

Assisted Object Placement

Master's Thesis – Andreas Kirsch

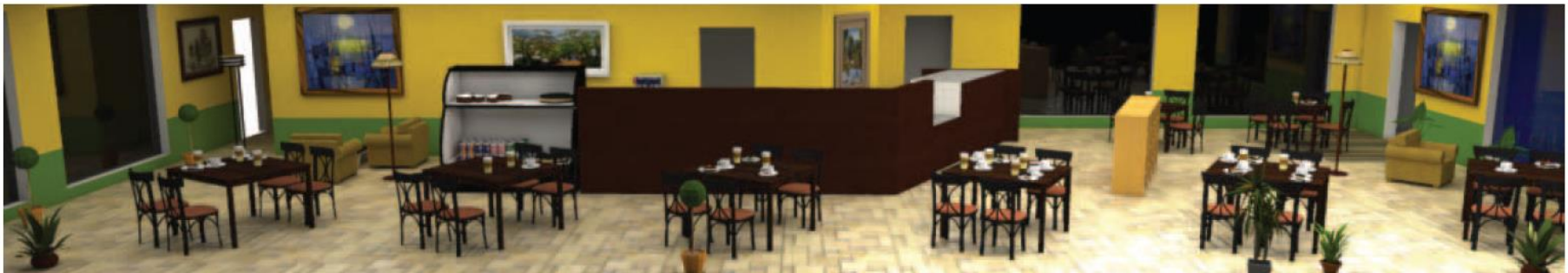


Motivation

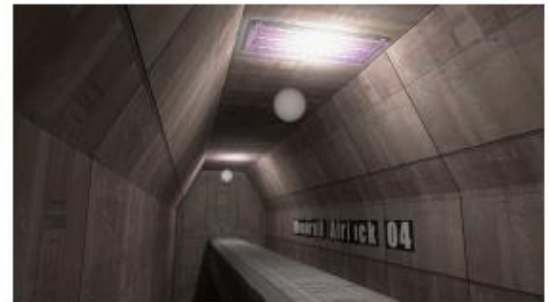
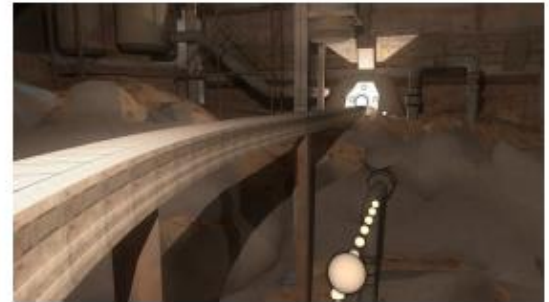
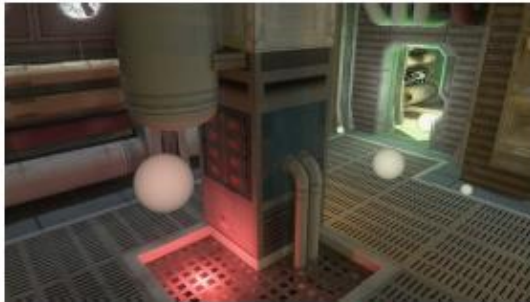
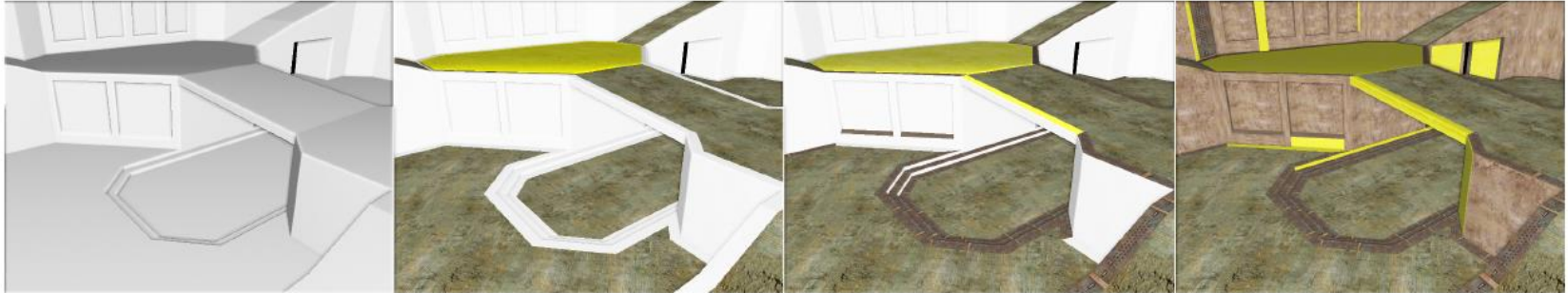


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Background



Background



Shadowgrounds Survivor



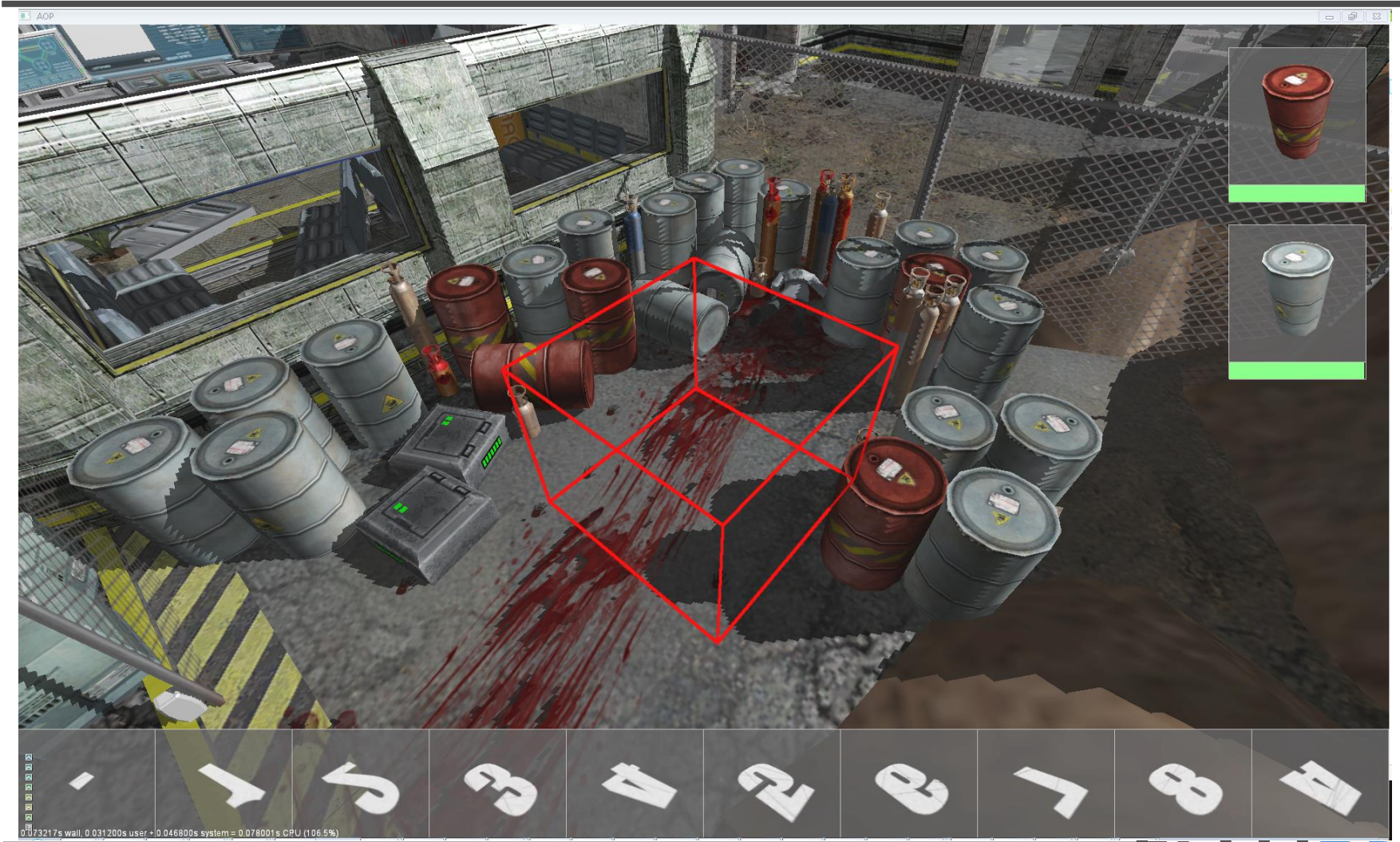
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Goals

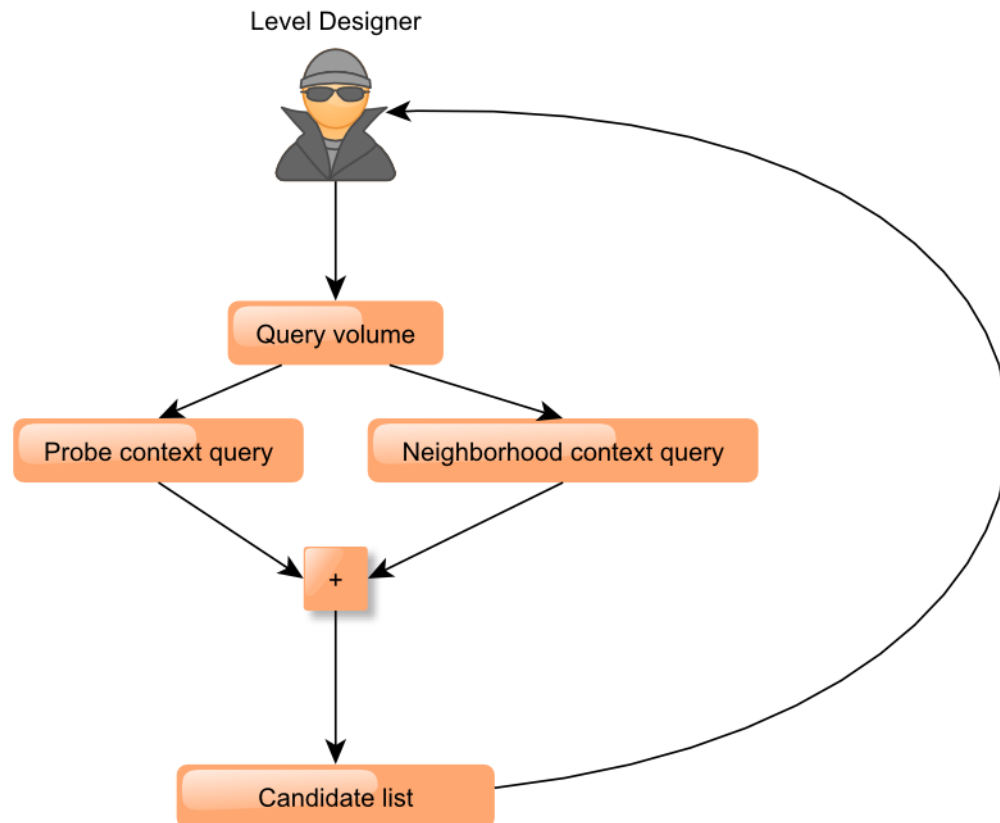


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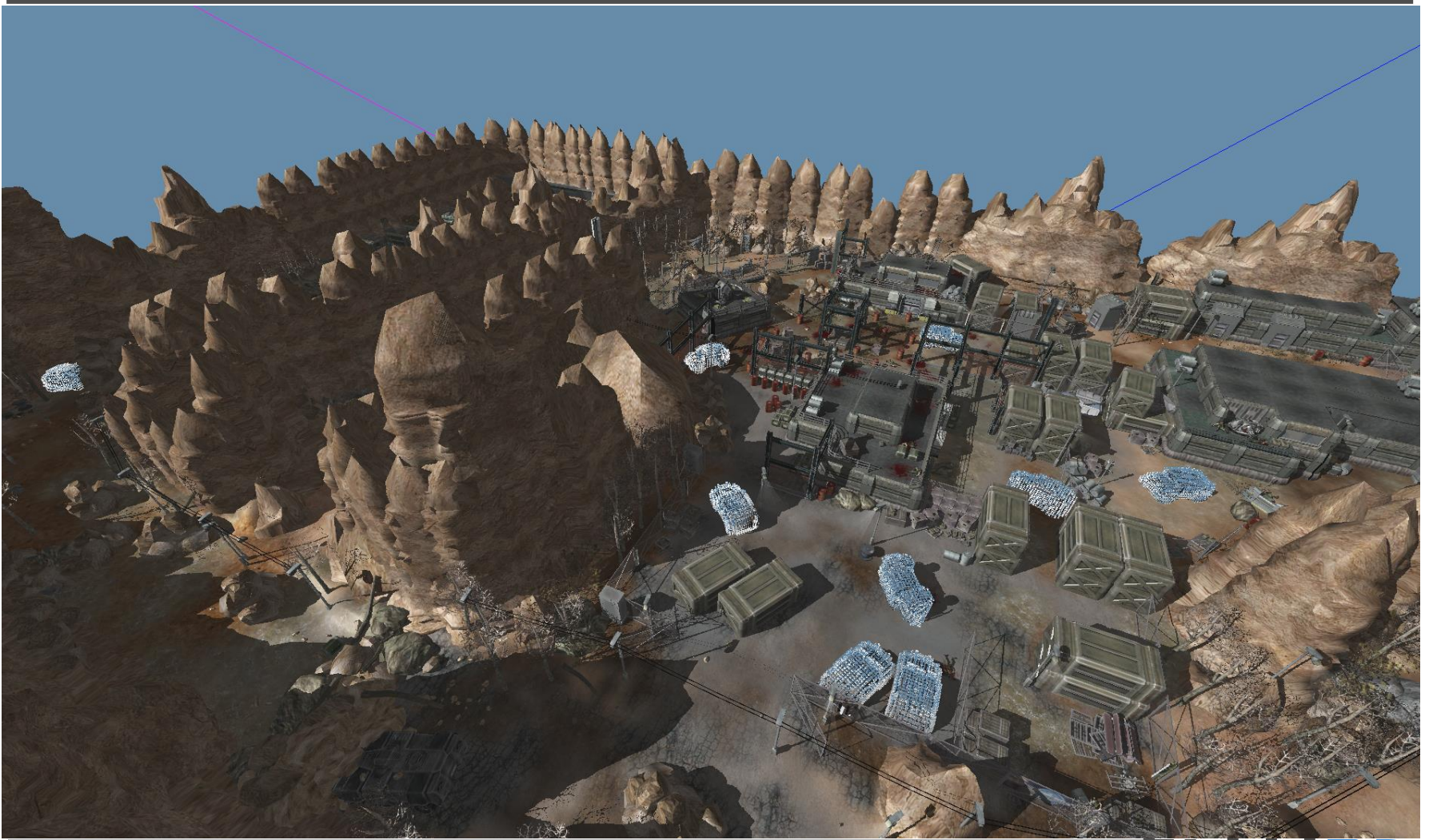
Goals



