

Méthodes basées sur les pavages pour informatique graphique

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Avec participation de

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Branko Grunbaum University of Washington

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Charles Donohue University of Montreal

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Florent Wachtel Université de Lyon

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Mathieu Desbrun Caltech

Fernando de Goes Caltech

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What is Tiling?





Victor Ostromoukhov

Professor (PR1)

Presently: [Université de Lyon 1, CNRS-LIRIS](#)

Formerly: [Université de Montréal, DIRO](#)

[Contact information](#)

Bio:

Victor Ostromoukhov is a professor at Université de Lyon 1/CNRS. He graduated from Phys-Tech (MIPT, Moscow, 1980). He then spent several years with prominent European and American industrial companies (SG2, Paris; Olivetti, Paris and Milan; Canon Information Systems, Cupertino, CA) as a research scientist and/or computer engineer. He received his PhD in Computer Science from Swiss Federal Institute of Technology (EPFL, Lausanne, 1995), where he then continued to work as a lecturer and senior researcher. He was an invited professor at University of Washington (Seattle, 1997) and at iMAGIS/INRIA (Grenoble, 2002). He spent a year as research scientist at Massachusetts Institute of Technology, Cambridge, MA, in 1999-2000. He has been an Associate Professor at Université de Montréal, from 2000 to 2009. He has been appointed as Full Professor at Université de Lyon 1, since September 2009, In 2010, he received a

Distinguished Chair from French National Research Agency (Chaire d'Excellence ANR).

Curriculum Vitae: [short](#) or [detailed](#)

Research:

- Research interests: [Digital Halftoning](#), [Photorealistic](#) and [Nonphotorealistic Rendering](#), [Importance Sampling](#), [Color Science](#), [Tilings and Patterns](#), [Low-Discrepancy Sequences](#), [Mote Carlo and Quasi-Monte Carlo Methods](#)
- Recent publications: [Abstracts & PDFs](#), [list](#)

Teaching:

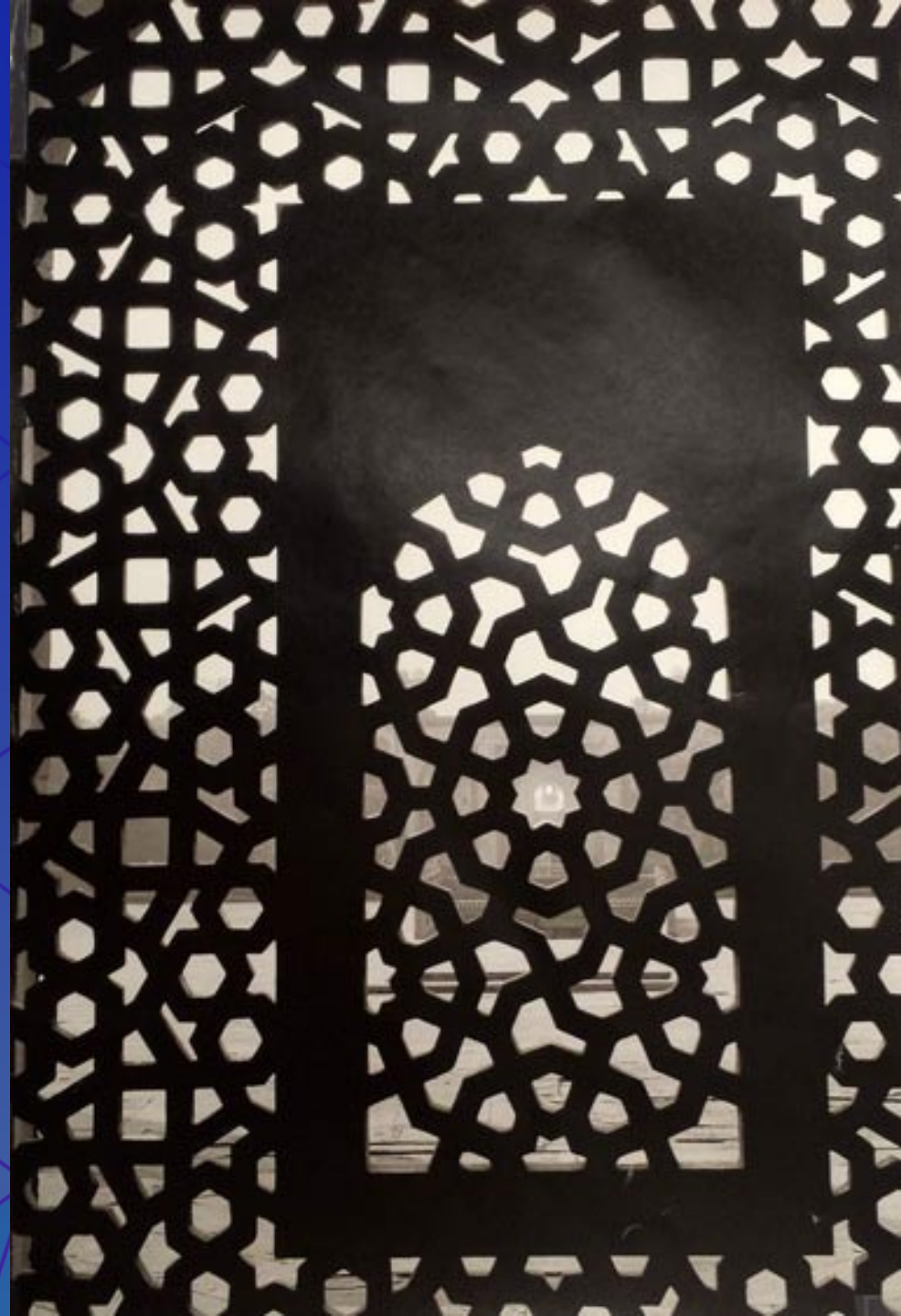
What is Tiling?

Definition **Tiling.** A *tiling* is a countable collection $\mathcal{T}$ of *tiles* $\{T_1, T_2, \dots\}$, such that:

1. Every tile is a closed topological disk.
2. Every point in the plane is contained in at least one tile;
3. The interiors of the tiles are pairwise disjoint; and
4. The tiles are uniformly bounded.



