction
- Predictable spatial structure
- Multiple-rate compression



Simple, does not require geometry information



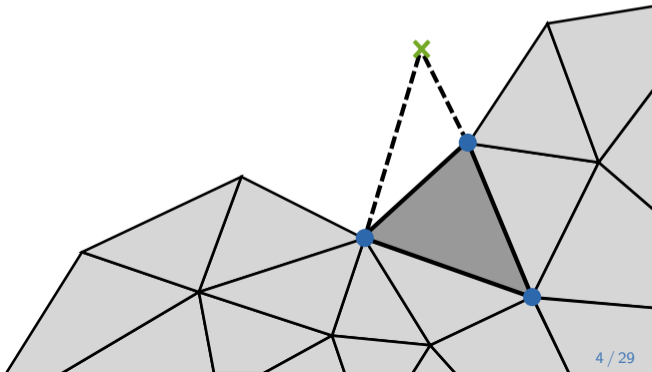
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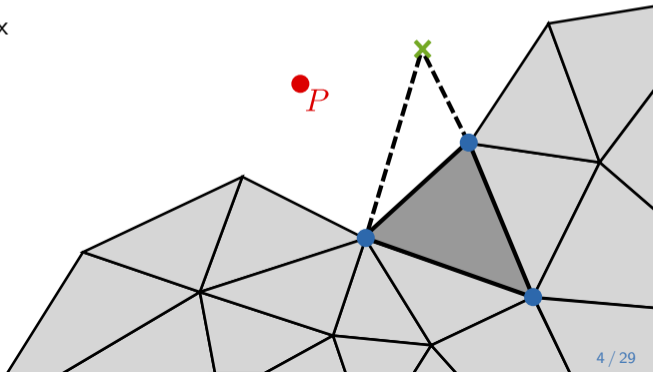
Causes reordering!

Permutation map: $\frac{1}{F} \log_2 (V!) \text{ bpf}$

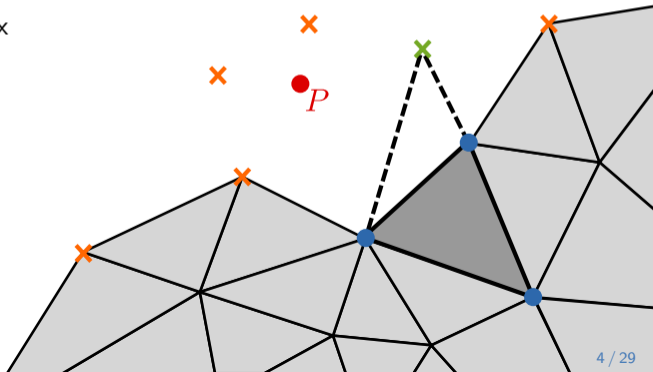
- Marais et al. 2007
- Fixed connectivity traversal through edges



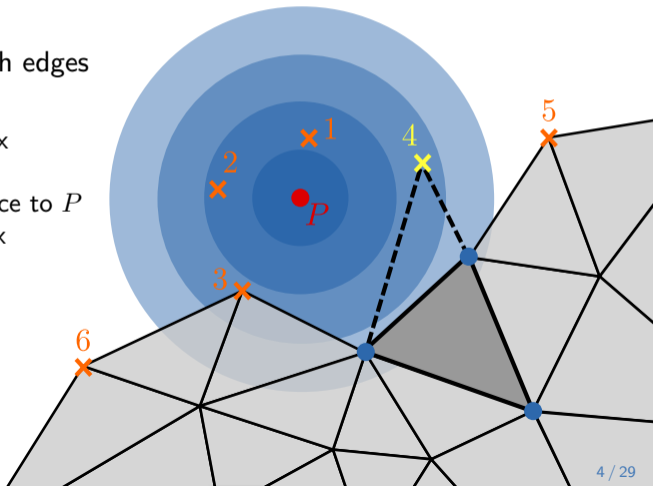
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- To encode a triangle:
 - 1 Make a prediction P of tip vertex