Overview

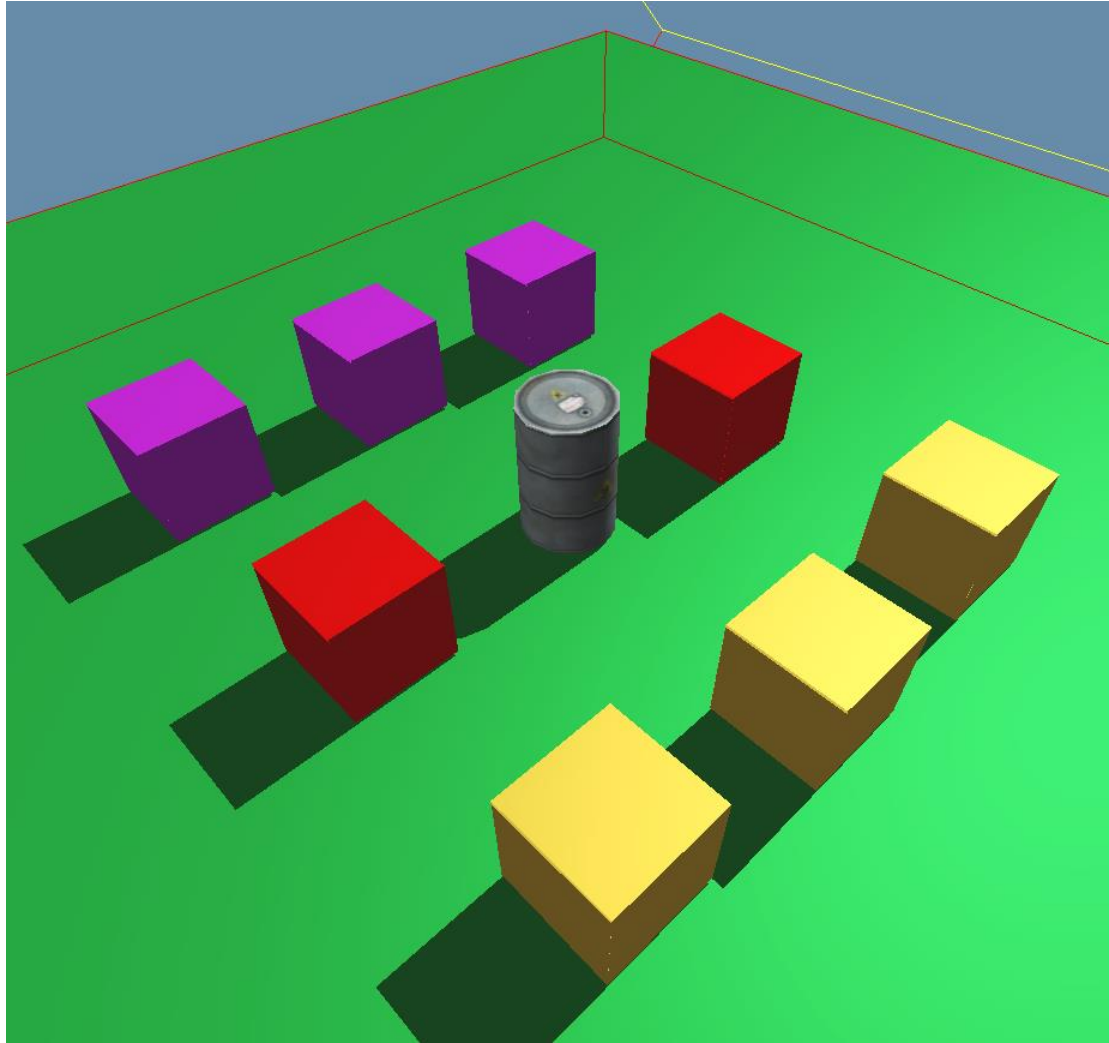


Probe context

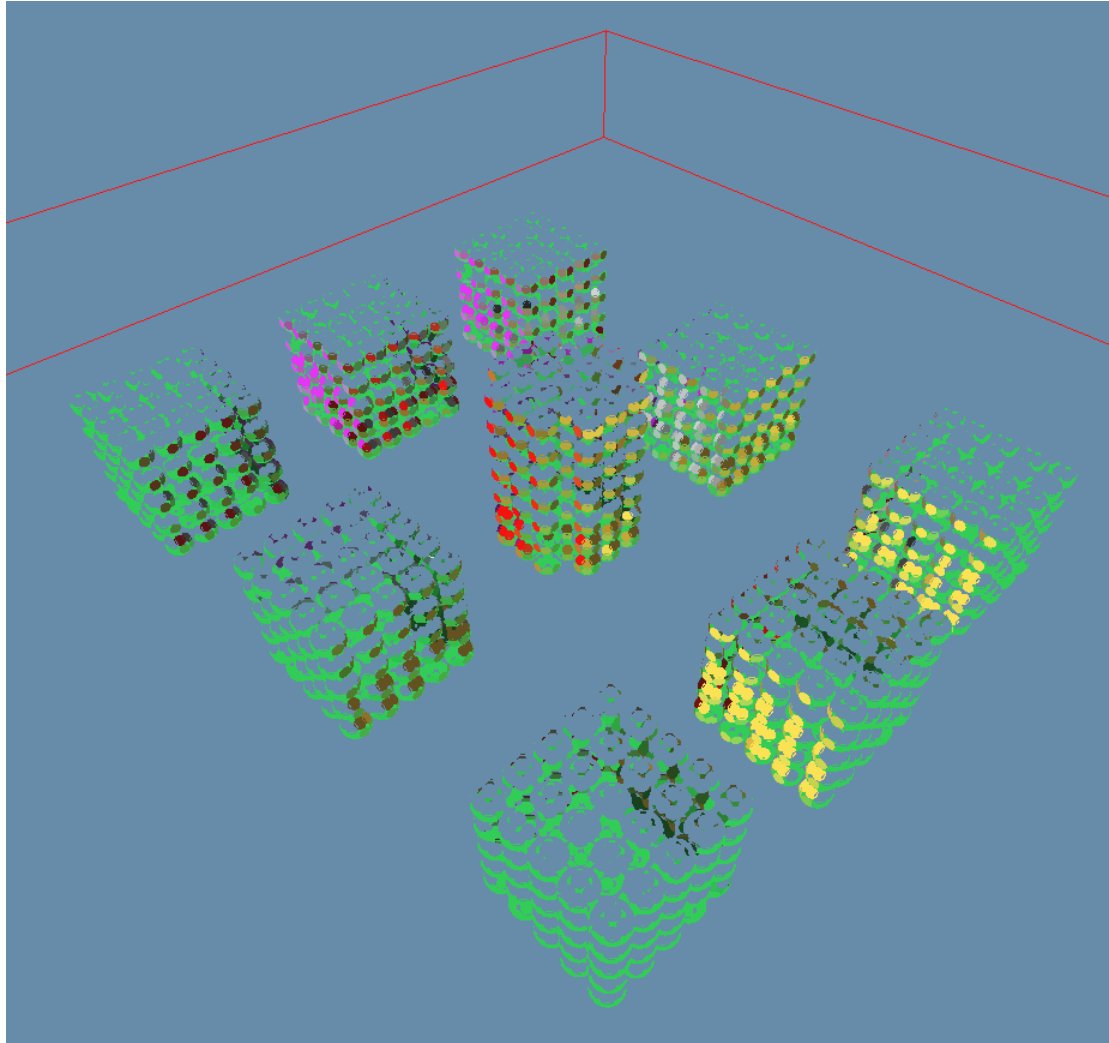


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Probe samples – Example



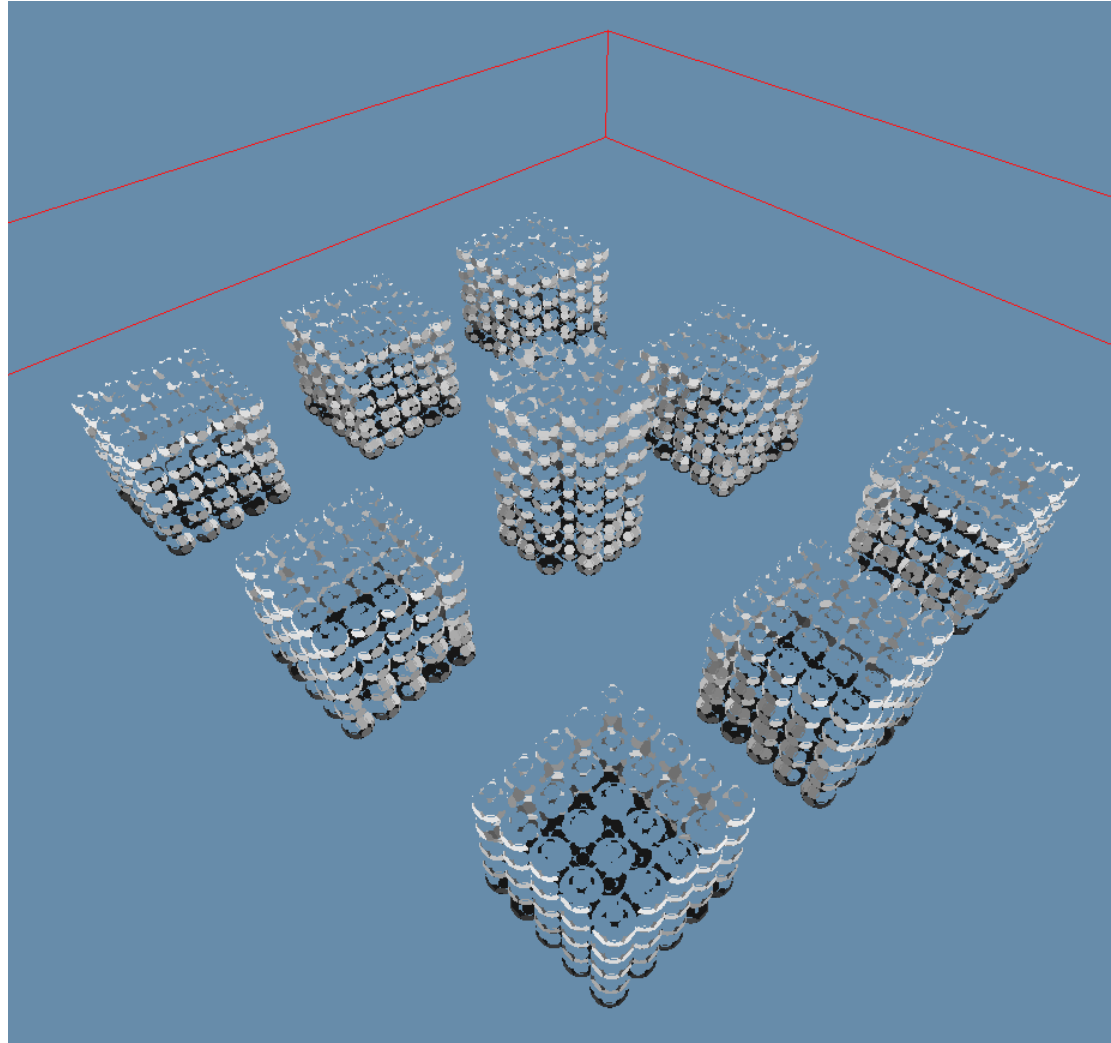
Probe samples – Color



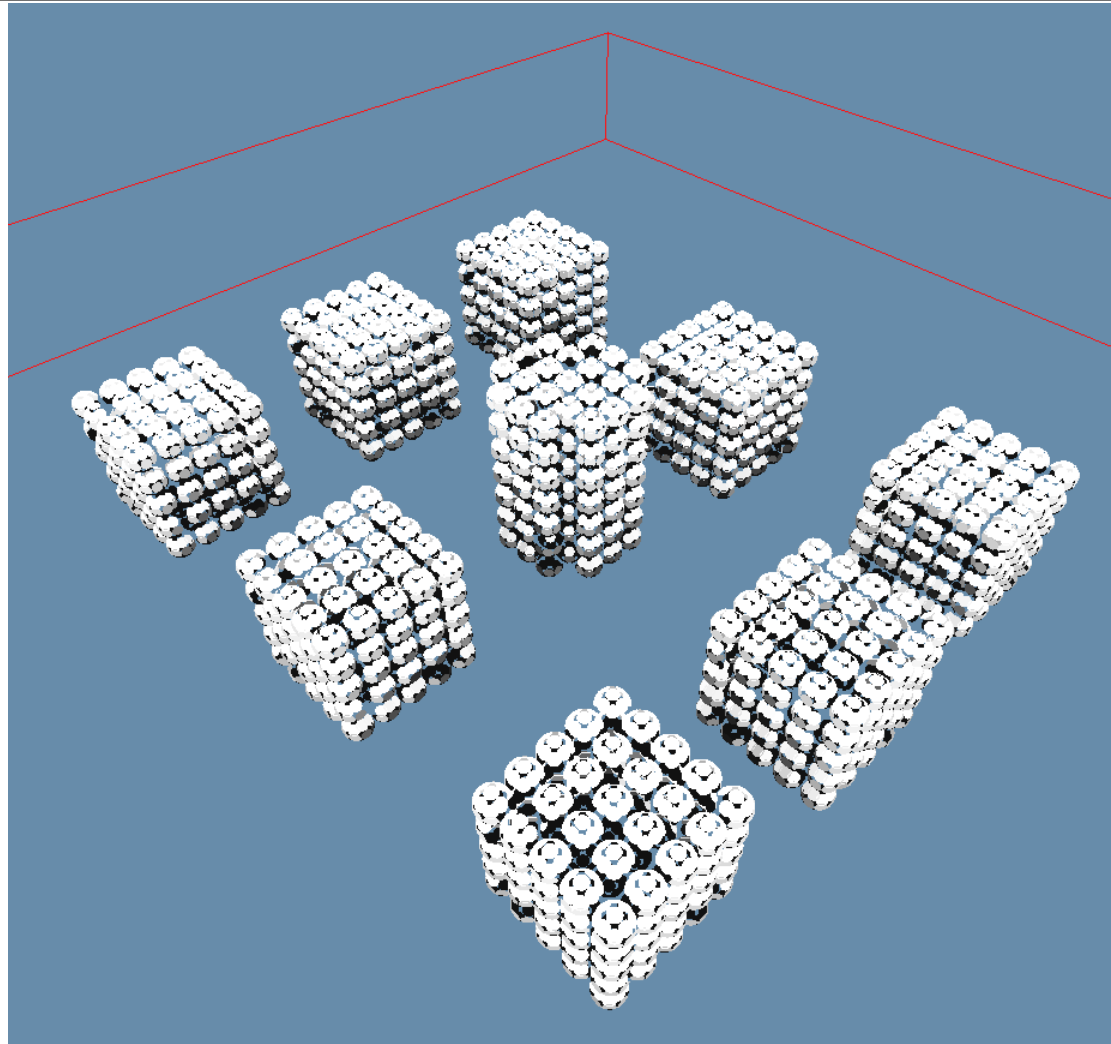
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Probe samples – Distance



Probe samples – Occlusion



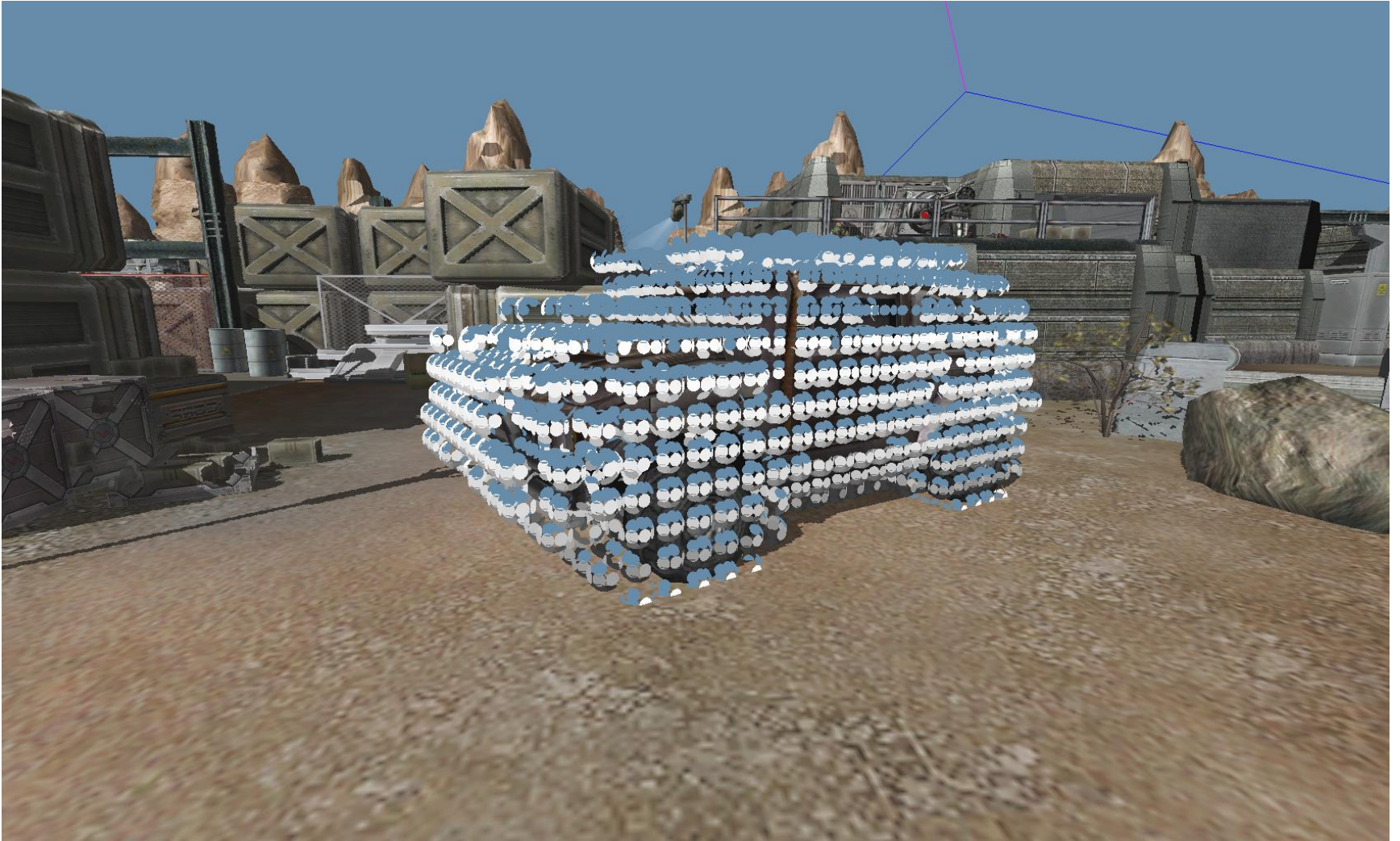
Probe placement



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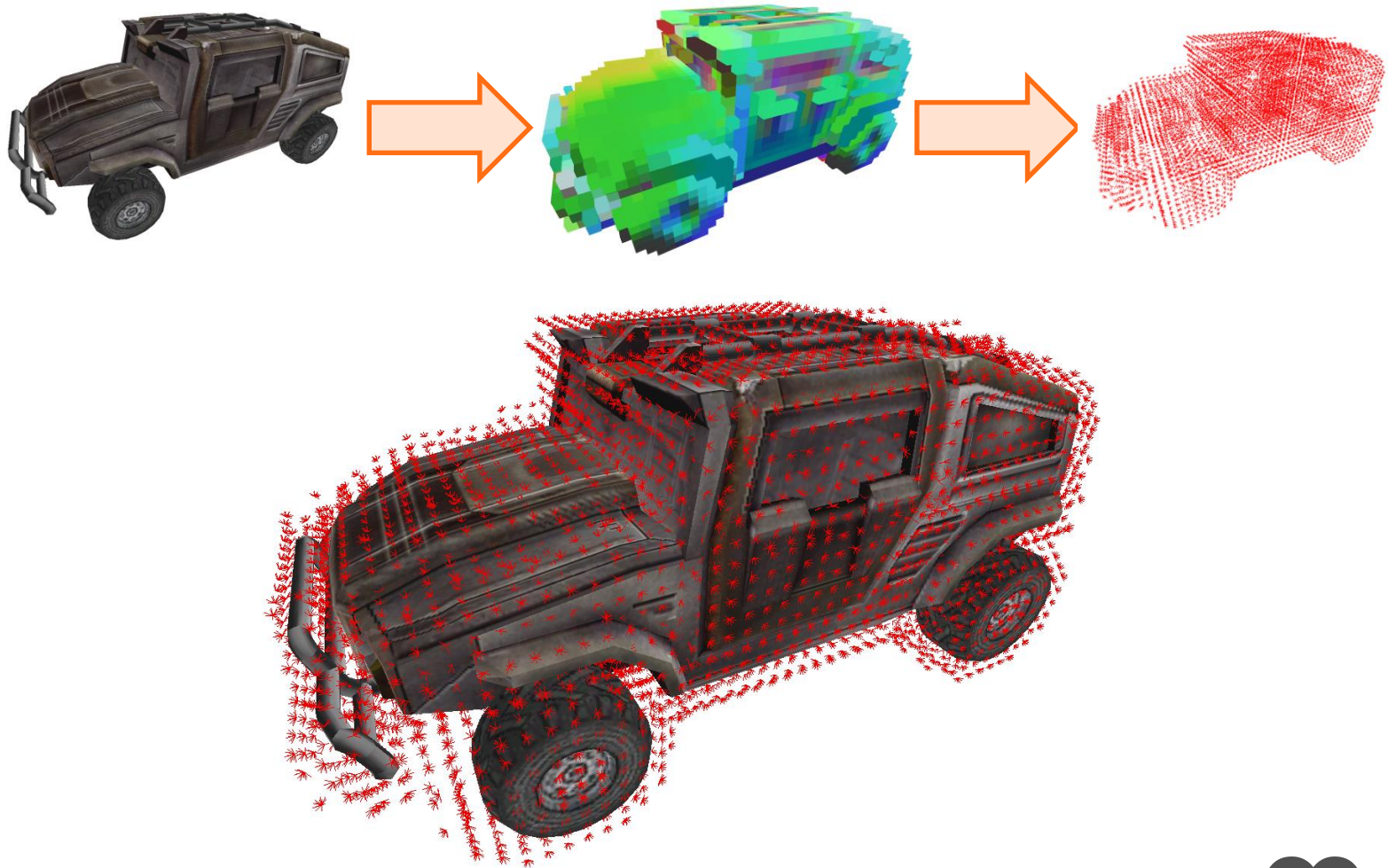
Probe placement