Tiling in Architecture



Tiling in Architecture



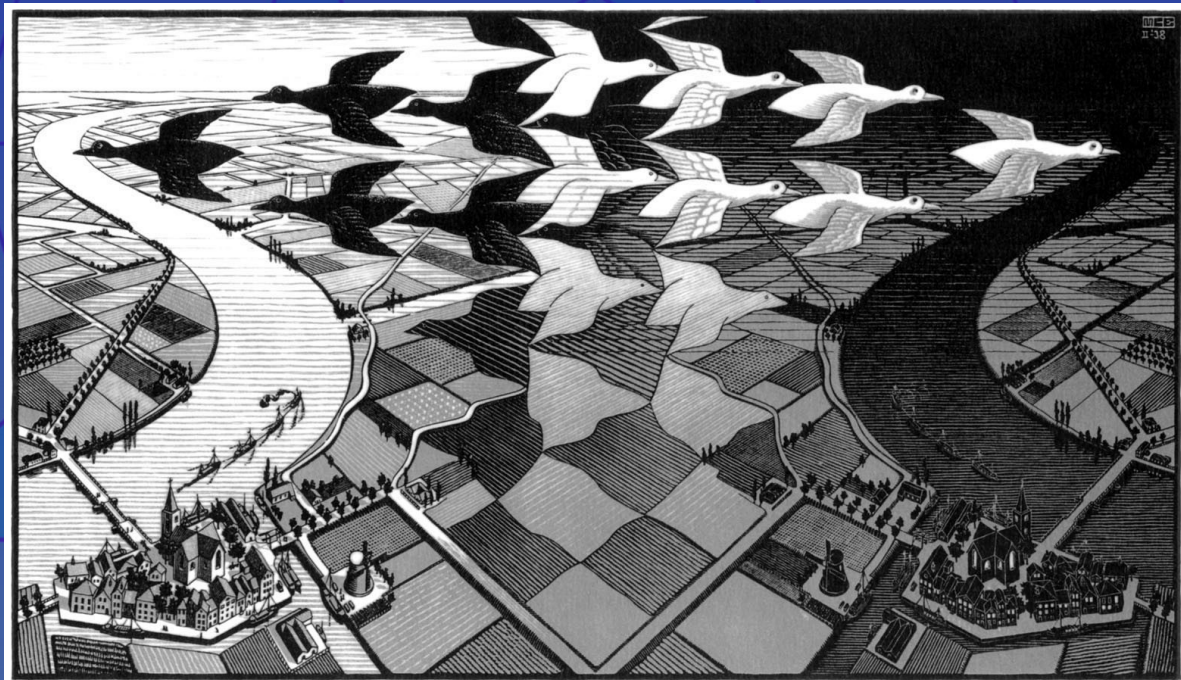
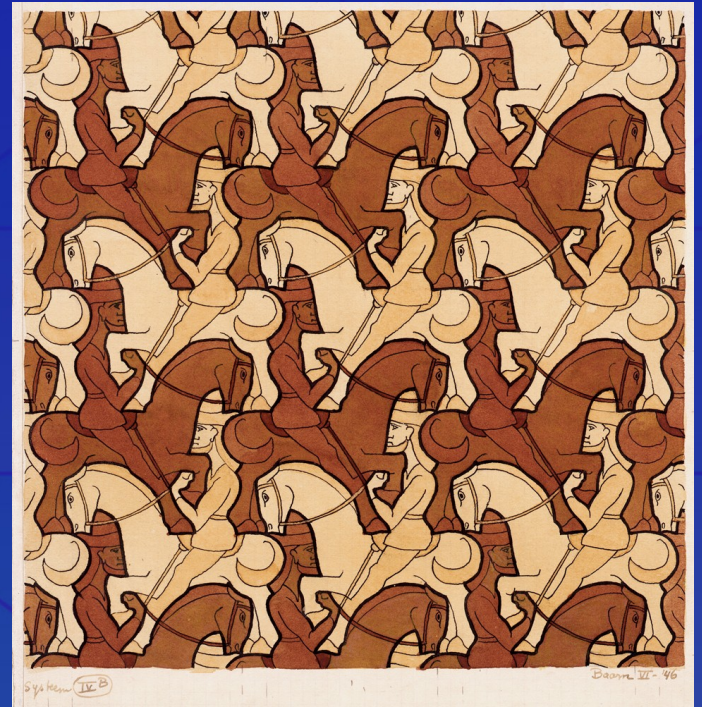
Tiling in Architecture