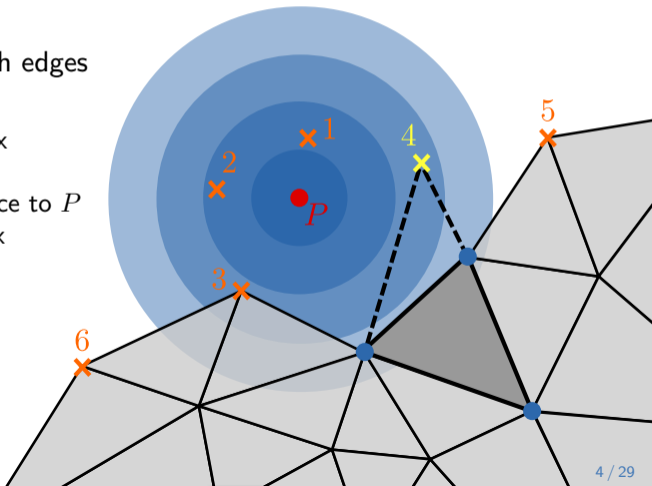
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- Symbol **0** reserved for boundary



- Directly built upon Distance-ranked algorithm

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- Fixed traversal → Priority-driven traversal

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- Boundary edge prediction

Priority-driven traversal

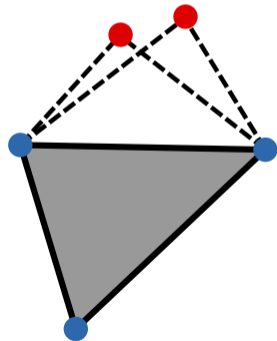
Triangles where encoder most certain processed first

$$p = q_{\max} - q_{\max 2}$$

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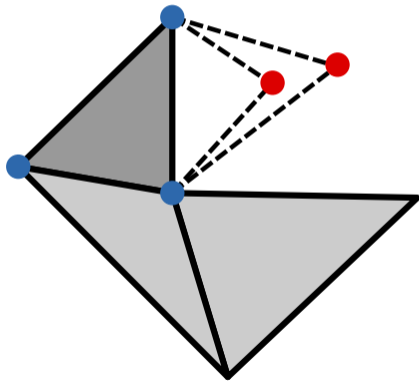
- More feasible situation in future



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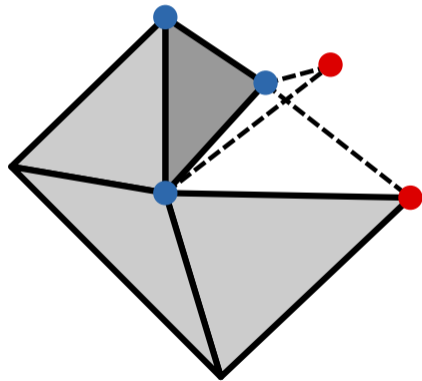
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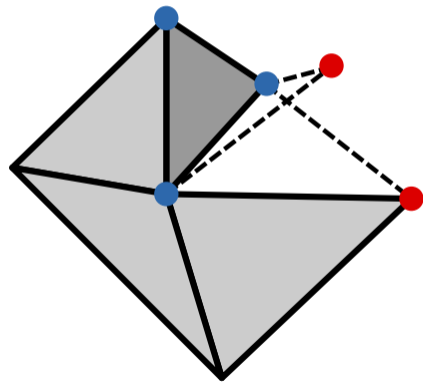
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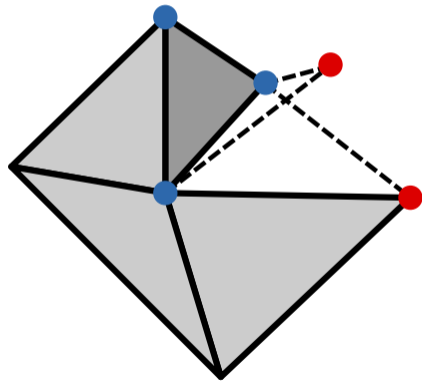
- More feasible situation in future
- Exponential distribution expected
- exp-Golomb code + CABAC

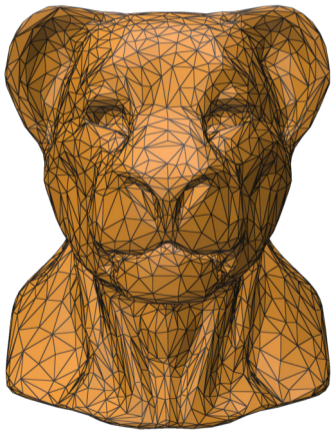


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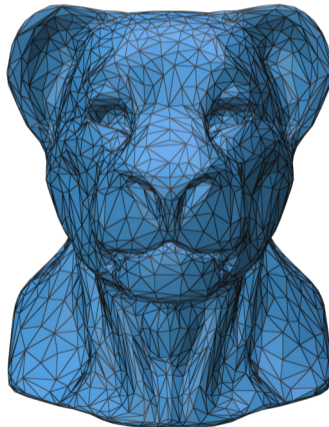
$$p = q_{\max} - q_{\max2}$$