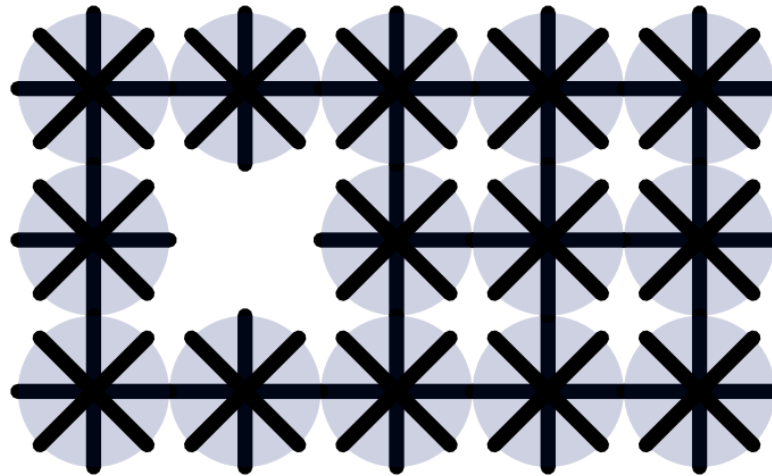
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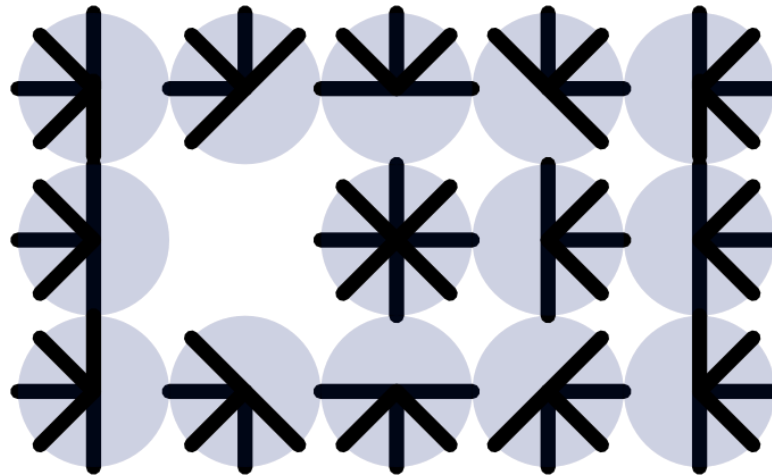
Probe placement



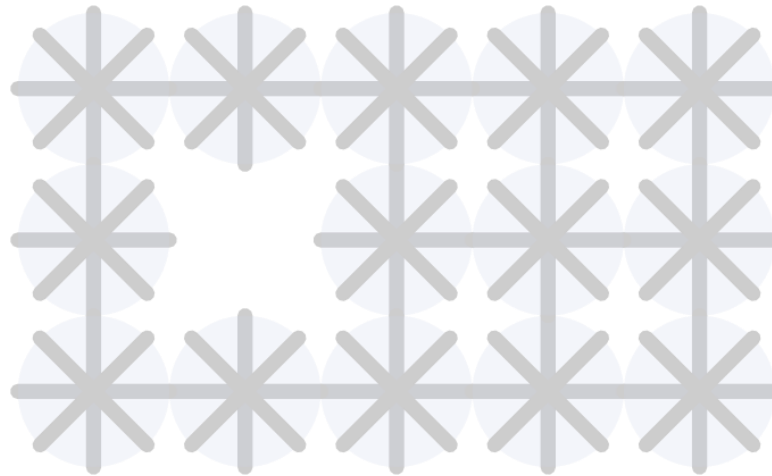
Probe placement – All directions



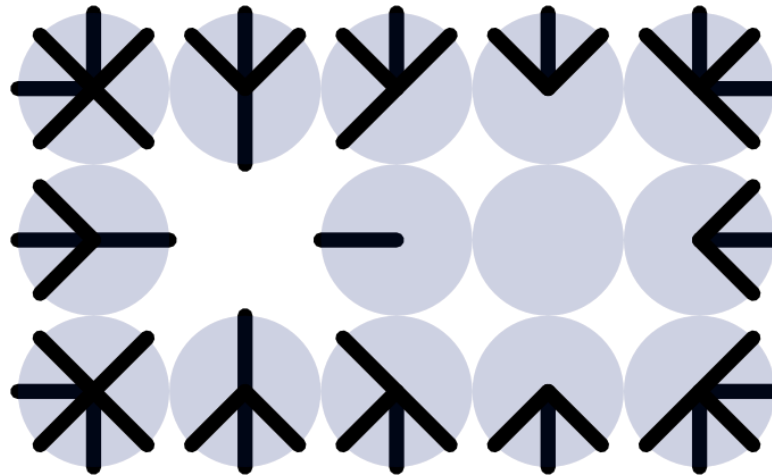
Probe placement – Relative position



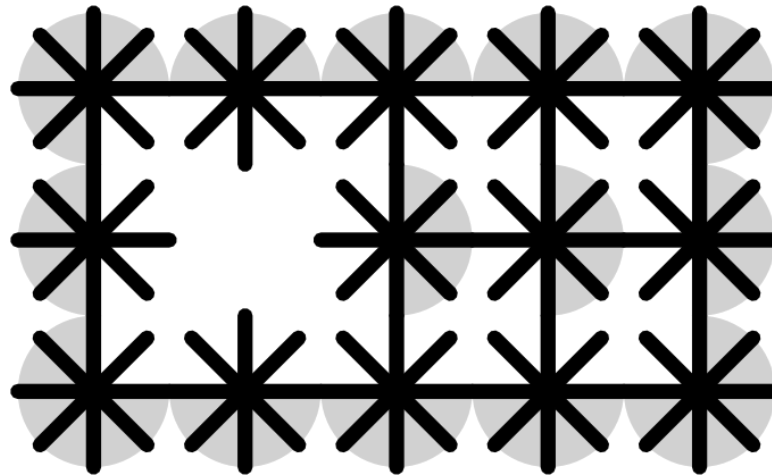
Probe placement – Neighbors



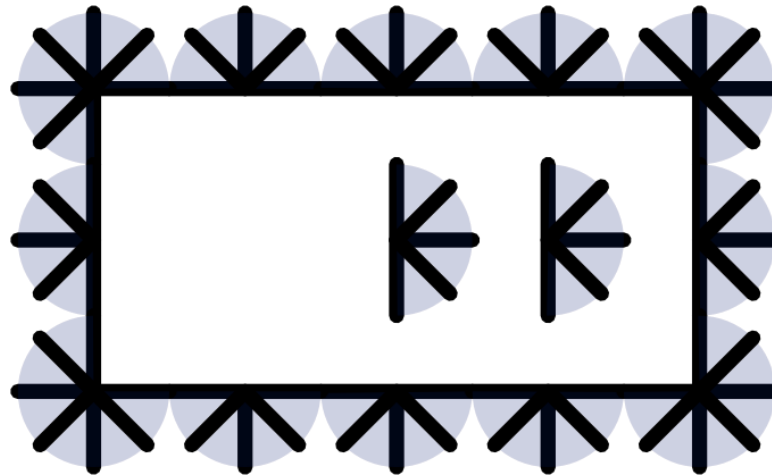
Probe placement – Neighbors



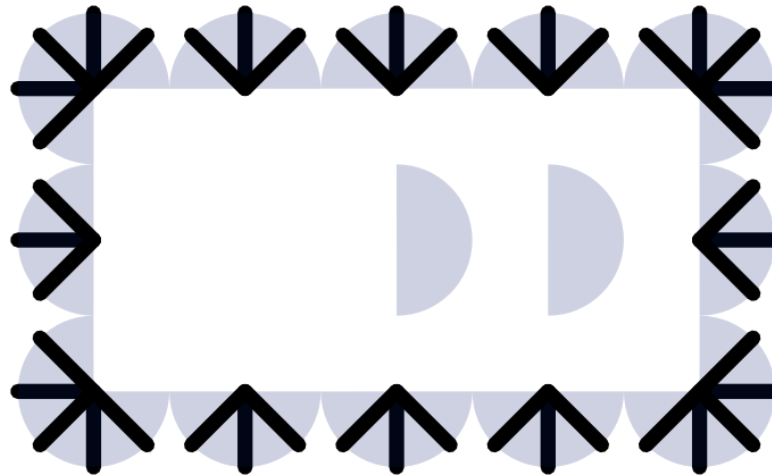
Probe placement – Avg normals



Probe placement – Avg normals



Probe placement – Combined

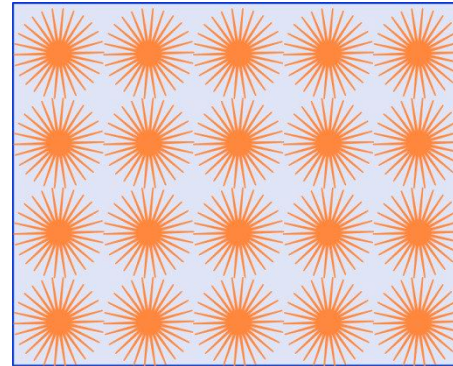
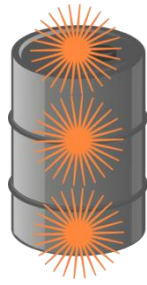


Queries

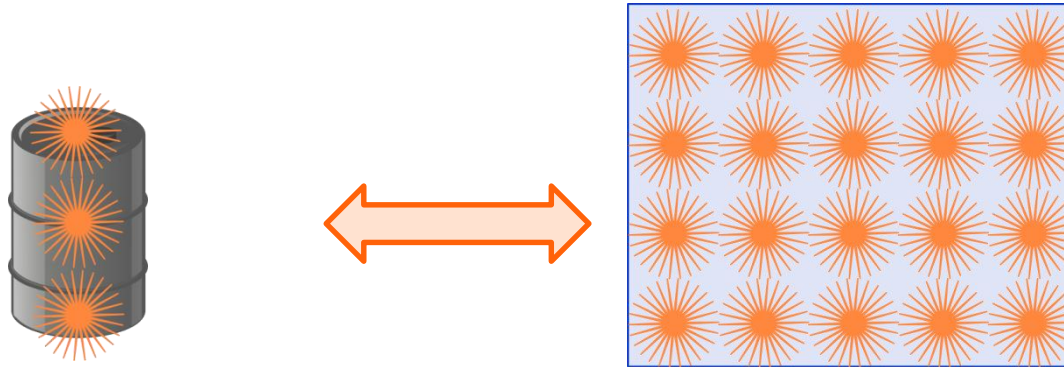


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Bidirectional match query

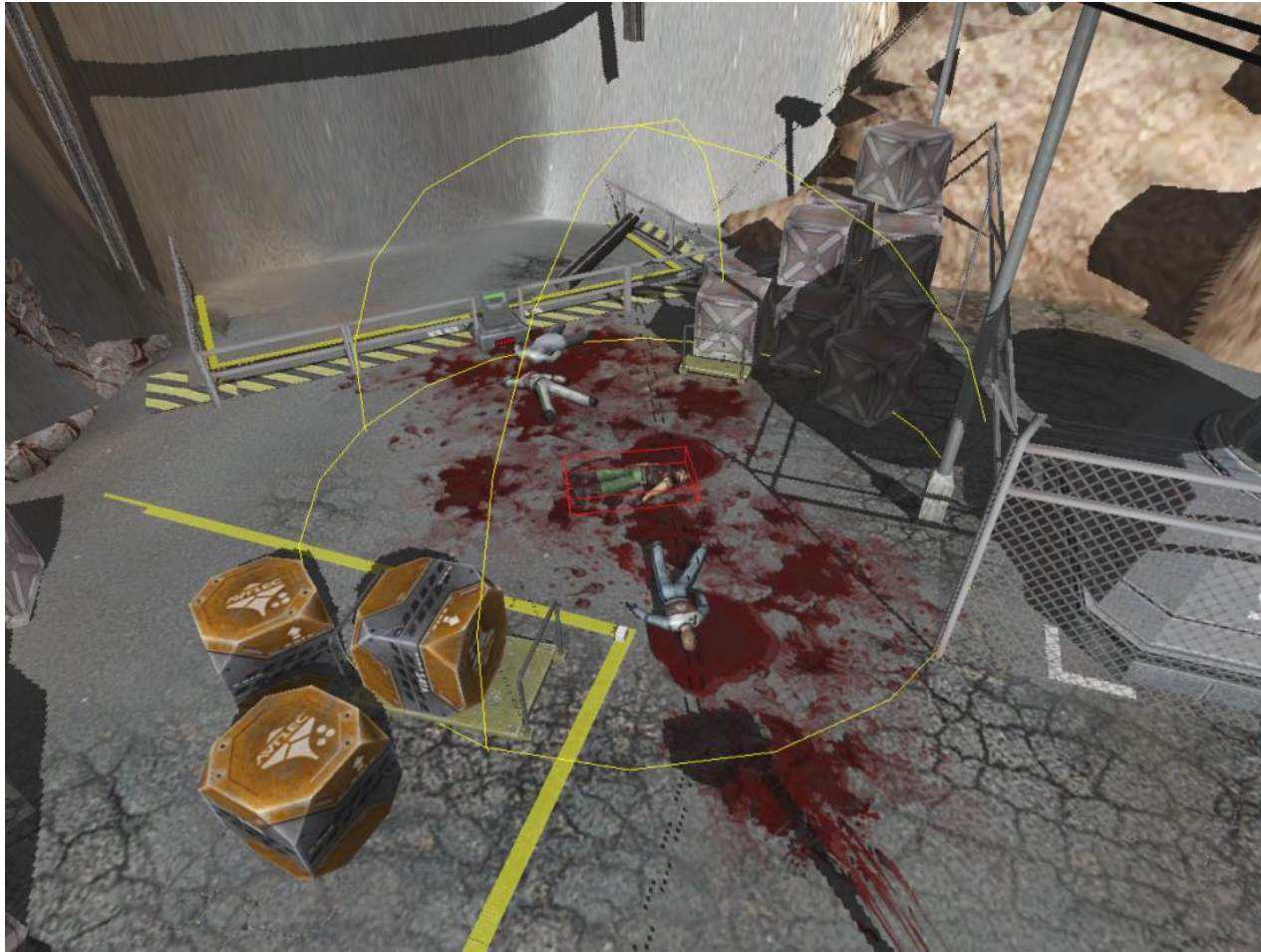


Bidirectional match query

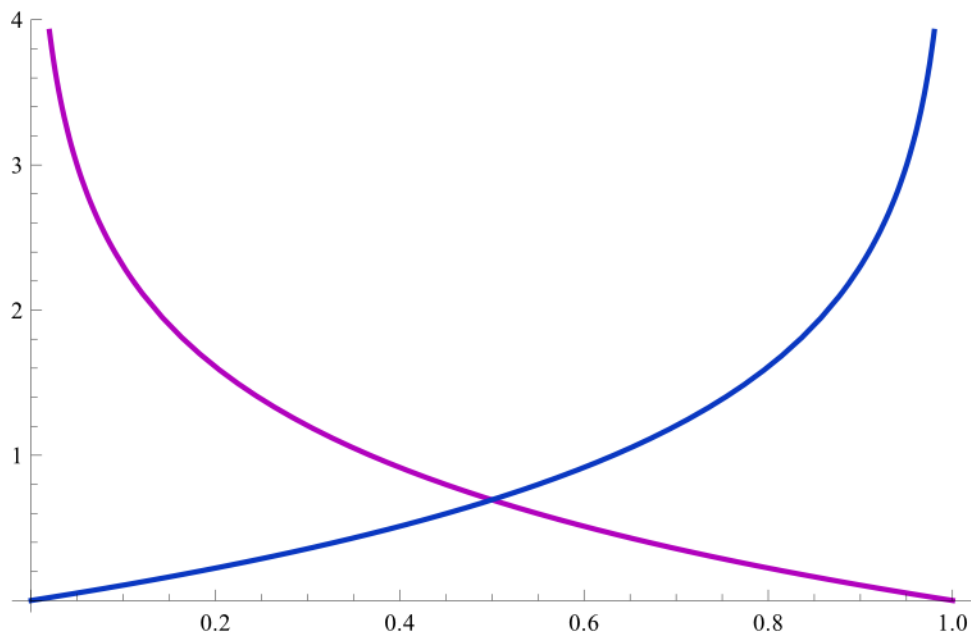


$$S = \frac{\# \text{matches model}}{\# \text{samples model}} \cdot \frac{\# \text{matches query}}{\# \text{samples query}}$$