Tiling in Architecture



Escher



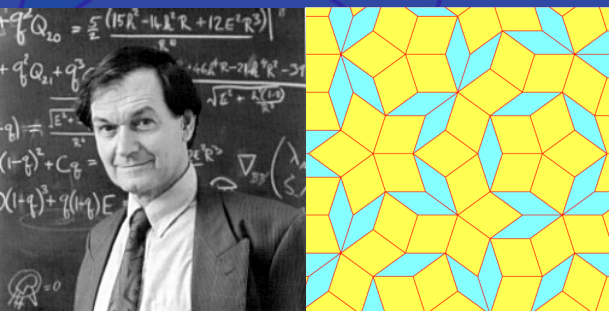
Penrose Tiling: Milestones



- Circa 1200 AD, Fibonacci (Leonardo of Pisa)
Rabbit Sequence, Fibonacci Numbers



- 1619, Johannes Kepler
Harmonice Mundi, 5-fold Tiling Problem



- 1974, Sir Roger Penrose
Pentaplexity, Penrose Tiling

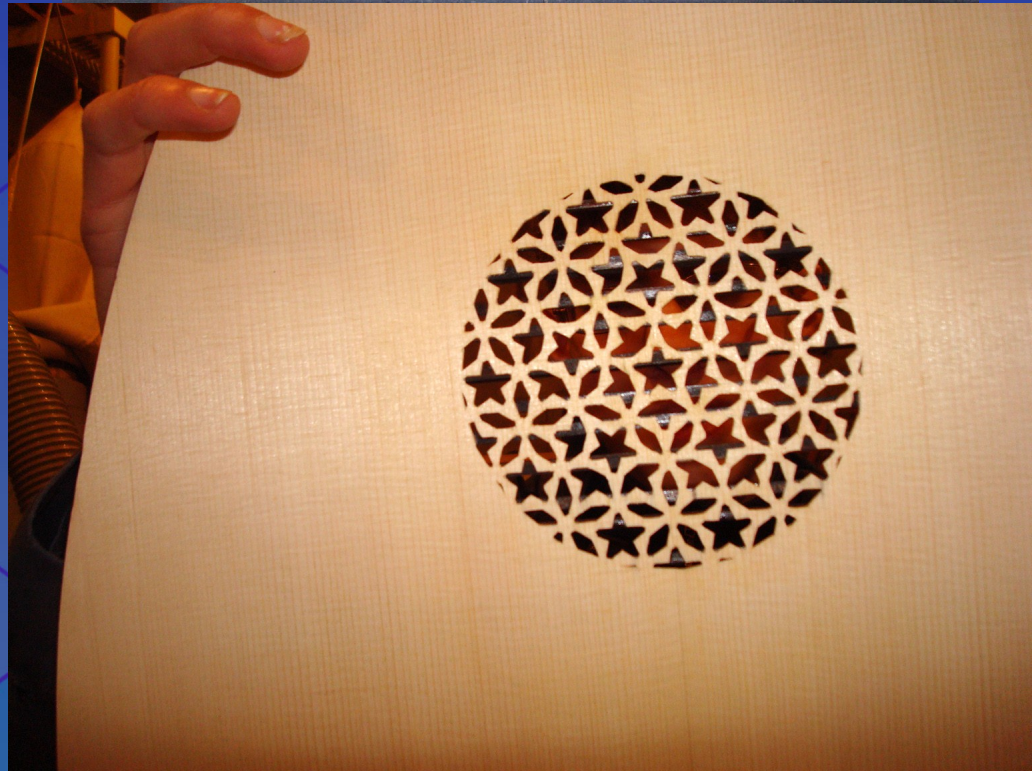
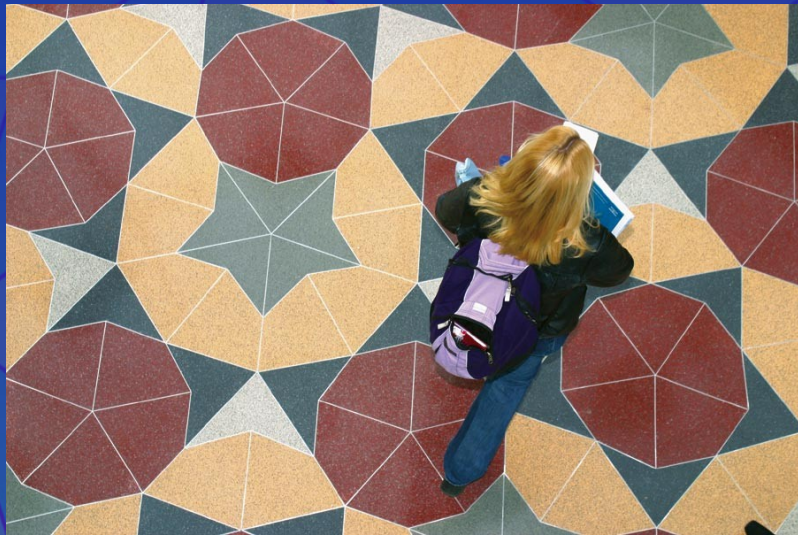
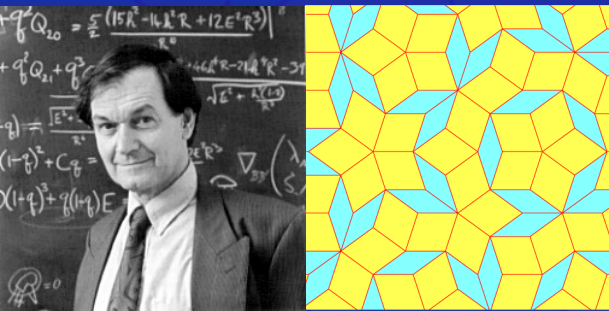
Diffraction
Pattern



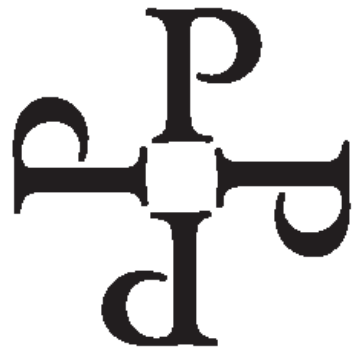
- 1984, Dan Shechtman *et al.*
Discovery of Quasi-Crystals



Penrose Tiling



SYMMETRY



$4\bullet \cdot c_4$



$5\bullet \cdot c_5$



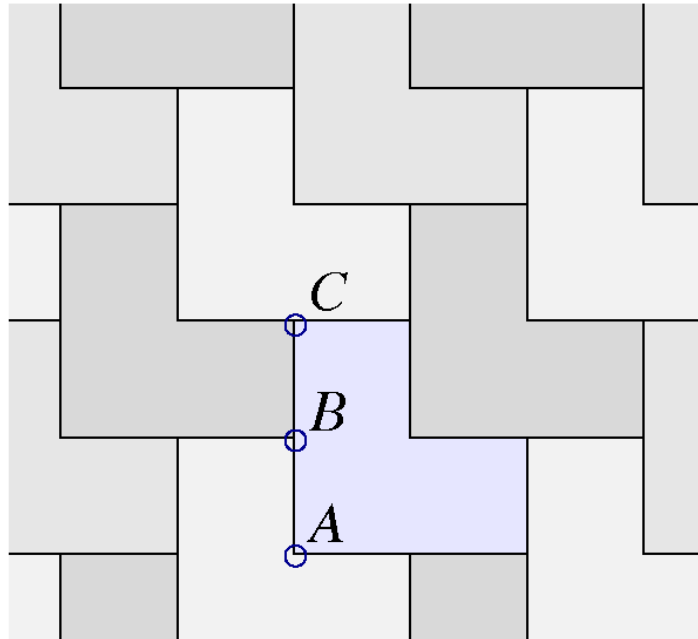
$*4\bullet \cdot d_4$



$*5\bullet \cdot d_5$

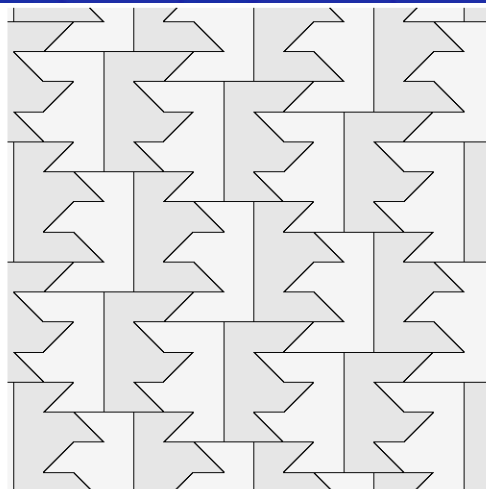
Examples of figures in the Euclidean plane with cyclic or dihedral symmetry. Each figure is labeled with both its orbifold signature and the name of the abstract group given by its symmetries.

MONOHEDRAL TILING WITH TRANSLATIONAL SYMMETRY

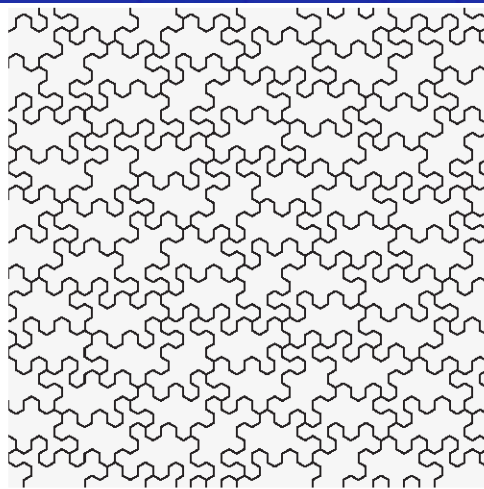


The features of a polygonal tiling. For the tile highlighted in blue, A is a shape vertex but not a tiling vertex, B is a tiling vertex but not a shape vertex, and C is both a tiling vertex and a shape vertex. Because the tiling vertices and shape vertices do not coincide, the tiling is not edge-to-edge.