- More feasible situation in future
- Exponential distribution expected
- exp-Golomb code + CABAC
- Benefits when smaller values are encoded first





Distance-ranked

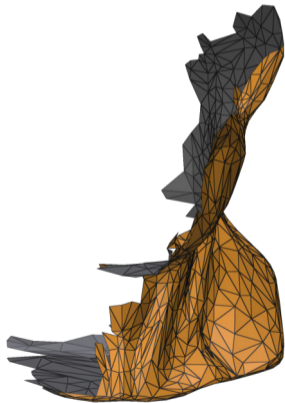


Priority-driven

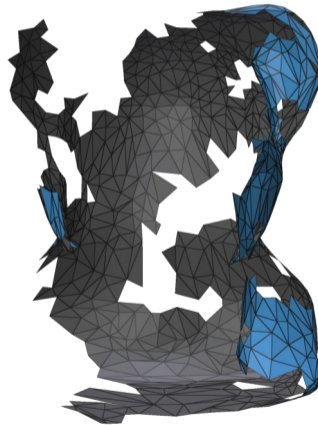
Fixed vs. Priority-driven traversal

Distance-ranked

Priority-driven



Distance-ranked



Priority-driven

Candidate vertex quality

- Measure the feasibility of the potentially formed triangle

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$$q = \theta - w_1 \cdot \bar{d} + w_2 \cdot \phi + w_3 \cdot S$$

$$w_1, w_2, w_3 \in \mathbb{R}_{>0}$$

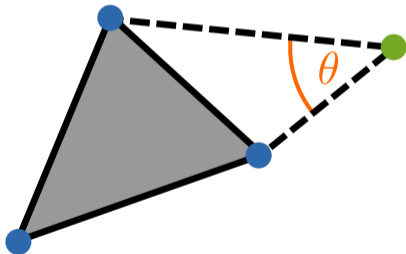
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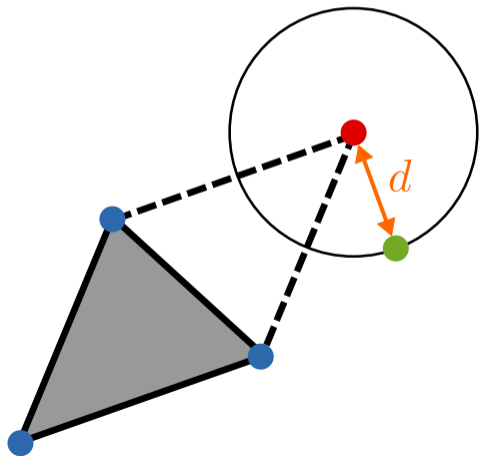
- Optimally - regular planar triangulation

Penalize long triangles



Best value: π

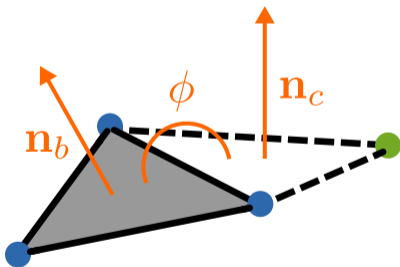
Planarity, similar properties (area, edge length, inner angles)



$$\bar{d} = \frac{d}{l_{\text{avg}}}$$

Best value: 0

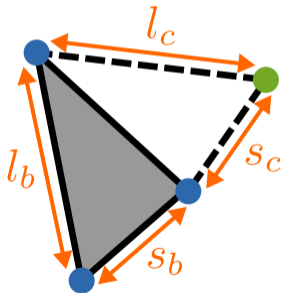
Planarity



$$\phi = \pi - \arccos(\mathbf{n}_b \cdot \mathbf{n}_c)$$

Best value: π

Alternative to parallelogram rule without enforced planarity



$$r_s = s_b/s_v, \quad r_l = l_b/l_c$$

$$r = (r_s + r_l)/2$$

$$S = -(|r - r_s| + |r - r_l|)/2.$$

Best value: 0

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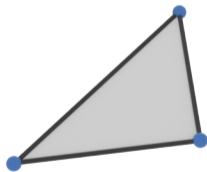
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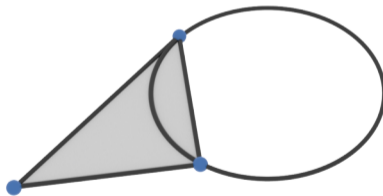
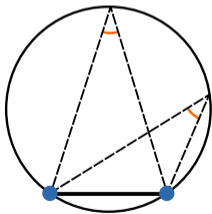
Still can be evaluated by radial search, the search area can be deduced by plugging optimal values into the equation for evaluating q .