Importance-weighted queries

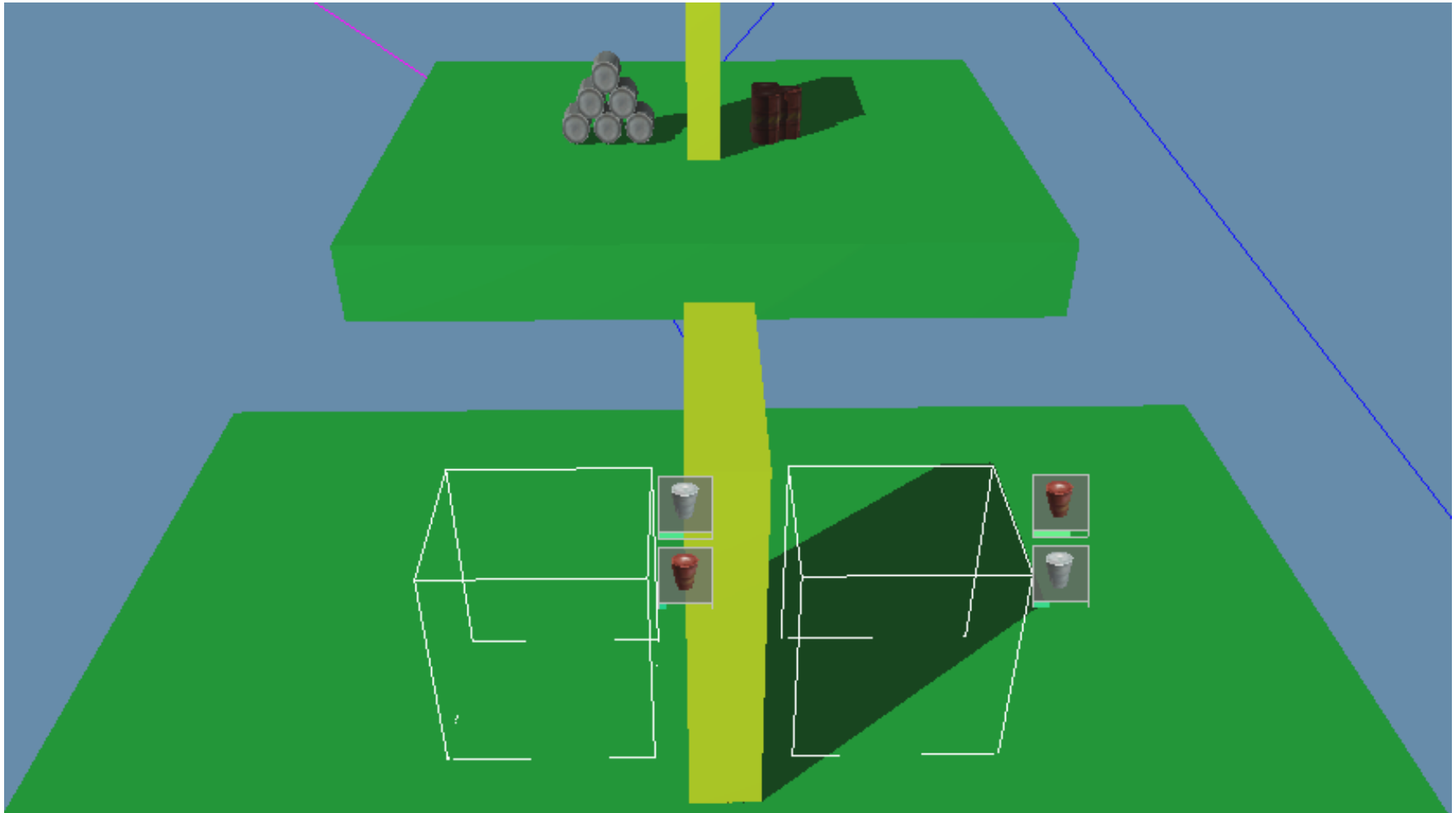


Importance-weighted queries

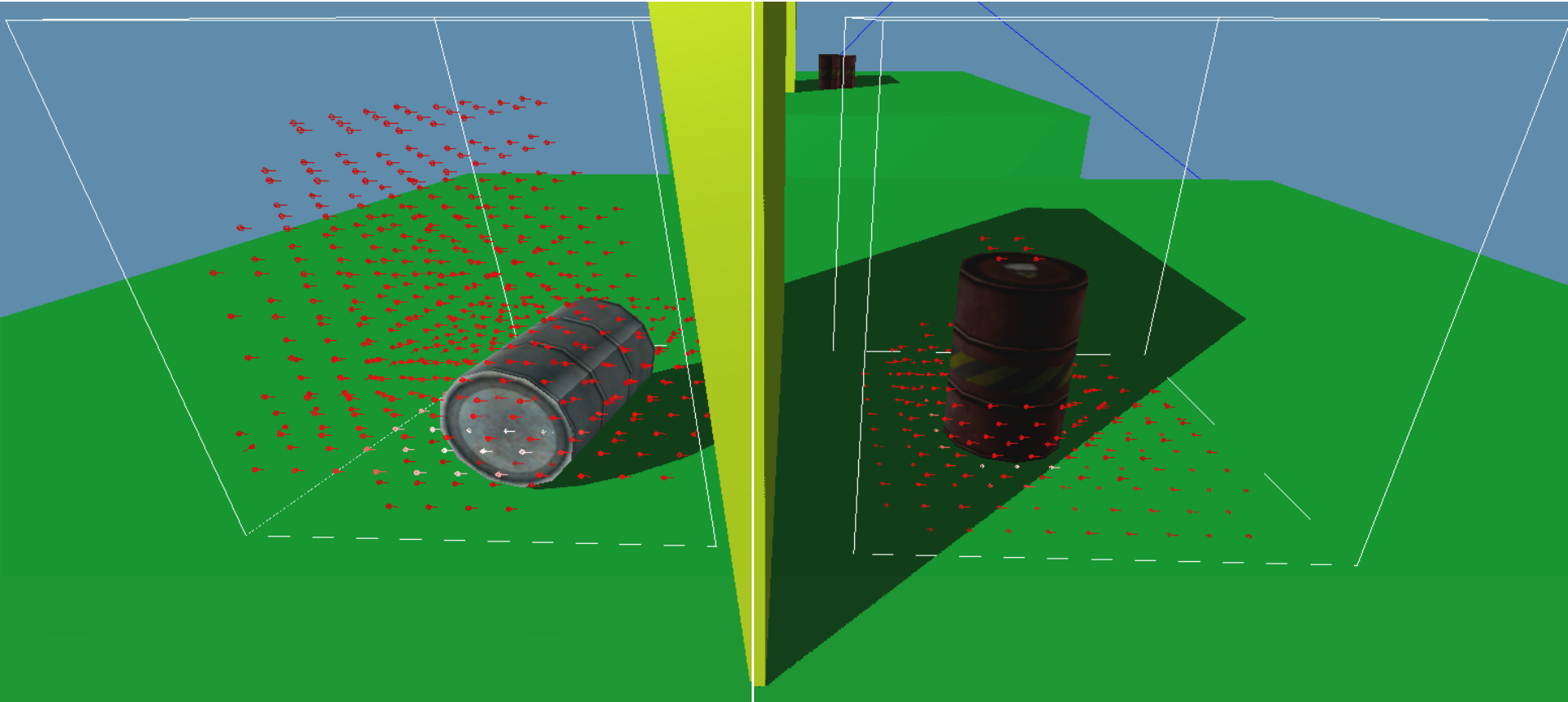


$$h(X) = -\ln P(X)$$

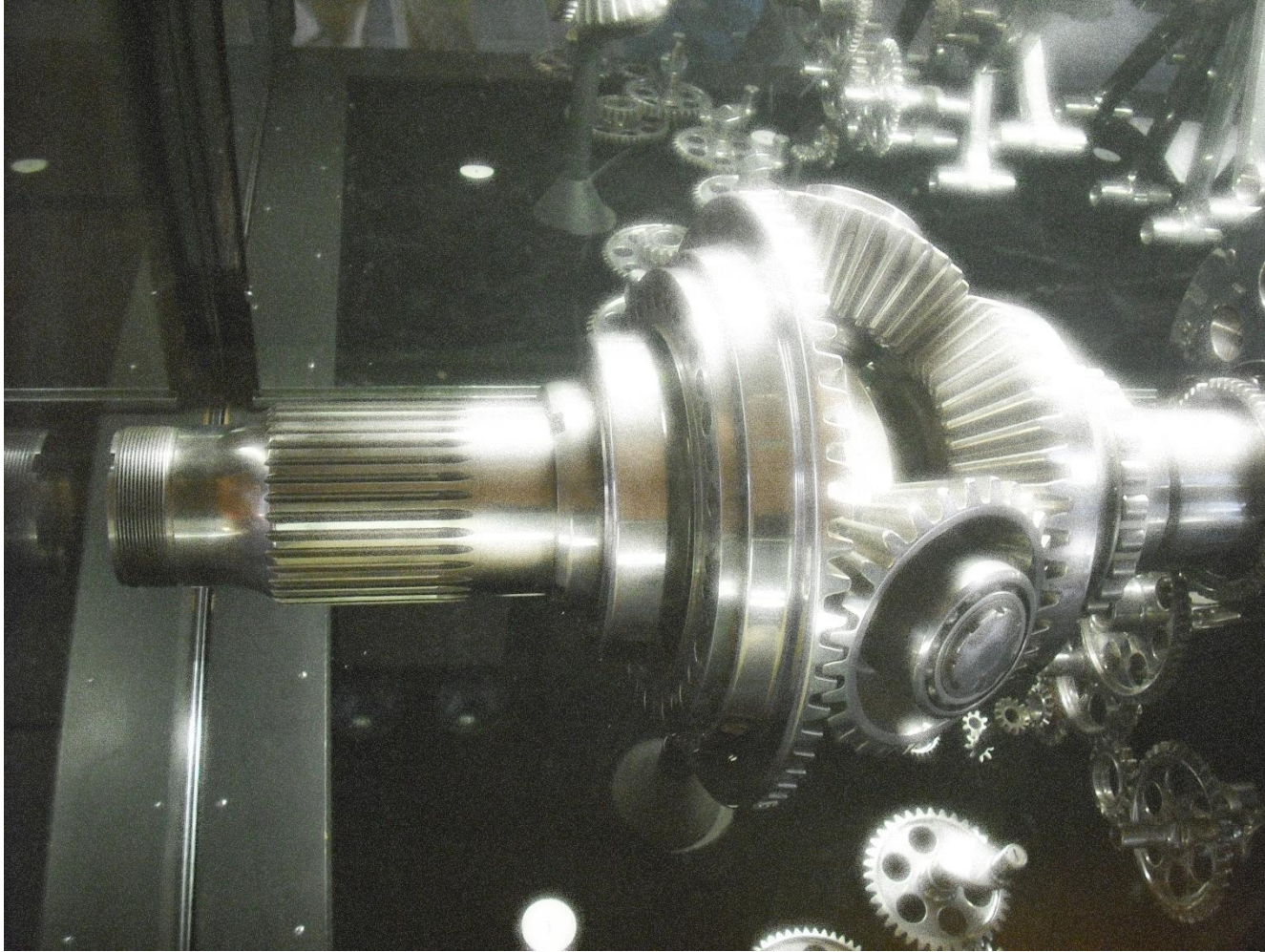
Configuration query



Configuration query



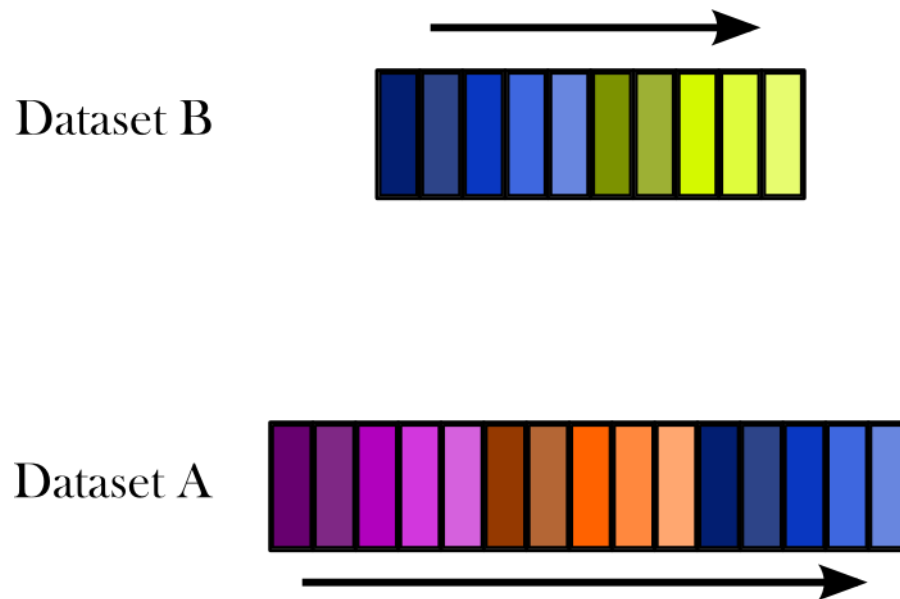
Algorithms



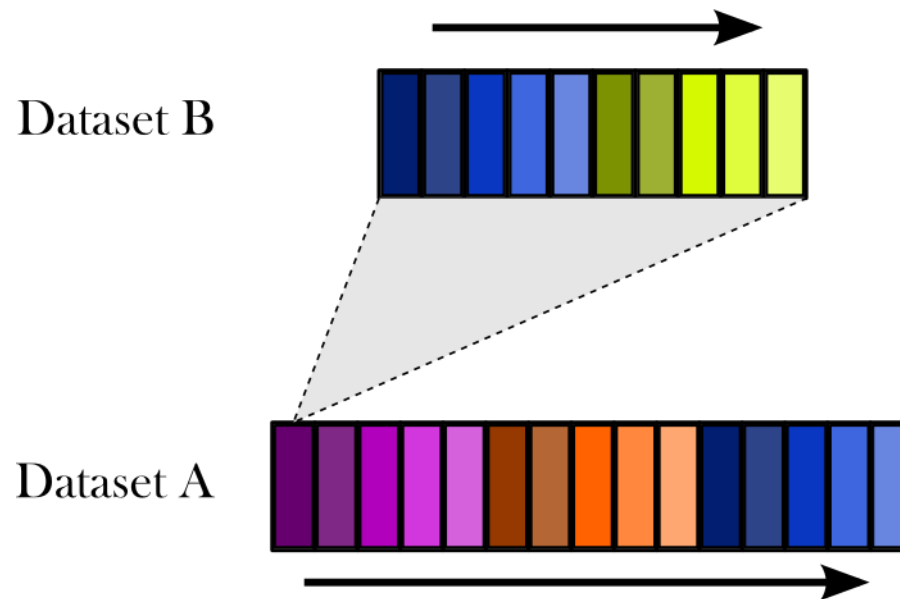
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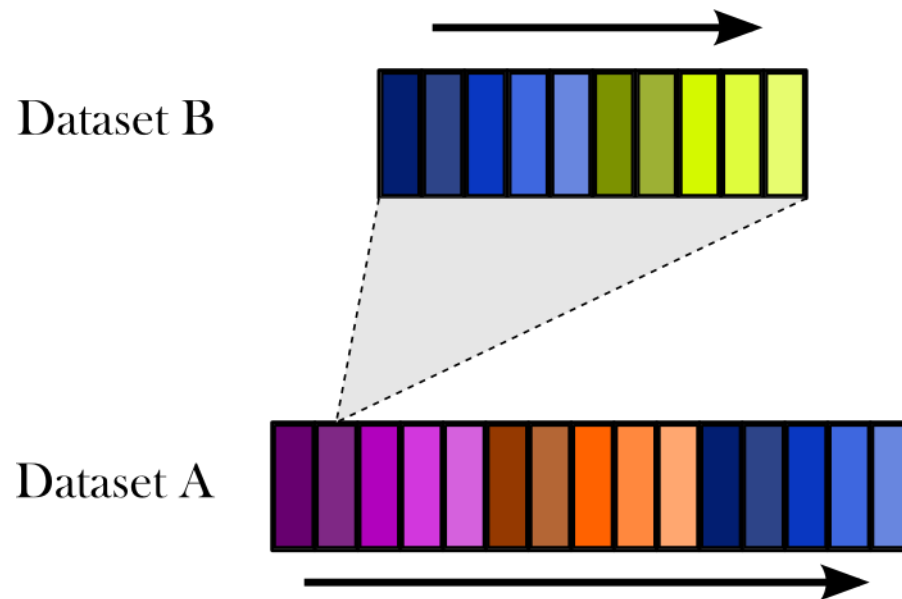
Naive implementation