MONOHEDRAL/POLYHEDRAL, ISOHEDRAL TILINGS

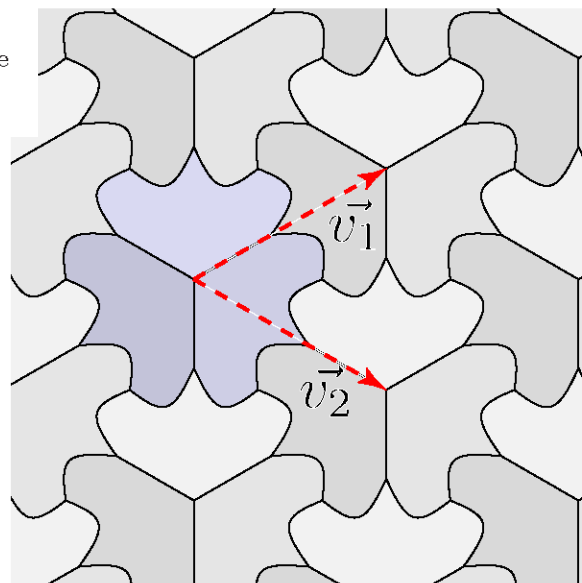


(a)



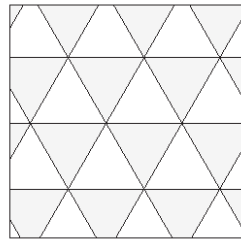
(b)

Anisohedral prototiles. Heesch's 1935 2-anisohedral prototile is shown in (a). The current record holder, a 10-anisohedral 16-hex discovered by Joseph Myers, is shown in (b).

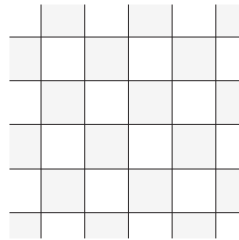


An example of an isohedral tiling of type IH16. A single translational unit of the tiling is shown through the two translation vectors $\vec{v}_1$ and $\vec{v}_2$ and the three coloured aspects

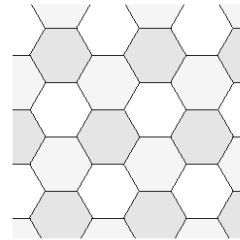
LAVES TILINGS



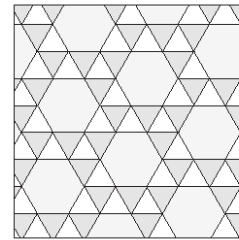
(3^6)



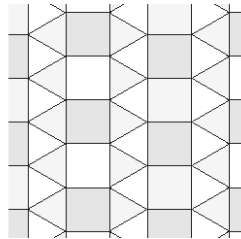
(4^4)



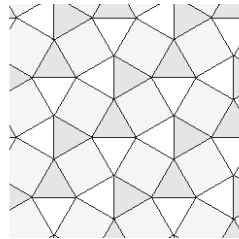
(6^3)



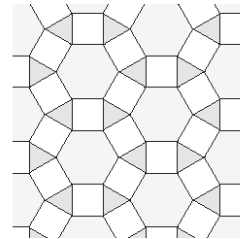
$(3^4.6)$



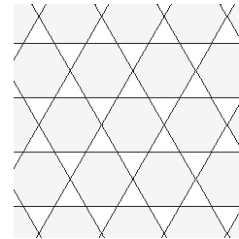
$(3^3.4^2)$



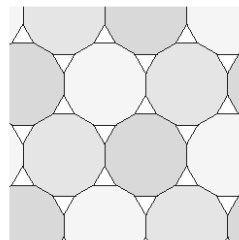
$(3^2.4.3.4)$



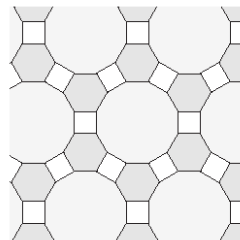
$(3.4.6.4)$



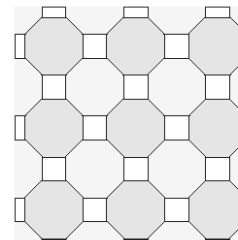
$(3.6.3.6)$



(3.12^2)



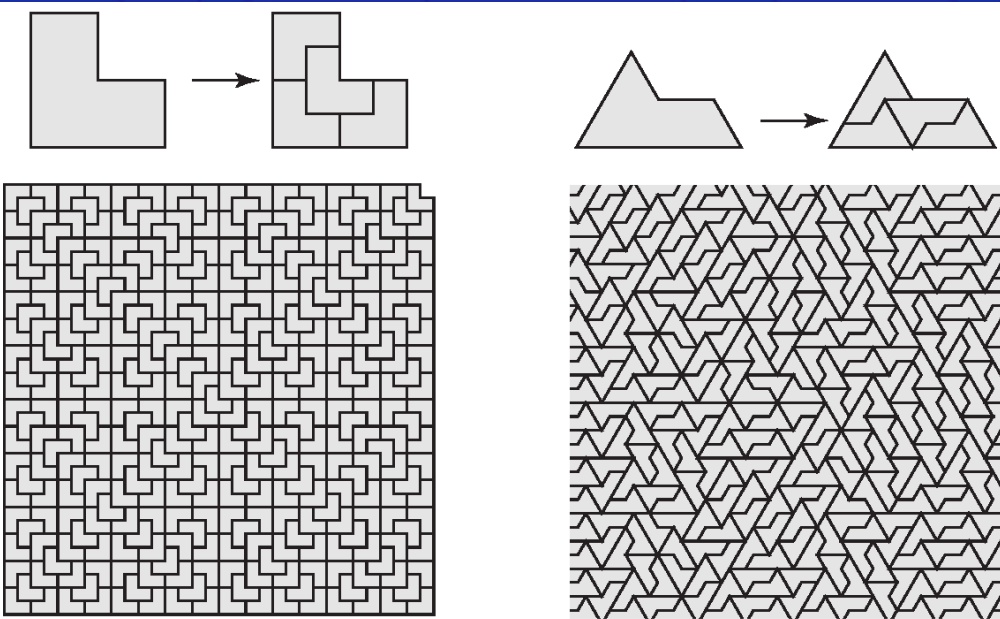
$(4.6.12)$



$(4.8.8)$

The eleven uniform Euclidean tilings, also known as Archimedean tilings. The tiling $(3^4.6)$ occurs in left-handed and right-handed forms.

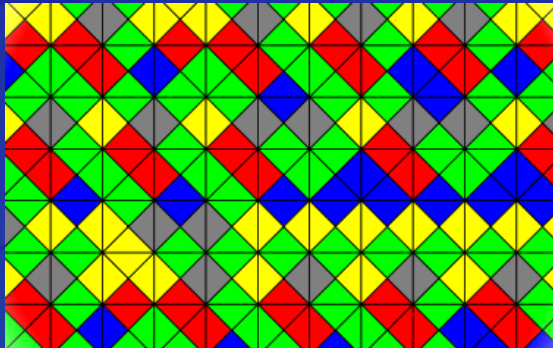
Non-Periodic vs. Aperiodic Tiling



Two well-known examples of rep-tiles: the Chair tiling on the left and the Sphinx tiling on the right.