$$q = \theta - w_1 \cdot \bar{d} + w_2 \cdot \phi + w_3 \cdot S$$



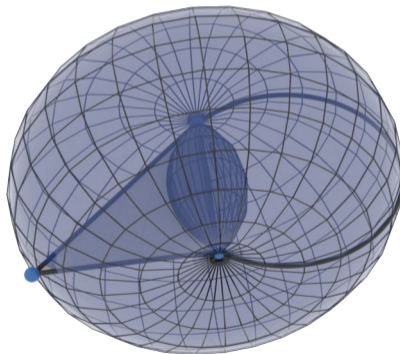
$$q_c = \theta_{\min} - w_1 \cdot 0 + w_2 \cdot \pi + w_3 \cdot 0$$

$$\theta_{\min} = q_c - w_2 \cdot \pi$$



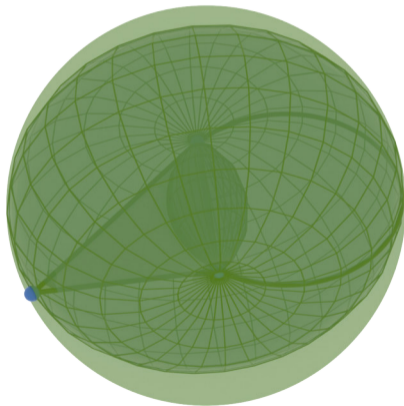
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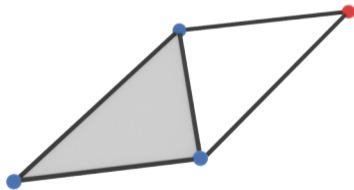


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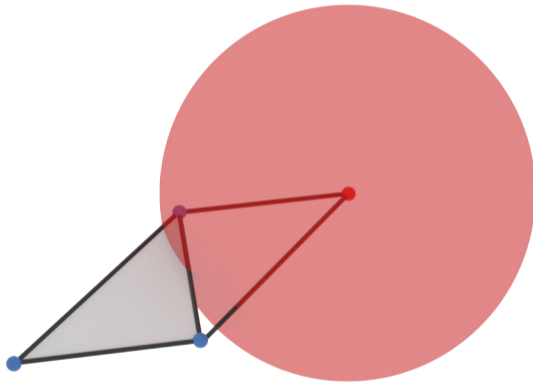


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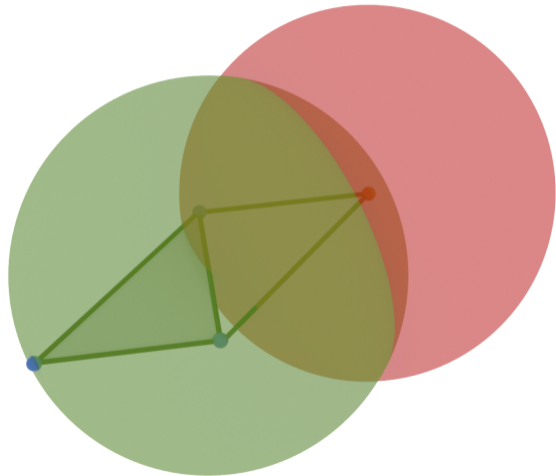


$$q_c = \pi - w_1 \cdot \bar{d}_{\max} + w_2 \cdot \pi + w_3 \cdot 0$$

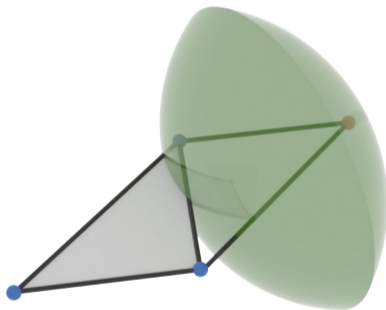
$$d_{\max} = \frac{(w_2 \cdot \pi + \pi - q_c) \cdot l_{\text{avg}}}{w_1}$$