Naive implementation



Naive implementation



Optimizations

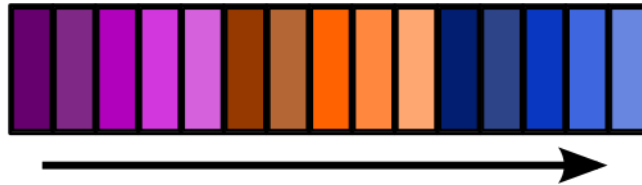


First optimization

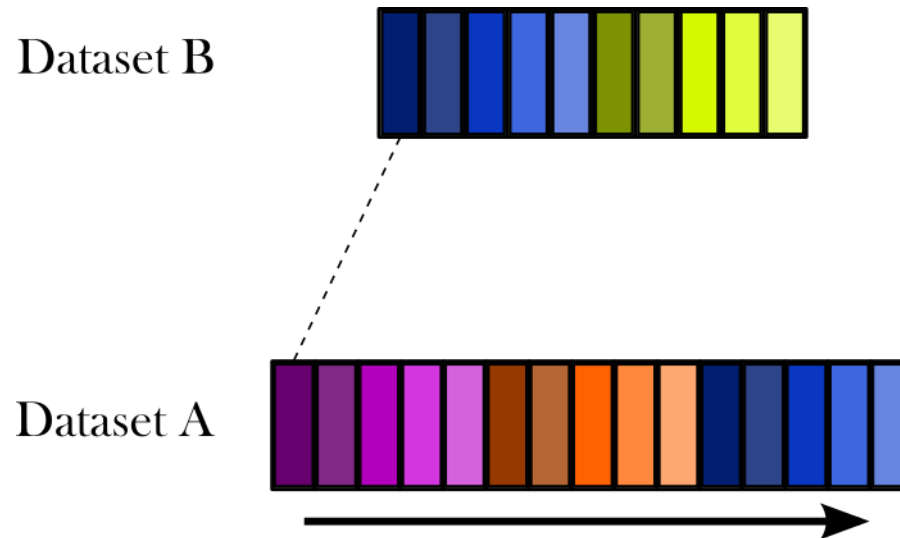
Dataset B



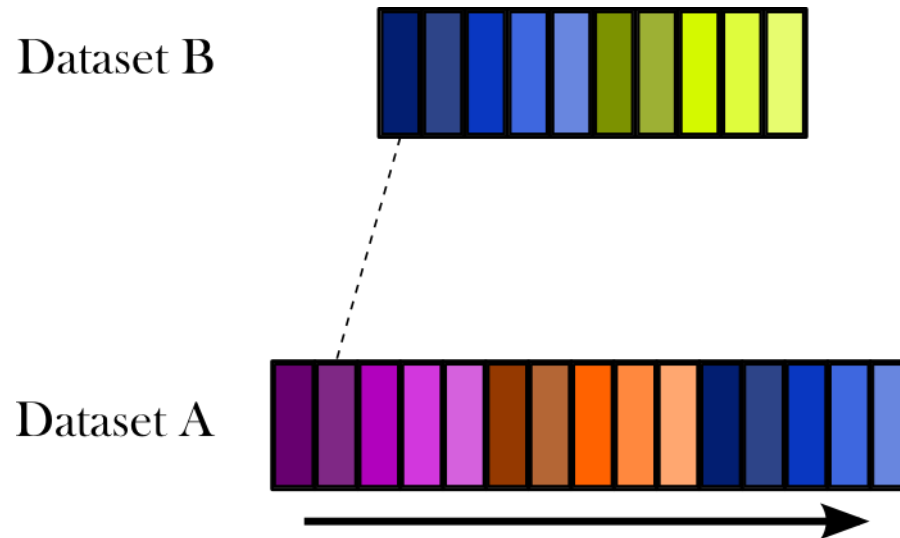
Dataset A



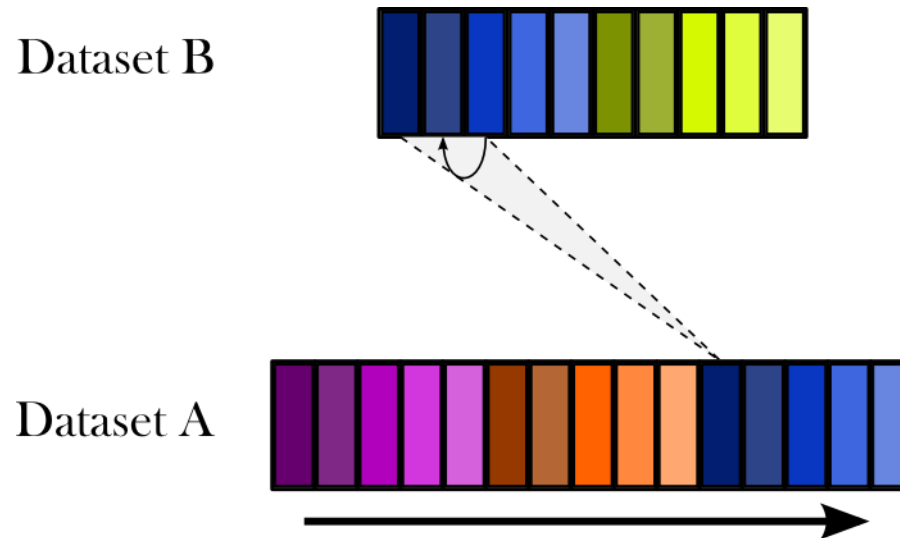
First optimization



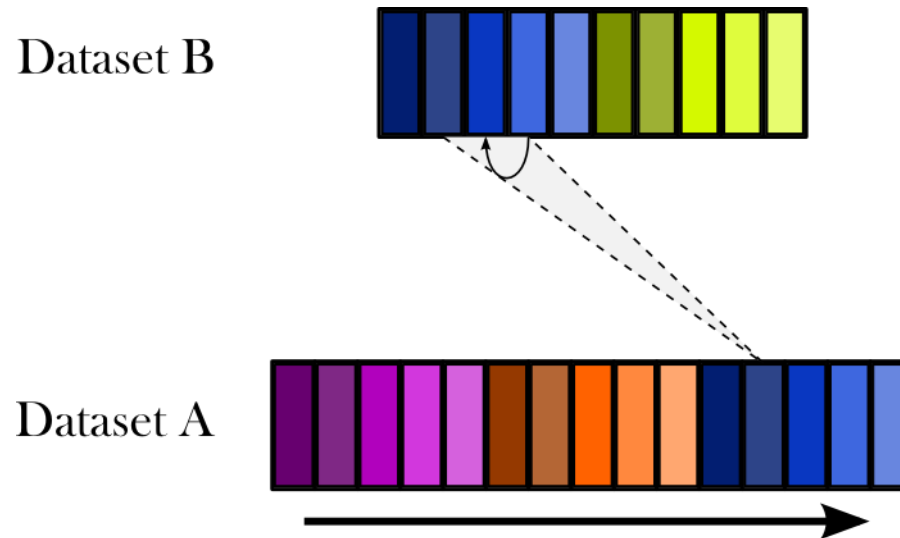
First optimization



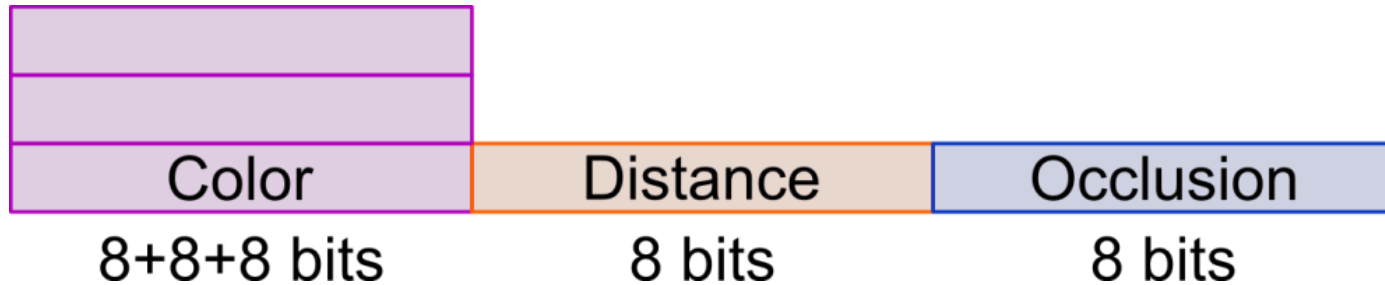
First optimization



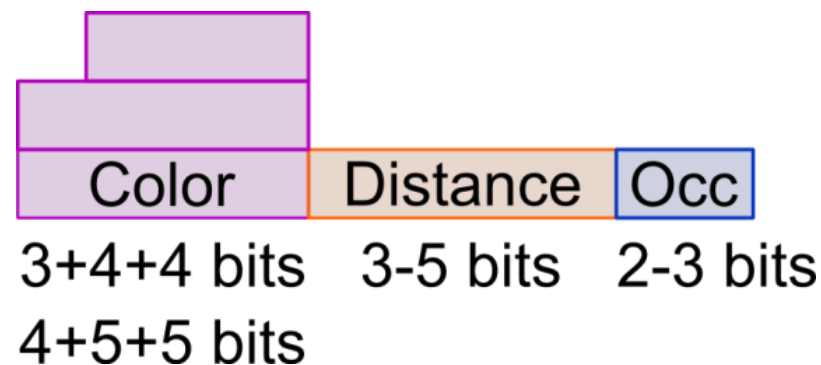
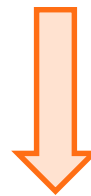
First optimization