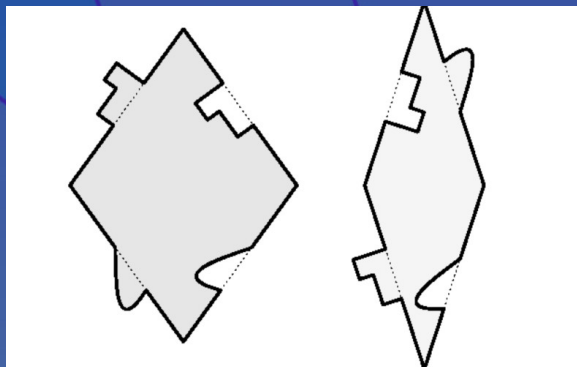
APERIODIC TILINGS



Wang Tiling



Robinson Tiling

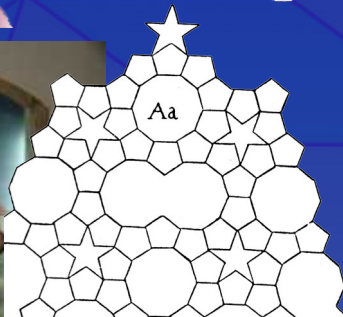


Penrose Tiling

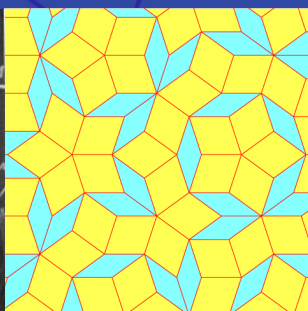
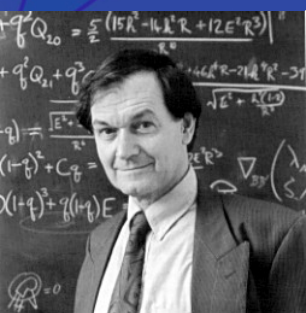
Penrose Tiling: Milestones



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Rabbit Sequence, Fibonacci Numbers

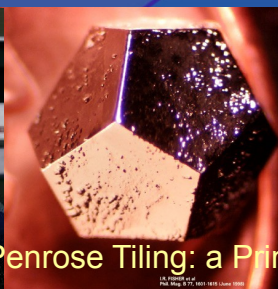
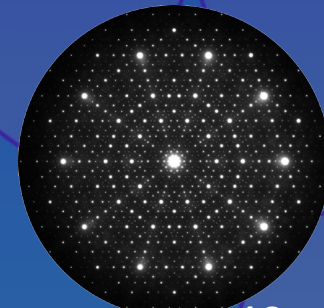


- 1619, Johannes Kepler
Harmonice Mundi, 5-fold Tiling Problem



- 1974, Sir Roger Penrose
Pentaplexity, Penrose Tiling

Diffraction
Pattern



- 1984, Dan Shechtman *et al.*
Discovery of Quasi-Crystals

Penrose-based Sampling System [SIGGRAPH'2004]

Fast Hierarchical Importance Sampling with Blue Noise Properties

Victor Ostromoukhov*
University of Montreal

Charles Donohue †
University of Montreal

Pierre-Marc Jodoin ‡
University of Montreal

Problem Statement

Given:

- Importance Density
 $I(x,y)$

high



low



Problem Statement

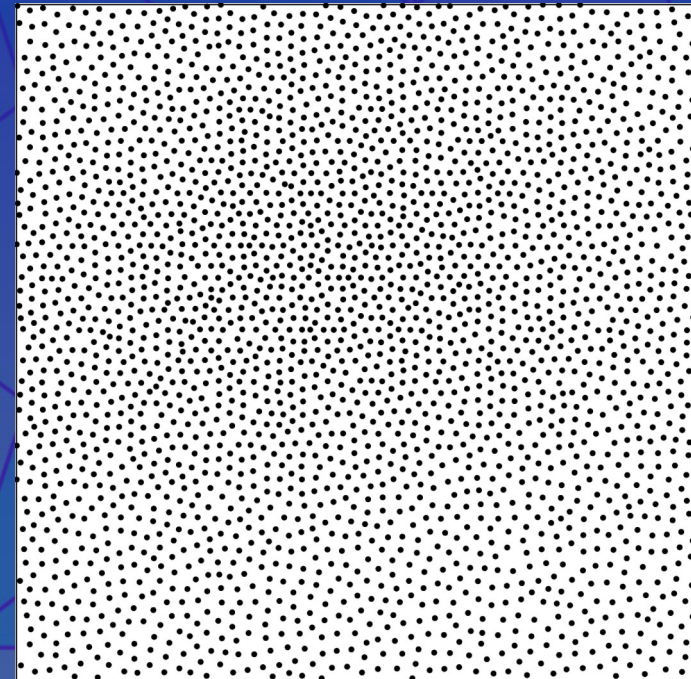
Given:

- Importance Density $I(x,y)$



Find:

- Discrete Sample Distribution Locally Proportional to $I(x,y)$



Problem Statement

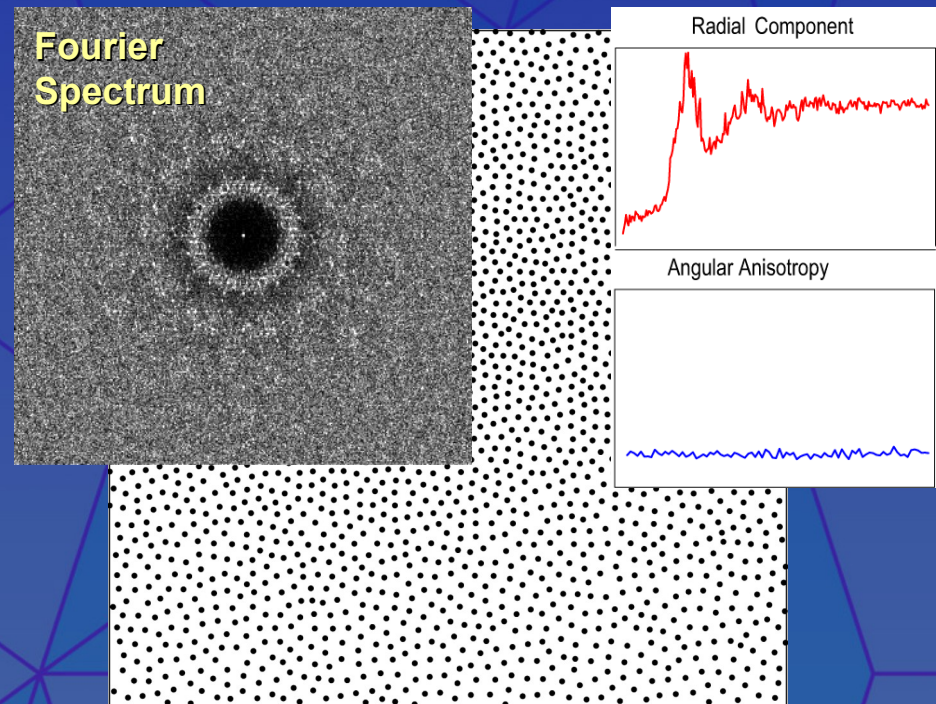
Given:

- Importance Density $I(x,y)$



Find:

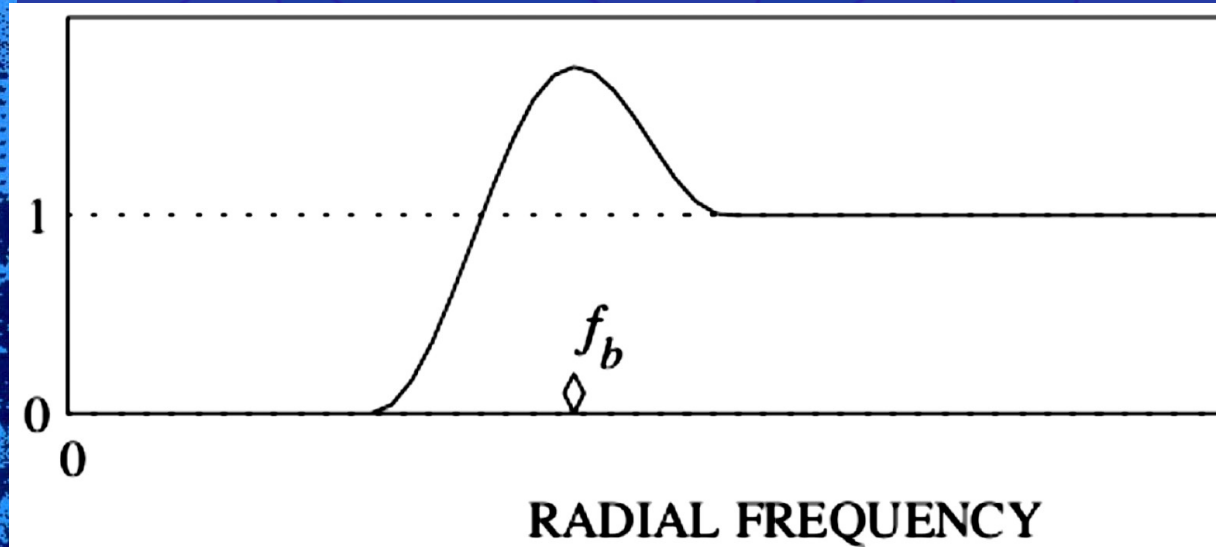
- Discrete Sample Distribution Locally Proportional to $I(x,y)$



- With Blue Noise Fourier Spectrum

Variance estimation in Monte Carlo Sampling and Blue-Noise Spectrum

Digital
Halftoning
Robert Ulichney



[Ulichney 1987]

Variance estimation in Monte Carlo Sampling and Blue-Noise Spectrum

$$V(\Delta) = \int V(\hat{\mathbf{S}}(\omega)) (\hat{f}_{\Pi}(-\omega))^2 d\omega$$

To appear in ACM TOG 32(4).

Fourier Analysis of Stochastic Sampling Strategies for Assessing Bias and Variance in Integration