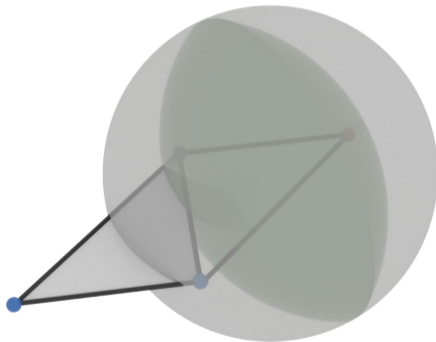
All vertices with $q \geq q_c$ lie
within $\mathcal{B}_d \cap \mathcal{B}_\theta$



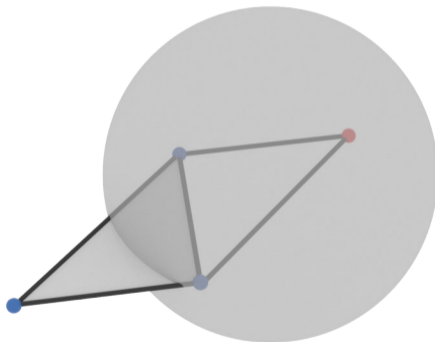
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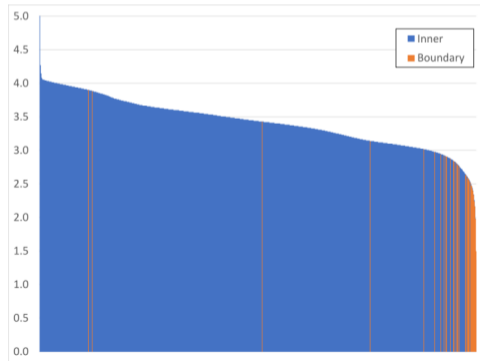
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Boundary edge prediction

Assumption

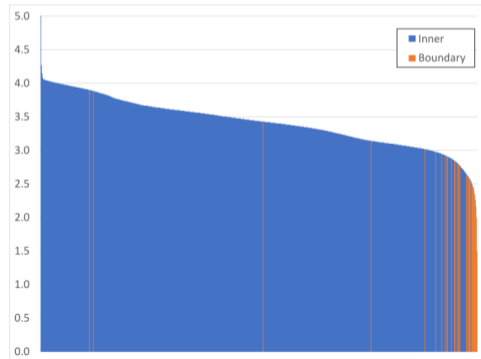
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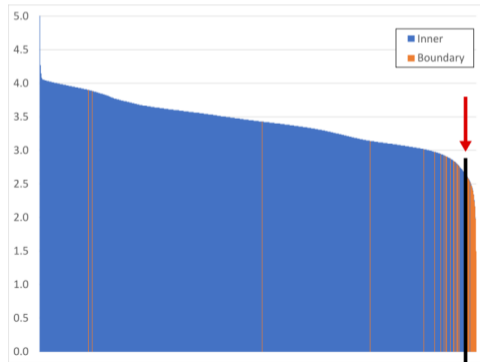
- Evaluate q_{\max} of each edge



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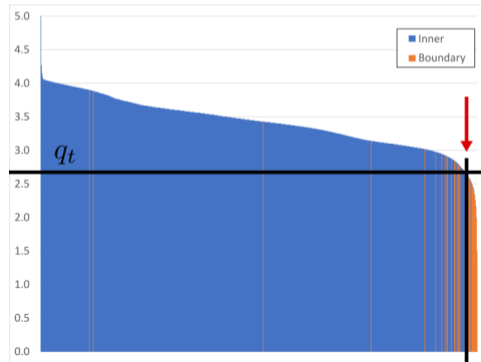
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- Find q_t which best separates q_{\max} of inner/boundary edges



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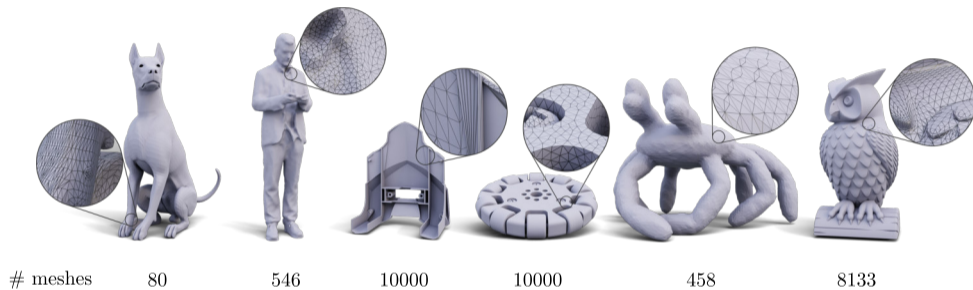
The boundary edge usually has no candidate vertex of high quality.