Second optimization

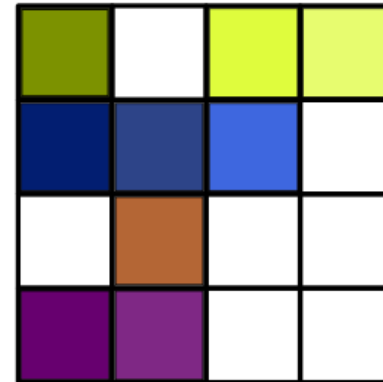


Second optimization



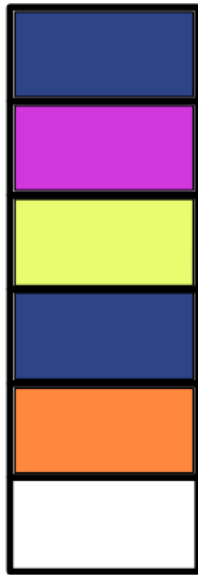
Second optimization

**Packed probe sample
bit set**



Second optimization

**Packed probe sample
sequence**

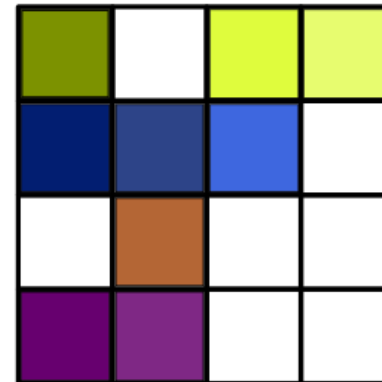


Second optimization

Sampled model

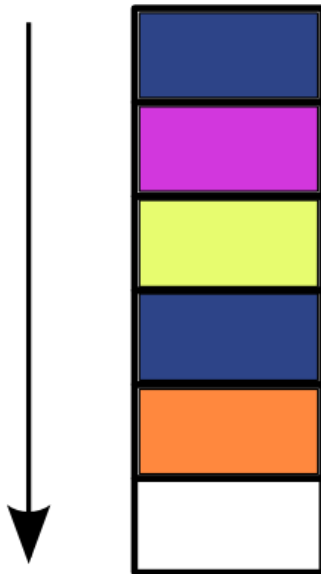


Query volume

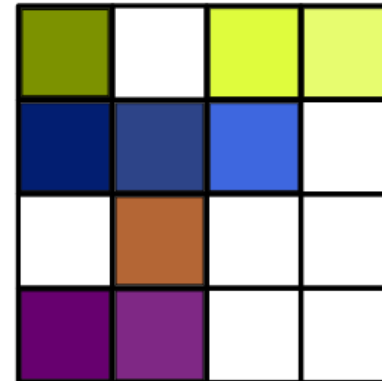


Second optimization

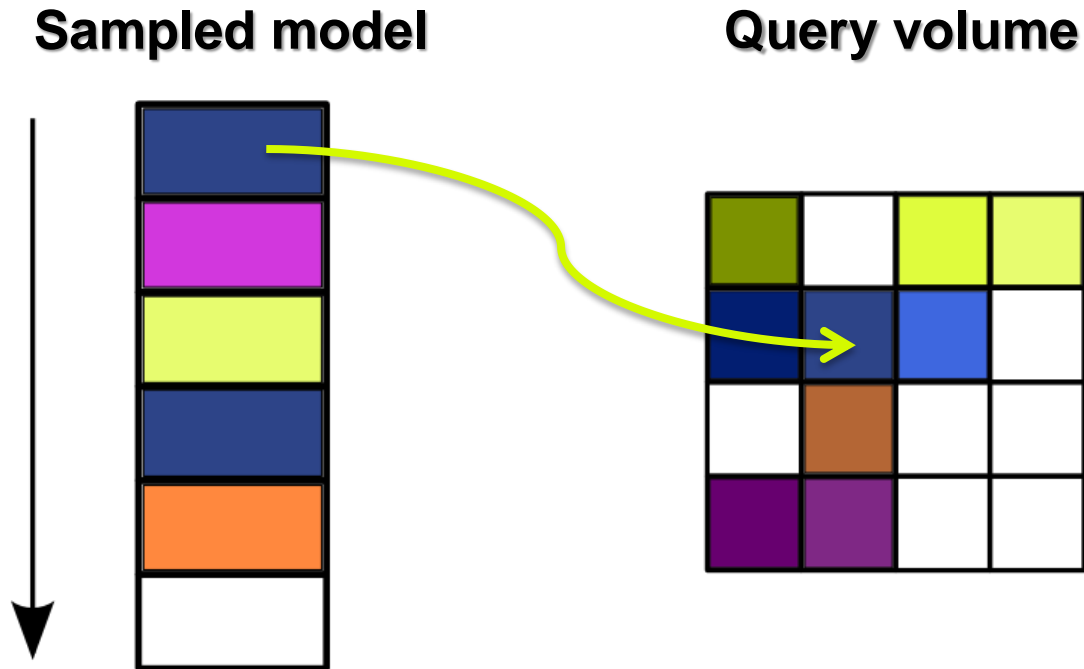
Sampled model



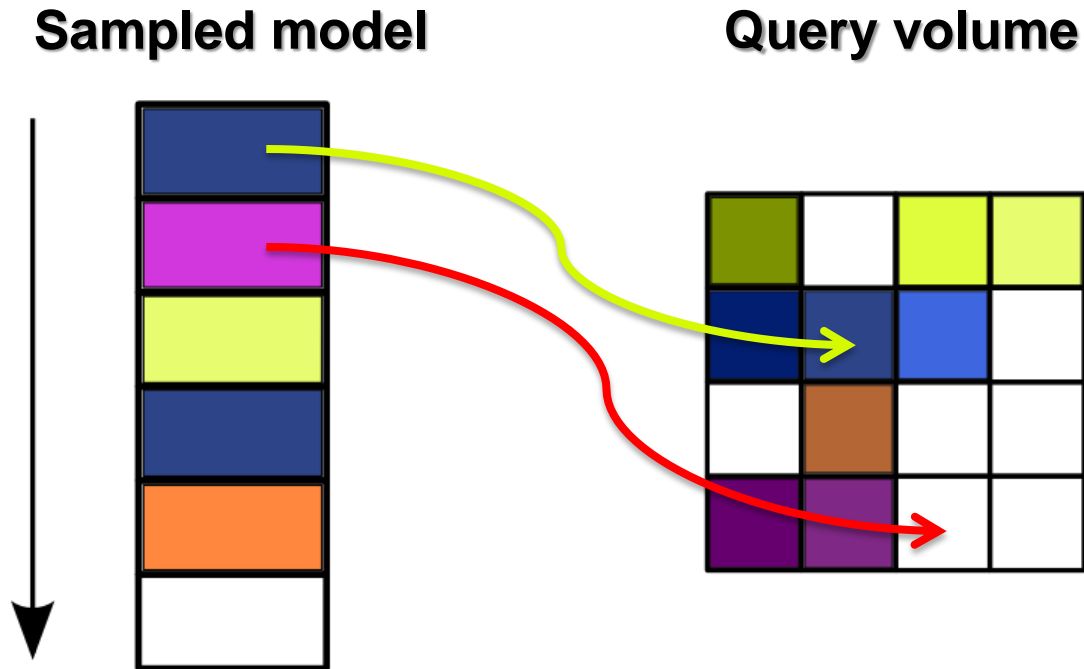
Query volume



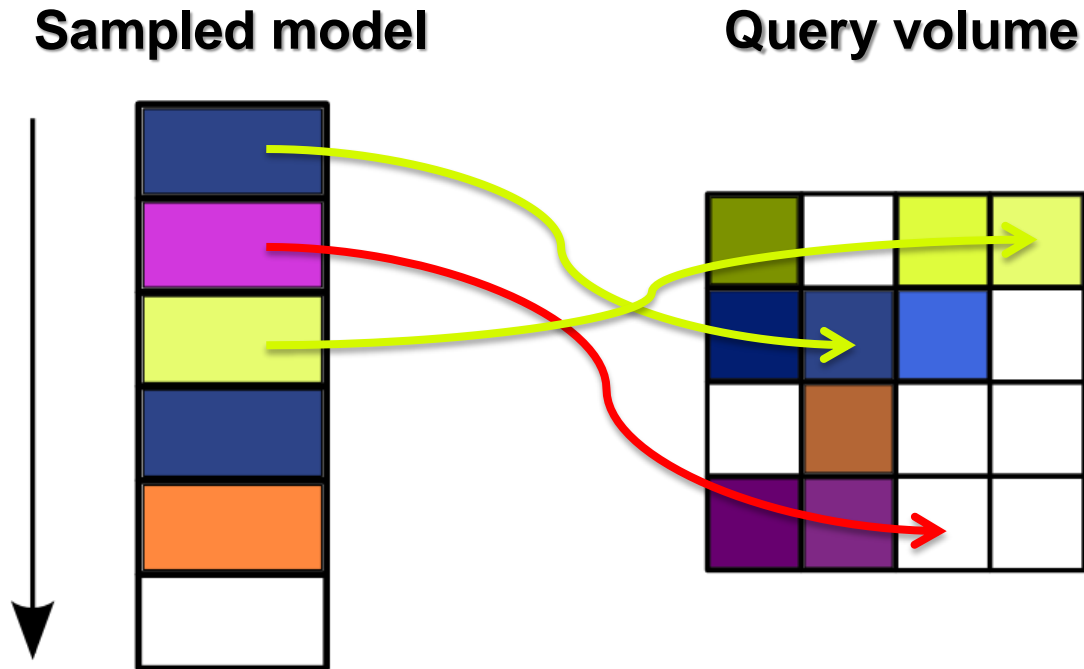
Second optimization



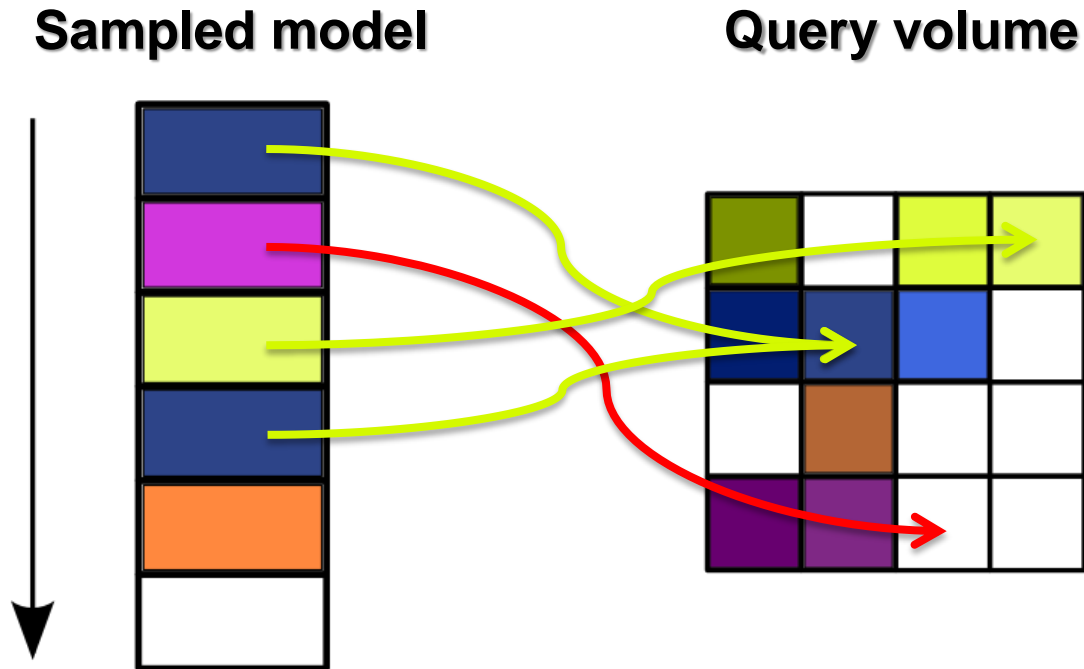
Second optimization



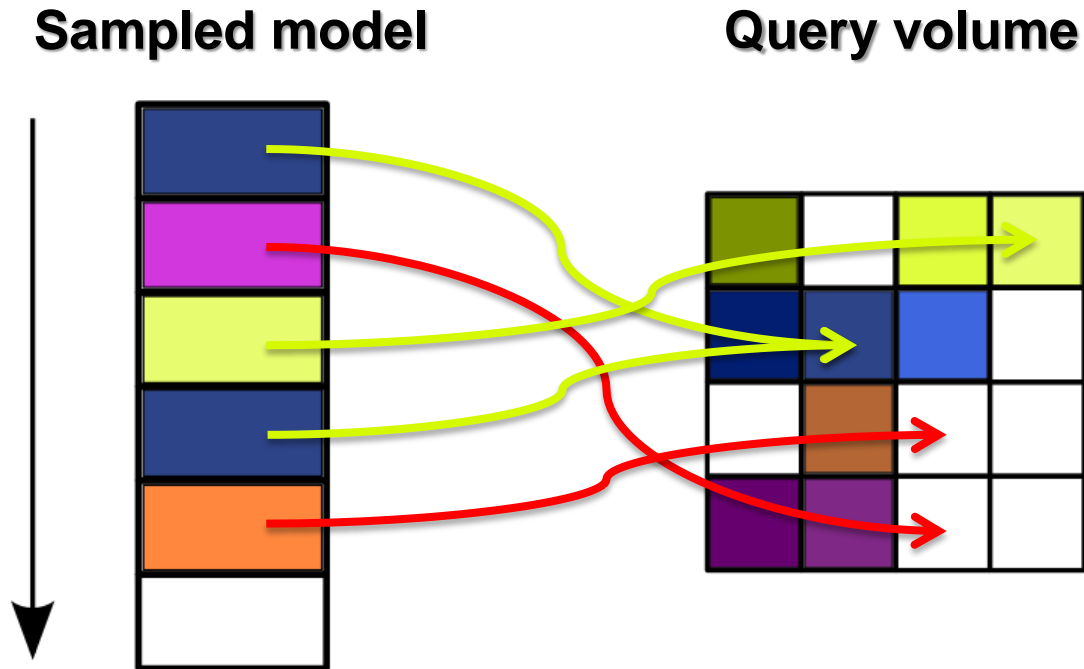
Second optimization



Second optimization



Second optimization



Second optimization

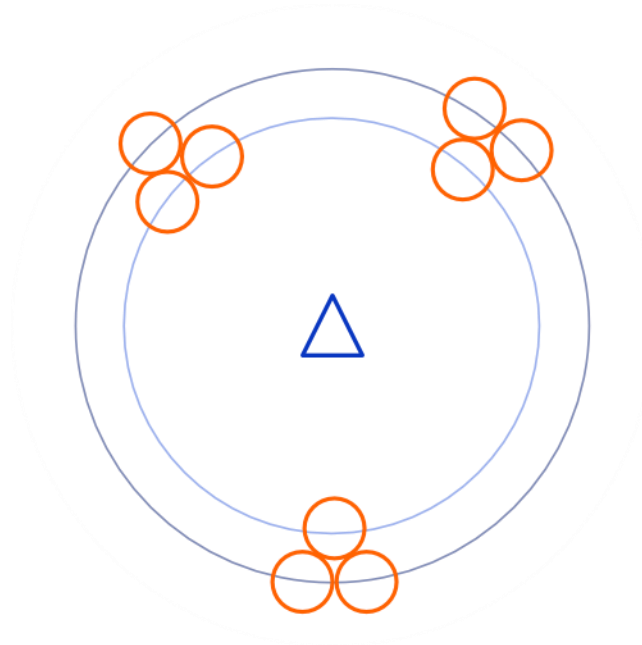
Sampled model



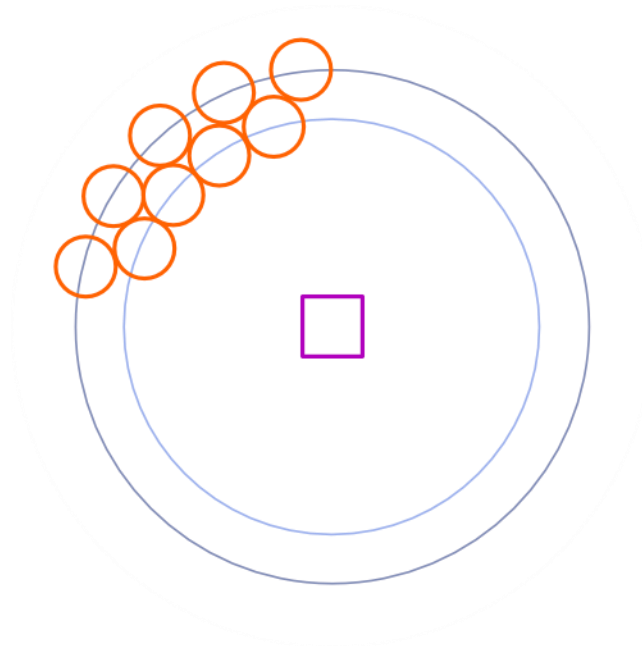
Query volume



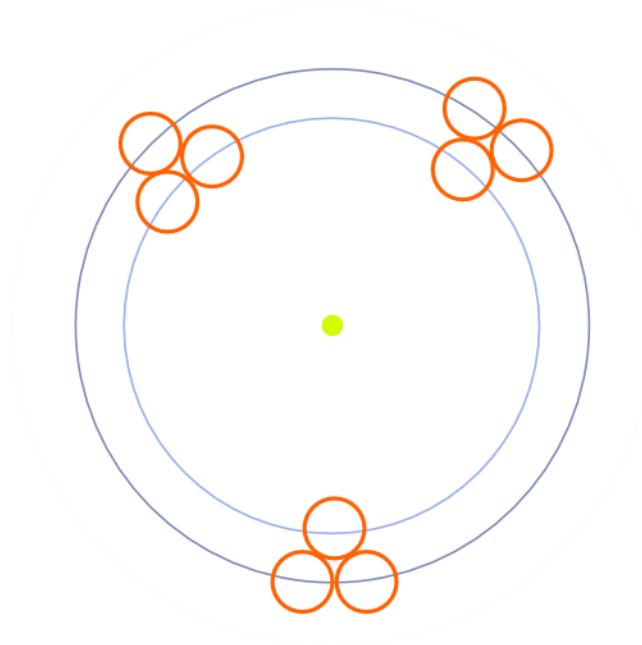
Neighborhood context



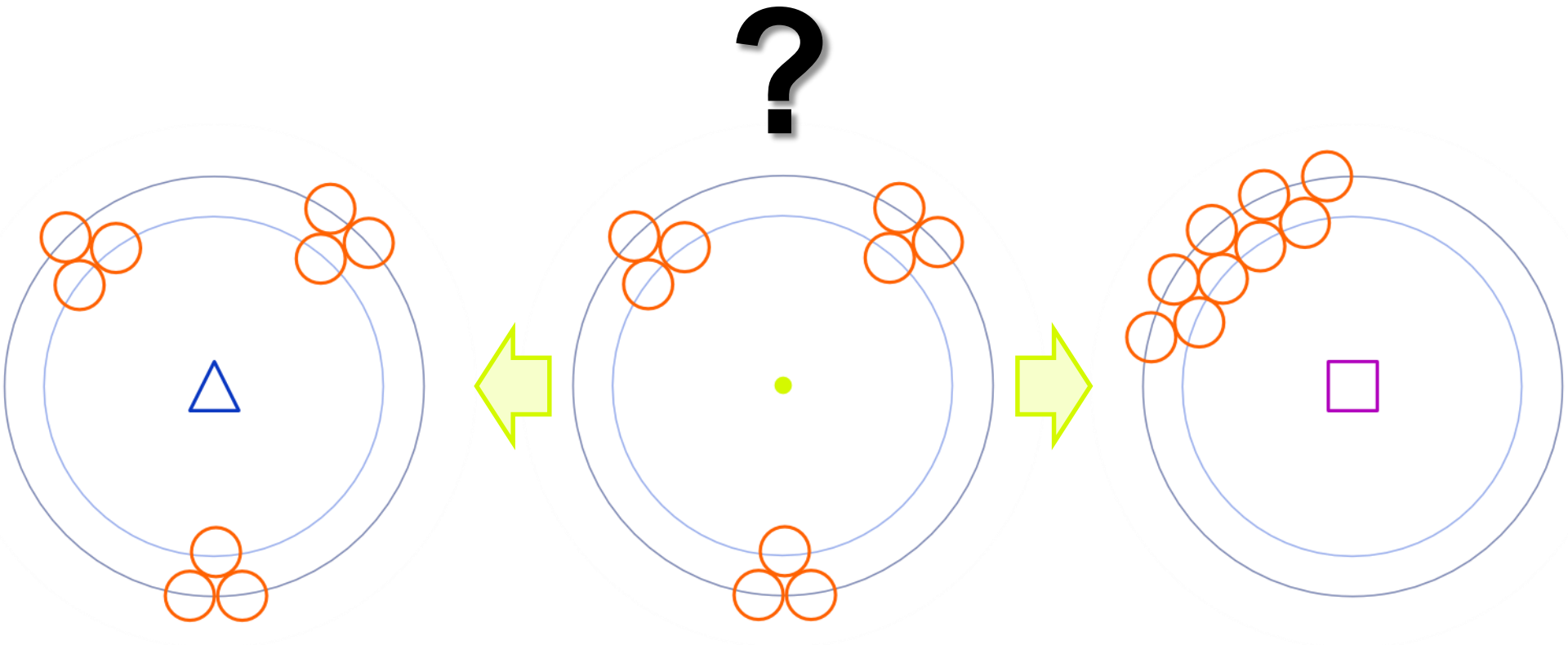
Neighborhood context



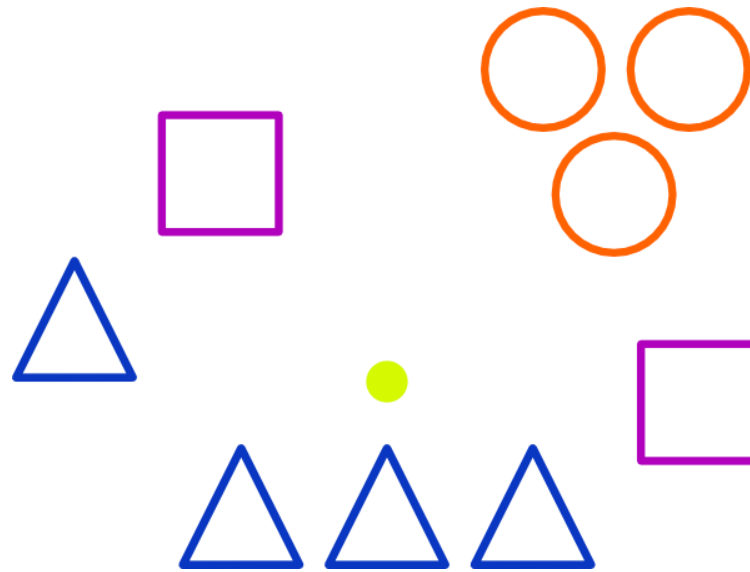
Neighborhood context



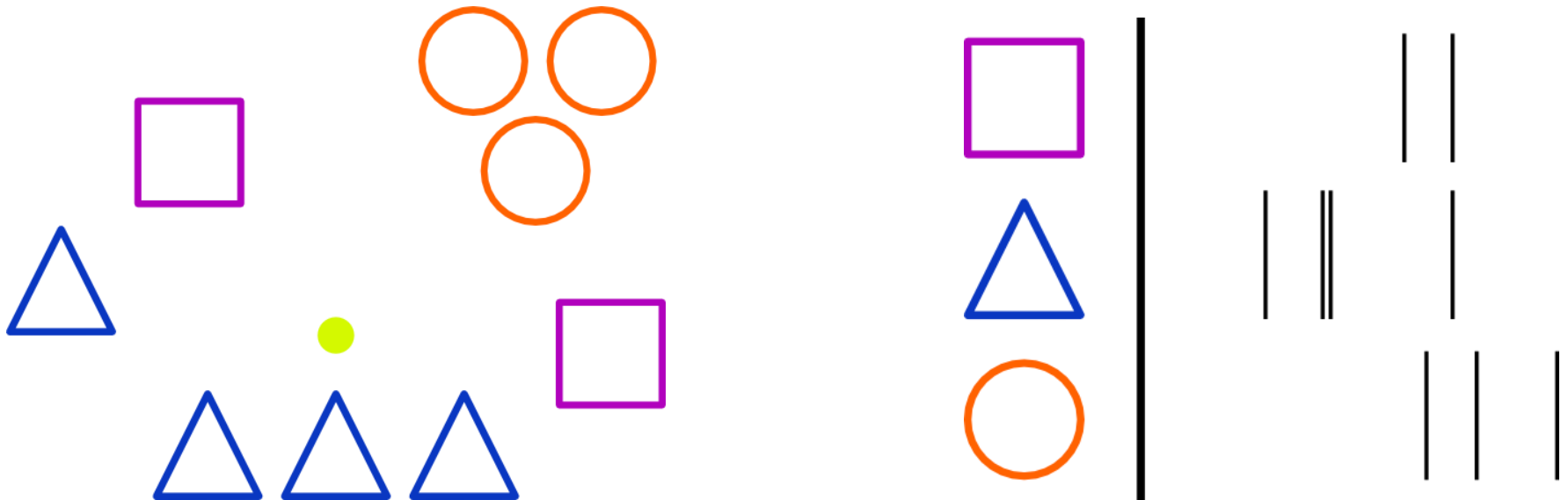
Neighborhood context



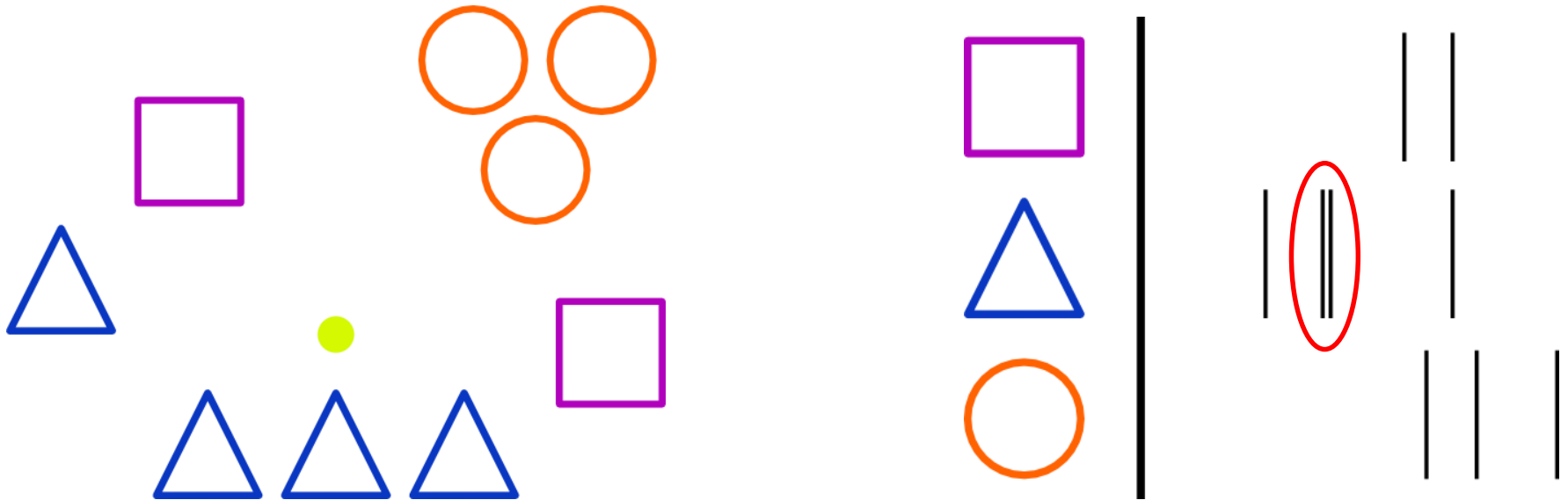
Comparing neighborhoods



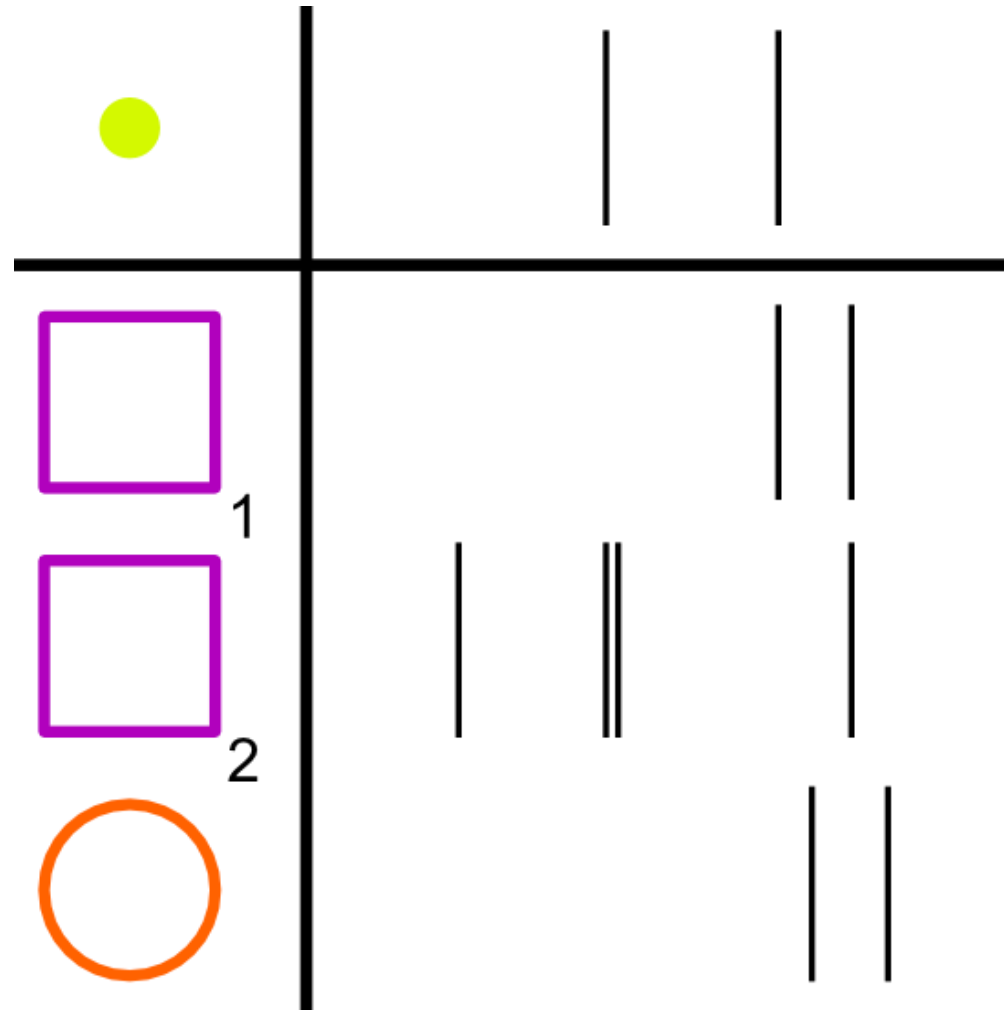
Comparing neighborhoods



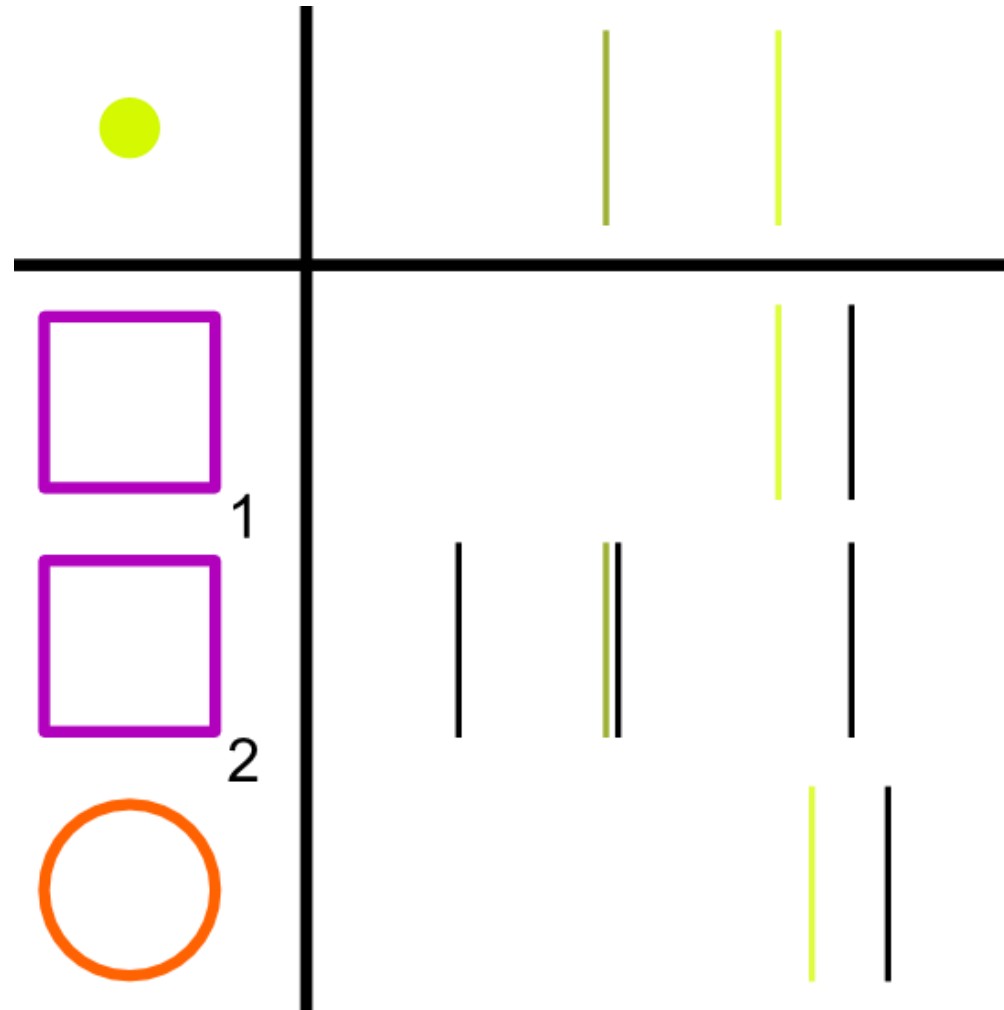