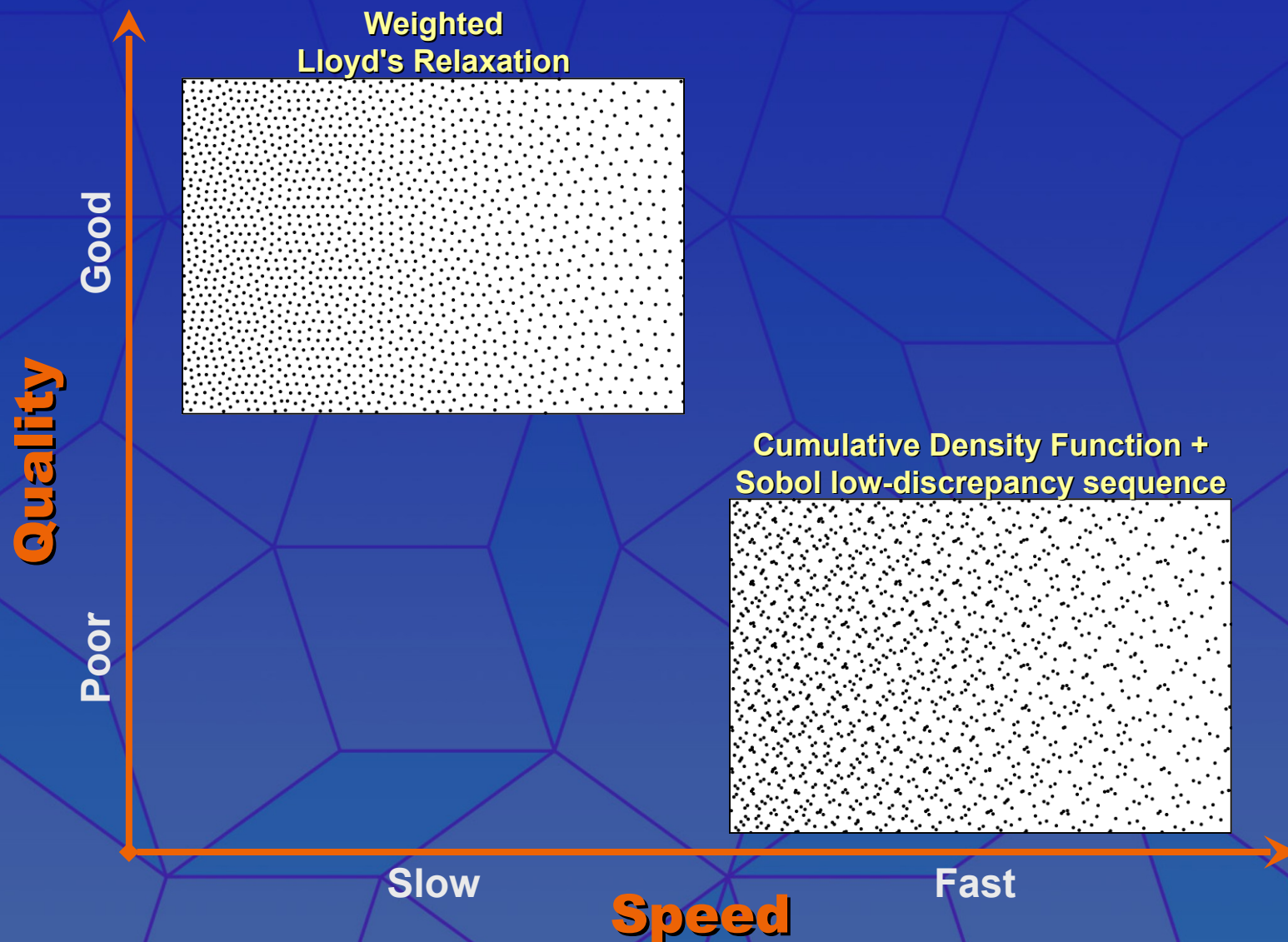
Kartic Subr*

University College London

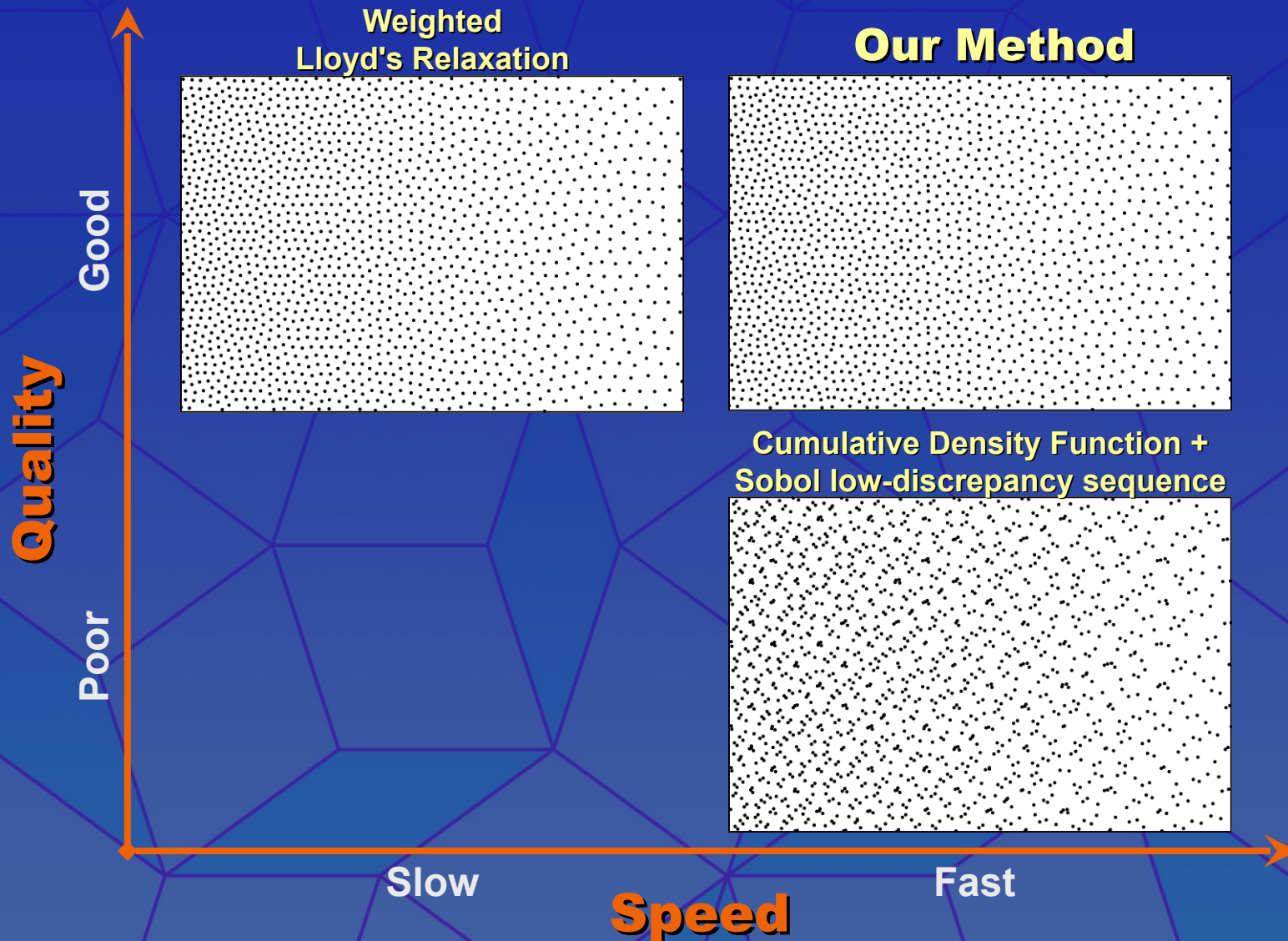
Jan Kautz[†]