- Evaluate q_{\max} of each edge
- Find q_t which best separates q_{\max} of inner/boundary edges
- Predict boundary if $q_{\max} < q_t$

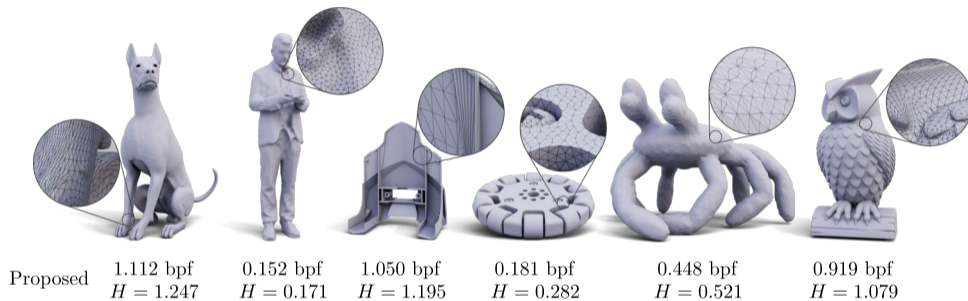
Prediction	Actual	Symbol
Inner	Inner	i
Boundary	Boundary	0
Boundary	Inner	$i + 1$
Inner	Boundary	$\max(i) + 1$

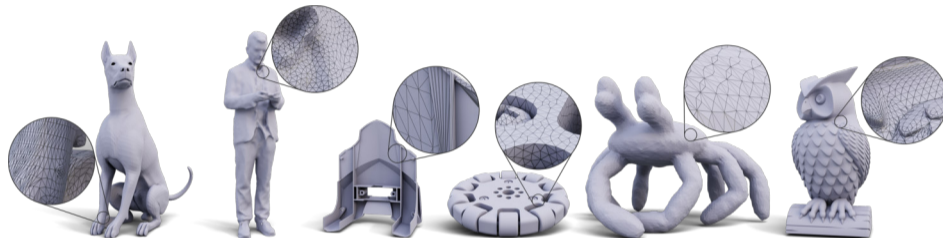
Experimental results

Main experiment

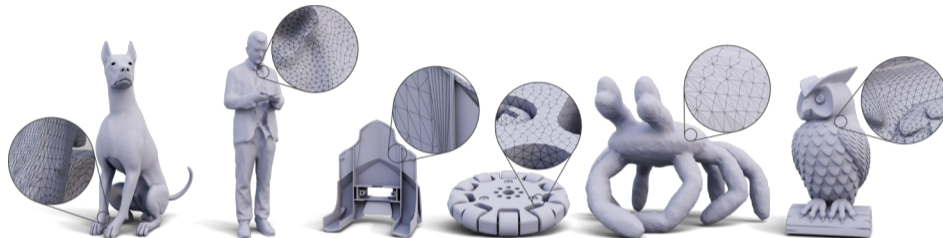


Main experiment





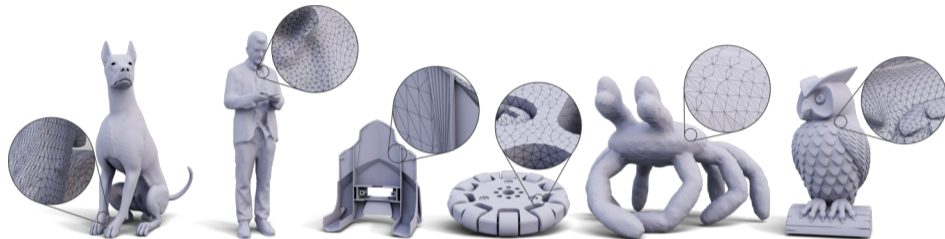
Proposed	1.112 bpf $H = 1.247$	0.152 bpf $H = 0.171$	1.050 bpf $H = 1.195$	0.181 bpf $H = 0.282$	0.448 bpf $H = 0.521$	0.919 bpf $H = 1.079$
Distance-ranked	1.343 bpf $H = 1.317$	0.264 bpf $H = 0.232$	2.261 bpf $H = 2.324$	0.331 bpf $H = 0.381$	0.708 bpf $H = 0.693$	1.523 bpf $H = 1.552$