Comparing neighborhoods



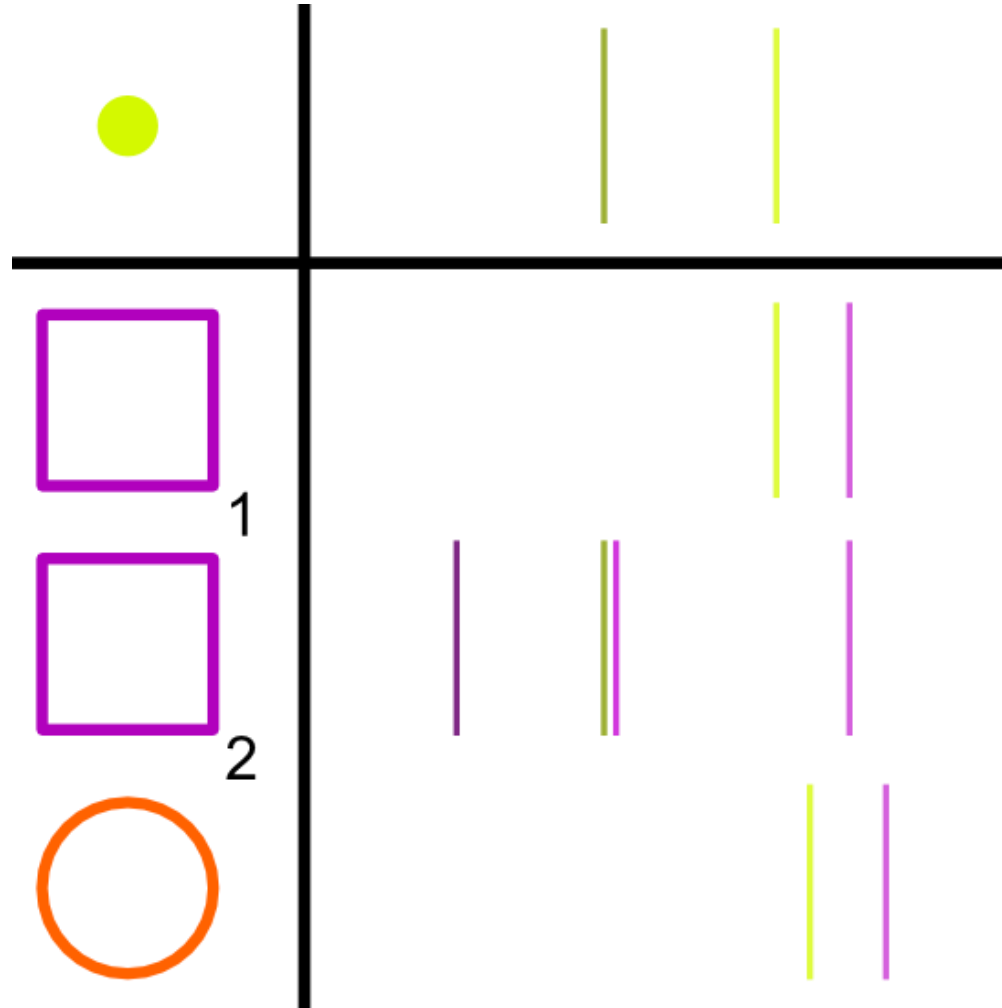
Matching distance groups



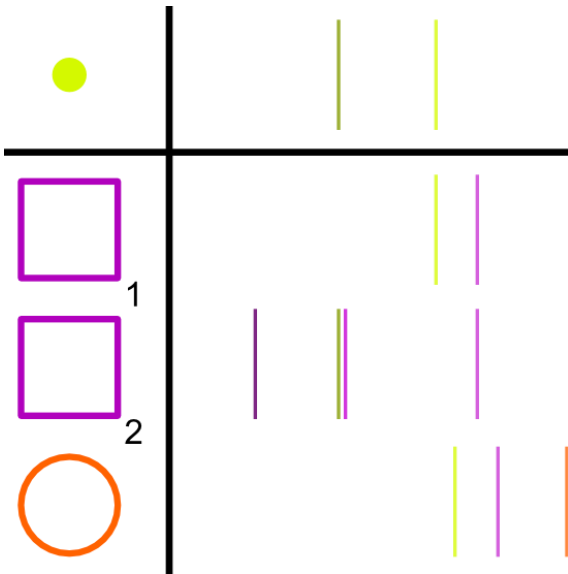
Matching distance groups



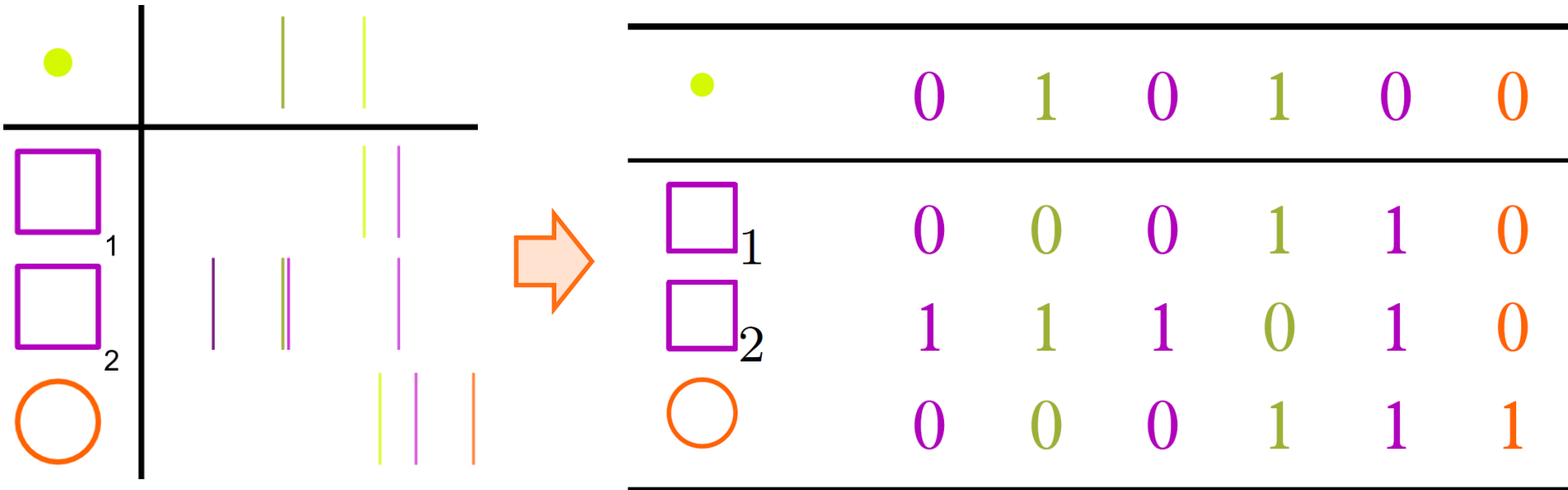
Matching distance groups







Matching distance groups



Matching distance groups



Similarity measures

	0	1	0	1	0	0
 ₁	0	0	0	1	1	0
 ₂	1	1	1	0	1	0
	0	0	0	1	1	1

Rand similarity measure

$$R = \frac{m_{11} + m_{00}}{m_{11} + m_{10} + m_{01} + m_{00}}$$

●	0	1	0	1	0	0
□ ₁	0	0	0	1	1	0
□ ₂	1	1	1	0	1	0
○	0	0	0	1	1	1

Rand similarity measure

$$R = \frac{m_{11} + m_{00}}{m_{11} + m_{10} + m_{01} + m_{00}}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	= $\frac{4}{6}$
□ ₂	1	1	1	0	1	0	
○	0	0	0	1	1	1	

Rand similarity measure

$$R = \frac{m_{11} + m_{00}}{m_{11} + m_{10} + m_{01} + m_{00}}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	
□ ₂	1	1	1	0	1	0	= $\frac{2}{6}$
○	0	0	0	1	1	1	





Rand similarity measure

$$R = \frac{m_{11} + m_{00}}{m_{11} + m_{10} + m_{01} + m_{00}}$$

●	0	1	0	1	0	0		
□ ₁	0	0	0	1	1	0		
□ ₂	1	1	1	0	1	0		
○	0	0	0	1	1	1	=	3/6

Jaccard index

$$R = \frac{m_{11}}{m_{11} + m_{10} + m_{01}}$$

	0	1	0	1	0	0
 ₁	0	0	0	1	1	0
 ₂	1	1	1	0	1	0
	0	0	0	1	1	1

Jaccard index

$$R = \frac{m_{11}}{m_{11} + m_{10} + m_{01}}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	= $\frac{1}{3}$
□ ₂	1	1	1	0	1	0	
○	0	0	0	1	1	1	