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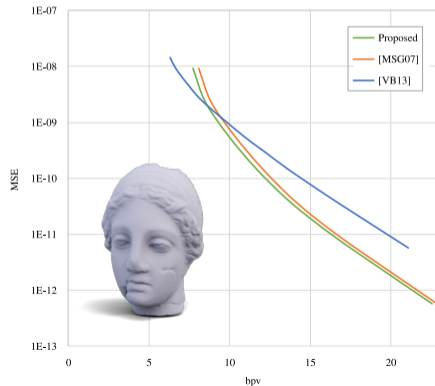
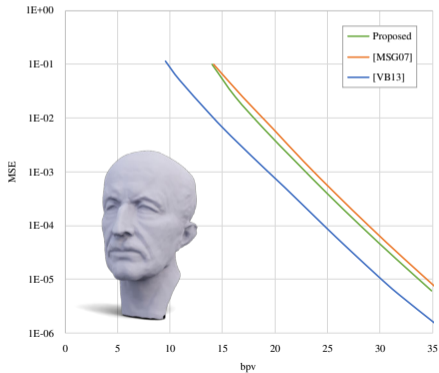
The Distance-ranked algorithm does not achieve a consistently lower data rate than H .

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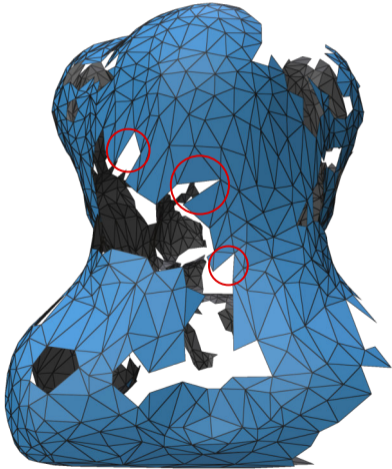


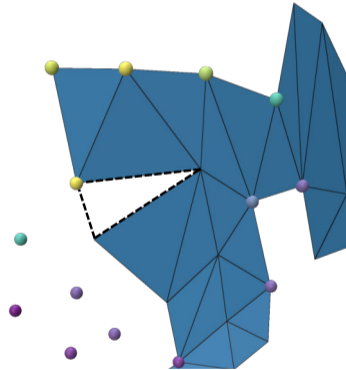
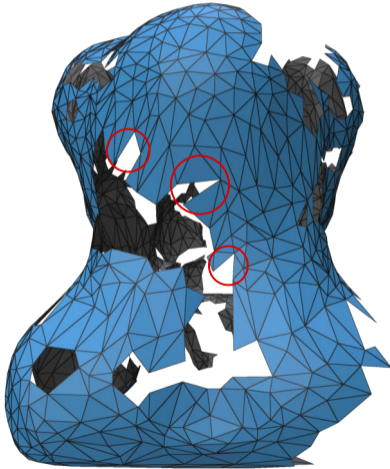
Improvement	17.17%	42.62%	53.54%	45.26%	36.68%	39.66%
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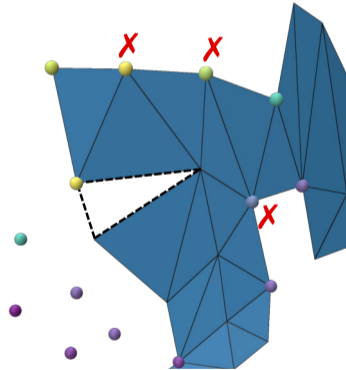
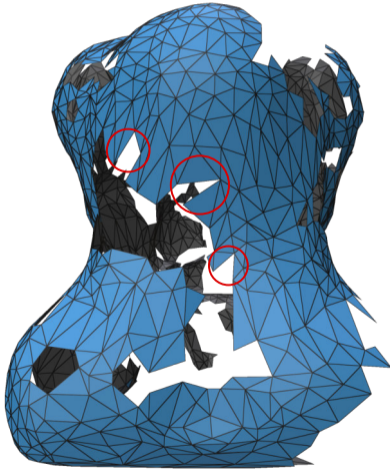
Combined with PC codec [Merry et al. 2006] vs. Weighted parallelogram [Váša-Brunnett 2013]