Jaccard index

$$R = \frac{m_{11}}{m_{11} + m_{10} + m_{01}}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	
□ ₂	1	1	1	0	1	0	= $\frac{1}{5}$
○	0	0	0	1	1	1	

Jaccard index

$$R = \frac{m_{11}}{m_{11} + m_{10} + m_{01}}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	
□ ₂	1	1	1	0	1	0	
○	0	0	0	1	1	1	= $\frac{1}{4}$

Importance-weighted Rand measure





$$R = \frac{\sum \Phi_k \mathbb{1}_{\{\text{query}_k = \text{instance}_k\}}}{\sum \Phi_k}$$

$$\Phi_k = h(\text{query}_k) + h(\text{instance}_k)$$

●	0	1	0	1	0	0
□ ₁	0	0	0	1	1	0
□ ₂	1	1	1	0	1	0
○	0	0	0	1	1	1

Importance-weighted Rand measure

$$R = \frac{\sum \Phi_k \mathbb{1}_{\{\text{query}_k = \text{instance}_k\}}}{\sum \Phi_k}$$

	0	1	0	1	0	0
 ₁	0	0	0	1	1	0
 ₂	1	1	1	0	1	0
	0	0	0	1	1	1





Importance-weighted Rand measure

$$R = \frac{\sum \Phi_k \mathbb{1}_{\{\text{query}_k = \text{instance}_k\}}}{\sum \Phi_k}$$

●	0	1	0	1	0	0	
□ ₁	0	0	0	1	1	0	= $\frac{\Phi_1 + \Phi_3 + \Phi_4 + \Phi_6}{\sum \Phi_k}$
□ ₂	1	1	1	0	1	0	
○	0	0	0	1	1	1	











Importance-weighted Rand measure

$$R = \frac{\sum \Phi_k \mathbb{1}_{\{\text{query}_k = \text{instance}_k\}}}{\sum \Phi_k}$$

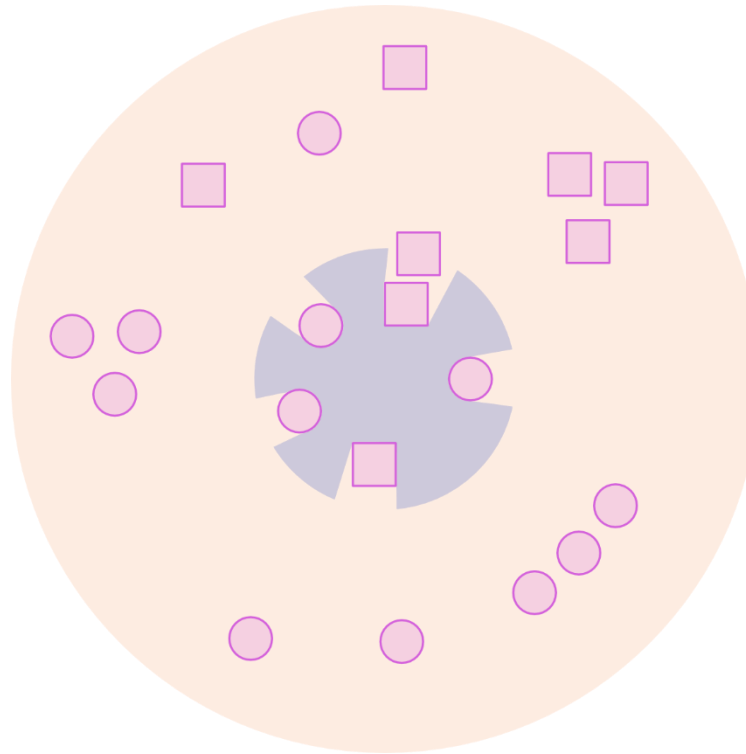
	0	1	0	1	0	0	
 ₁	0	0	0	1	1	0	
 ₂	1	1	1	0	1	0	= $\frac{\Phi_2 + \Phi_6}{\sum \Phi_k}$
	0	0	0	1	1	1	

Importance-weighted Rand measure

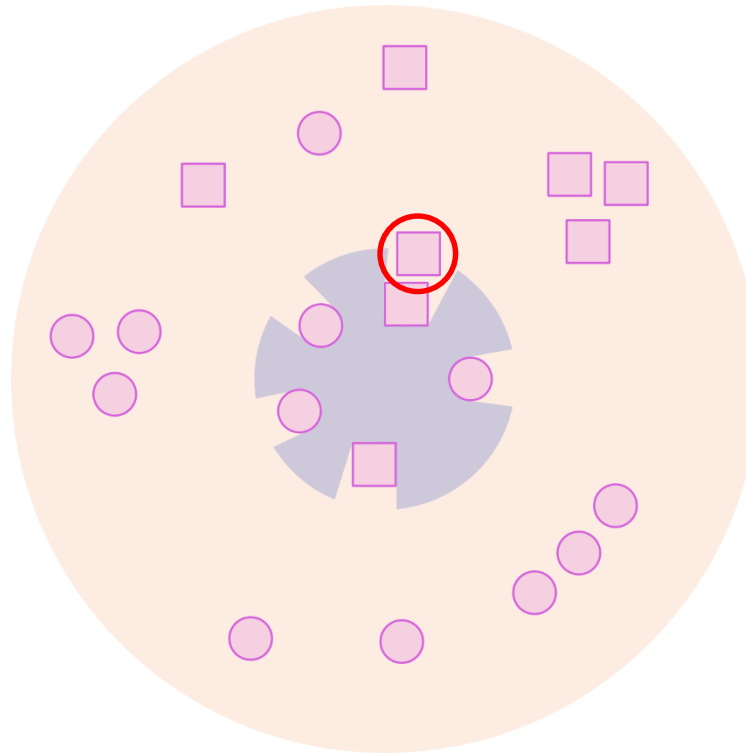
$$R = \frac{\sum \Phi_k \mathbb{1}_{\{\text{query}_k = \text{instance}_k\}}}{\sum \Phi_k}$$

	0	1	0	1	0	0	
 ₁	0	0	0	1	1	0	
 ₂	1	1	1	0	1	0	
							$= \frac{\Phi_1 + \Phi_3 + \Phi_4}{\sum \Phi_k}$

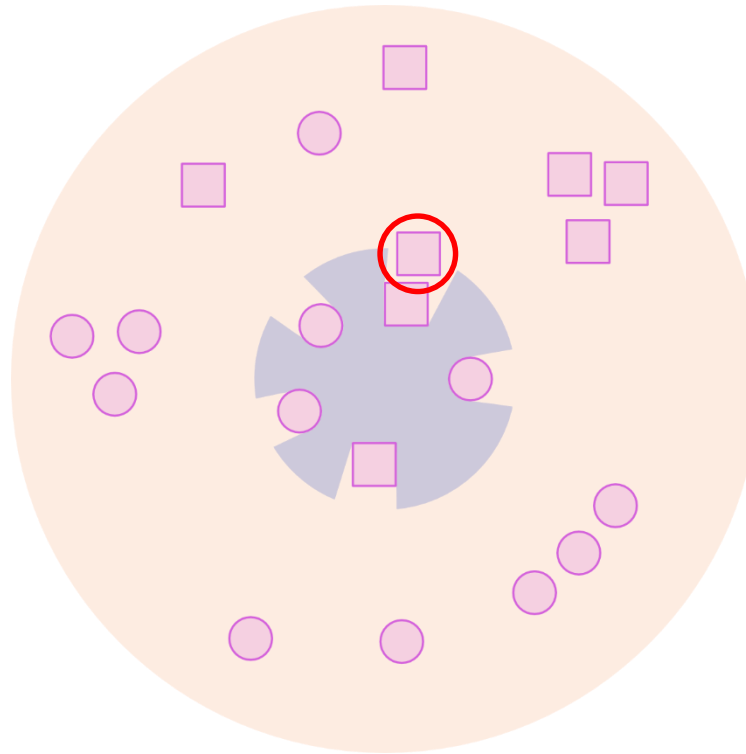
Combining context scores



Combining context scores



Combining context scores



$$S_{\text{final}} = S_{\text{probe}} \cdot S_{\text{neighborhood}}$$

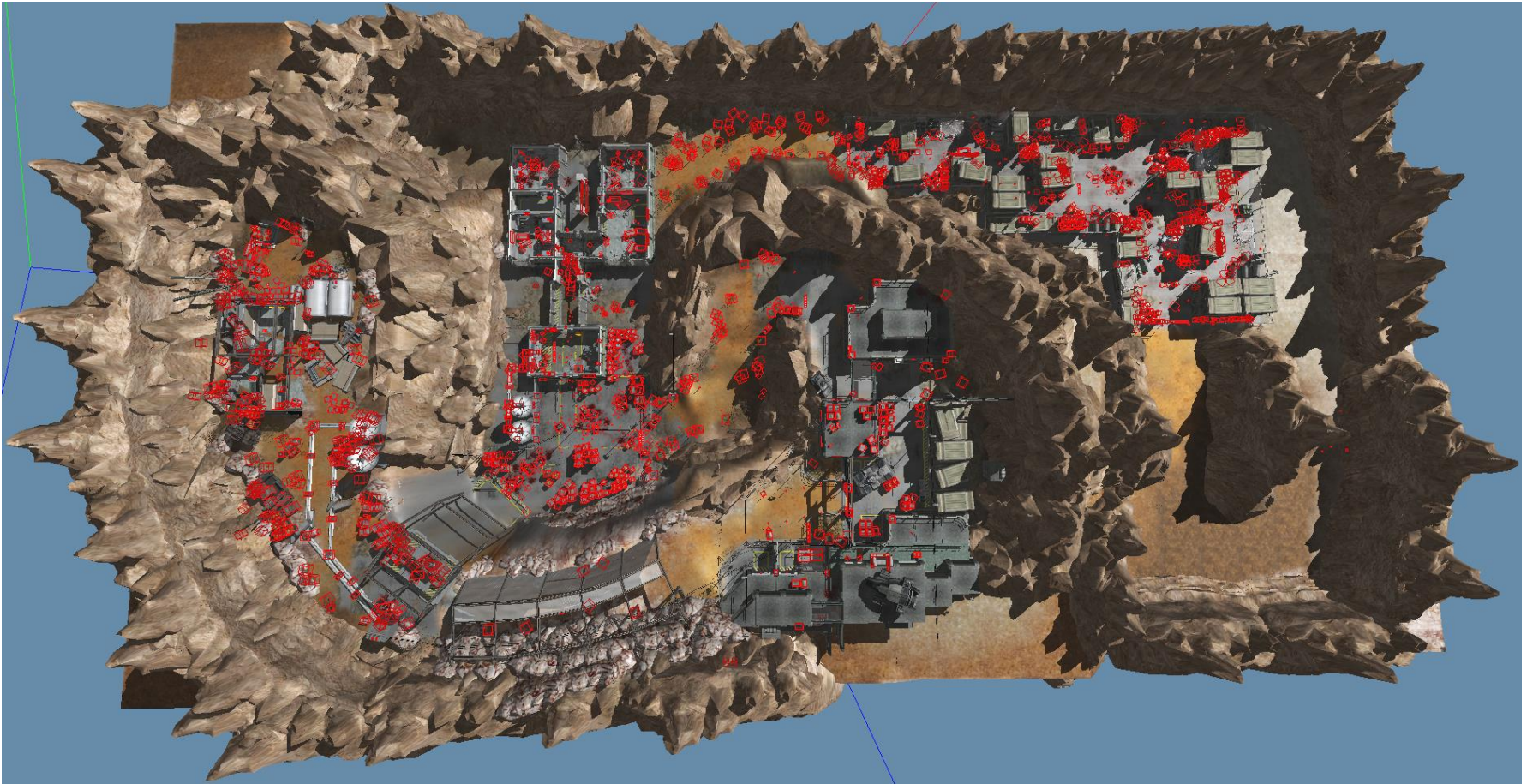
Results



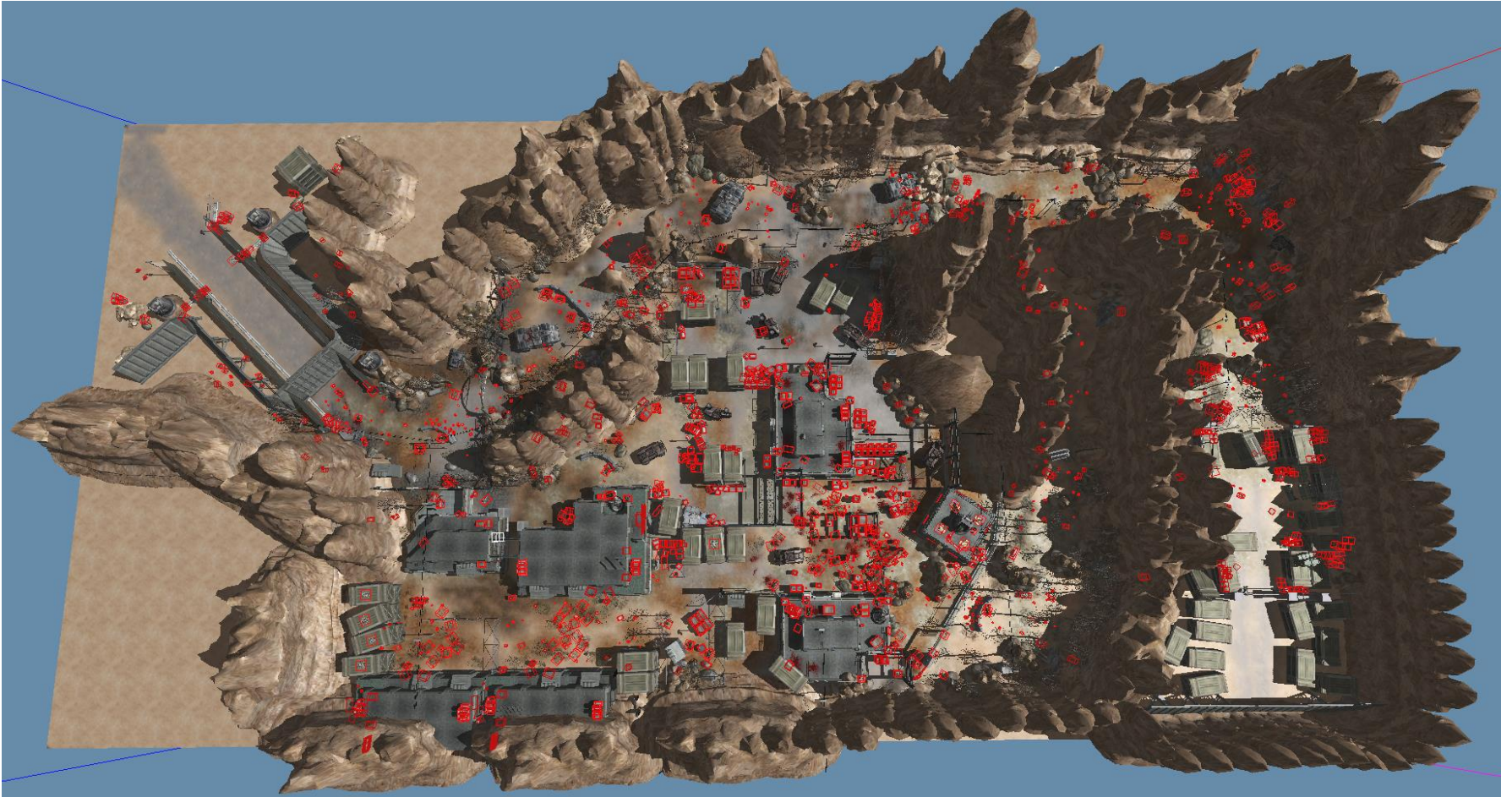
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Validation

marine01_wakeup



marine02_road



Validation: probe context

	# models	# instances	avg rank (model frequency)
marine02_road	94	1396	17.15
marine01_wakeup	150	2066	21.10

Level	marine02_road	marine02_road (5)	marine01_wakeup
Uniform bidirectional match	14.1	17.7	15.2
IW bidirectional match	12.9	17.6	14.3
Configuration	10.5	16.0	10.4

Validation: neighborhood context

	# models	# instances	avg rank (model frequency)
marine02_road	94	1396	17.15
marine01_wakeup	150	2066	21.10

Max random shift	2			4			8		
	10	20	60	10	20	60	10	20	60
Rand measure	0.38	0.43	0.46	1.16	1.20	1.23	3.38	2.90	2.91
IW measure	0.40	0.42	0.45	1.23	1.21	1.21	3.57	2.92	2.90
Jaccard index	0.38	0.43	0.46	1.14	1.20	1.23	2.86	2.88	2.90

Validation: combined contexts

	# models	# instances	avg rank (model frequency)
marine02_road	94	1396	17.15
marine01_wakeup	150	2066	21.10

	marine02_road			marine02_road (5)			marine01_wakeup		
	Rand	IW	Jaccard	Rand	IW	Jaccard	Rand	IW	Jaccard
Uniform bidirectional	7.14	11.0	0.230	7.34	12.6	0.176	8.85	12.3	0.651
IW bidirectional	8.09	11.0	0.399	9.81	14.4	0.320	9.79	12.5	0.926
Configuration	11.1	11.5	4.40	15.9	16.5	5.83	21.8	22.2	15.2

Performance

	Rand measure	Importance-weighted measure	Jaccard index
marine02_road	9.5	11	9.5
marine01_wakeup	38	44	39

	Bidirectional query	IW bidirectional query	Configuration query
marine02_road	1.2	5.4	900
marine02_road (5)	14	93	18000
marine01_wakeup	2.4	11	1500

Summary

Post Mortem



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Assisted Object Placement

Andreas Kirsch

Questions?

Demo