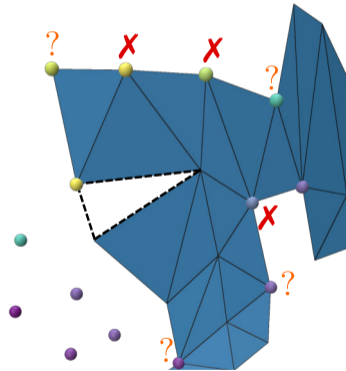
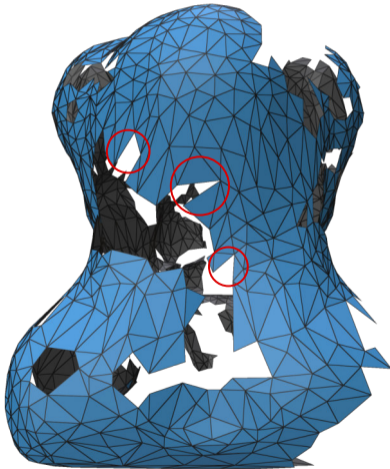


Limitations & Future Work





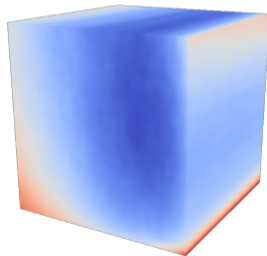




Difficult to find optimal w_1, w_2, w_3 for a certain model.

- Global trend towards a region of satisfactory rates

$bpf(w_1, w_2, w_3)$



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In experiments:

- Default parameters vs. Fine-tuned for each dataset
- Estimated on a subset
- Default parameters still better than Distance-ranked
- No significant improvement for irregular data

$$bpf(w_1, w_2, w_3)$$



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 - Vertex degrees
 - Inner angles
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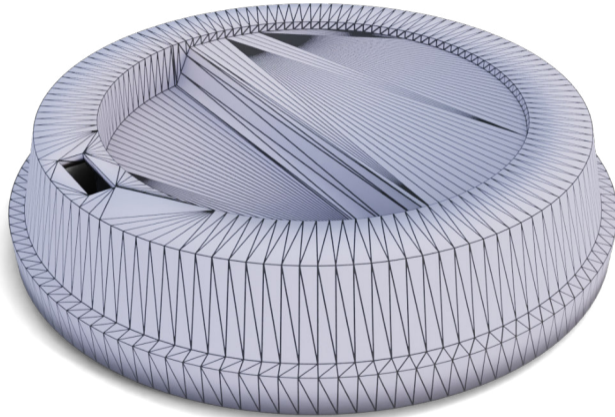
So complex, there might not be any.

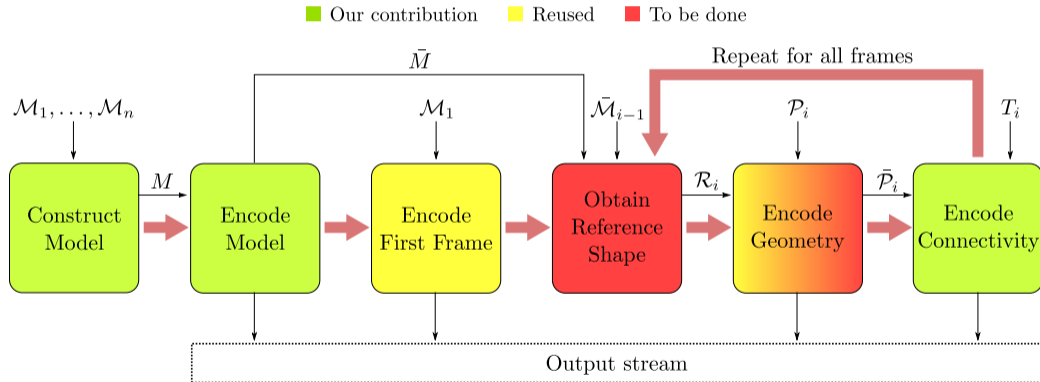
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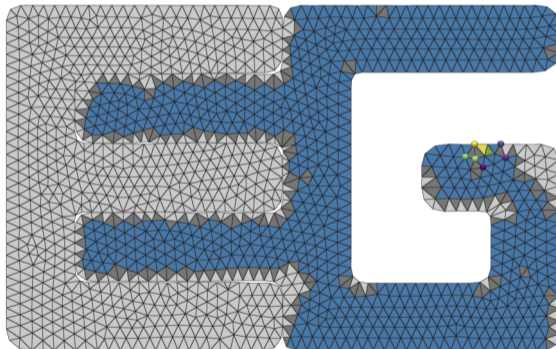
Does not consider all the aspects (e.g., priority).





Thank you

<https://gitlab.kiv.zcu.cz/jdvorak/priority-based-connectivity-coding>



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