University College London

Quality vs. Speed

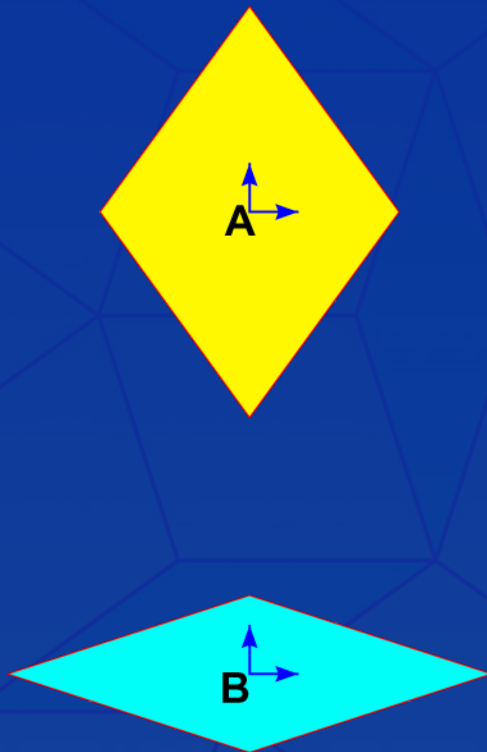


Quality vs. Speed



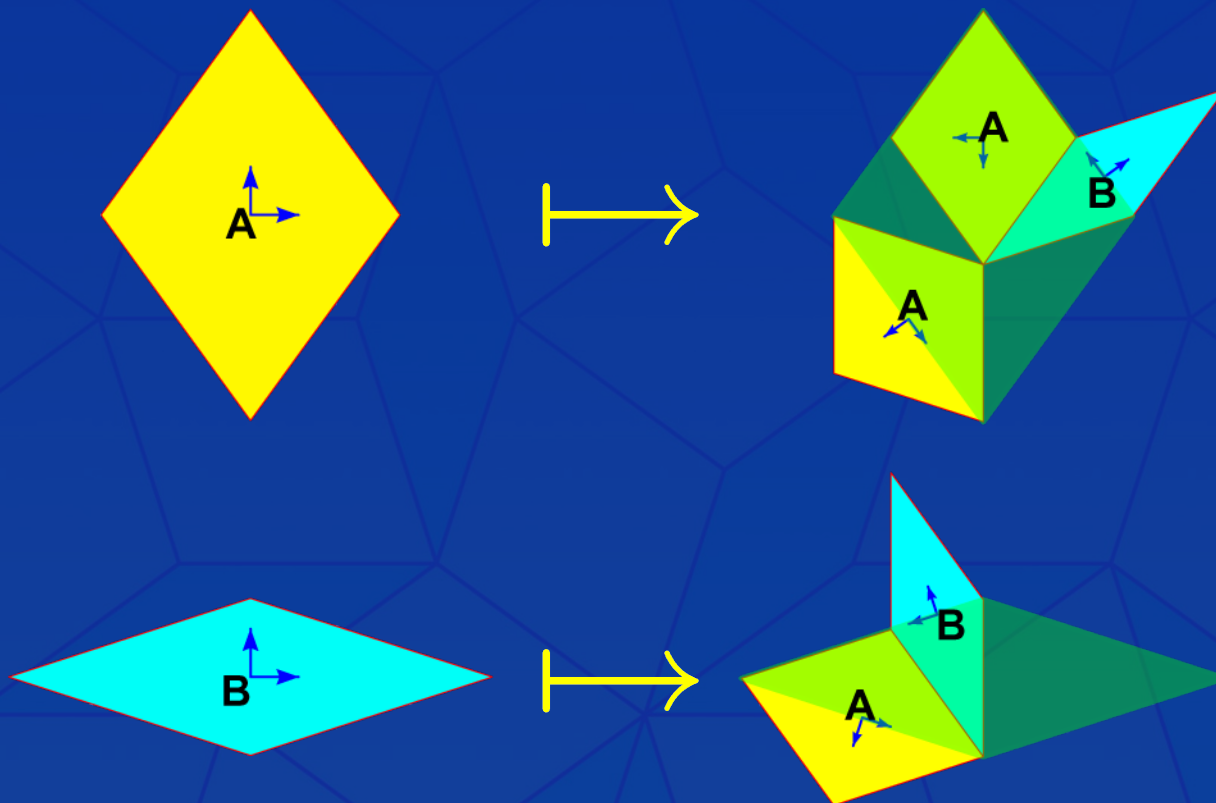
Penrose Tiling: A Primer

Penrose Tiling: Definition by Production Rules

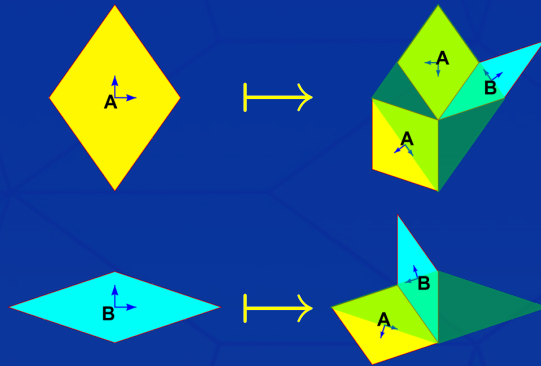


Penrose Tiling: Definition by Production Rules

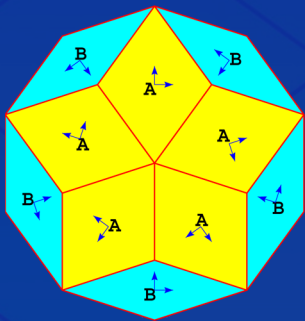
$$\mathcal{P}_{Penrose} := \begin{cases} A \mapsto \{A, B, A\} \\ B \mapsto \{A, B\} \end{cases}$$



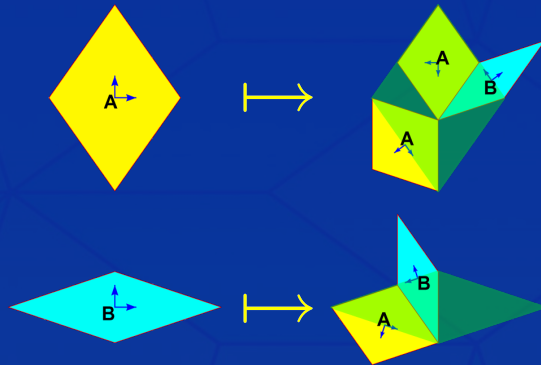
Penrose Tiling: Definition by Production Rules



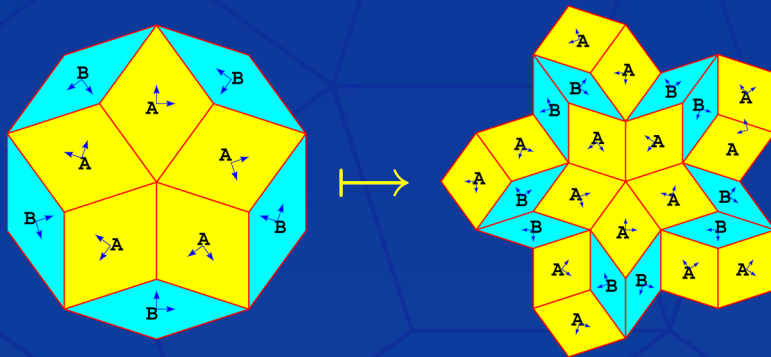
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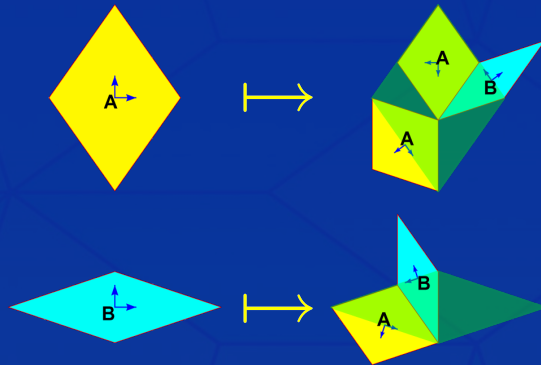
Penrose Tiling: Definition by Production Rules



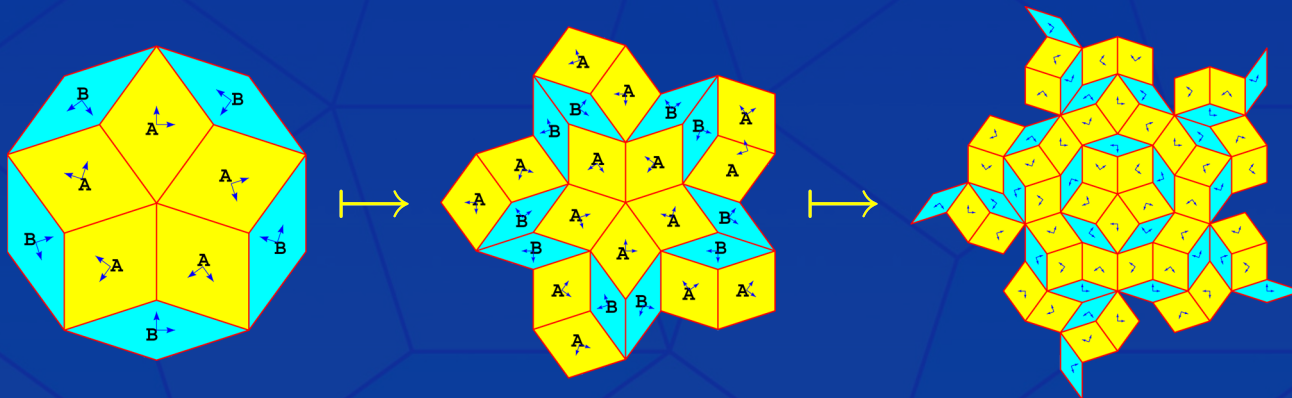
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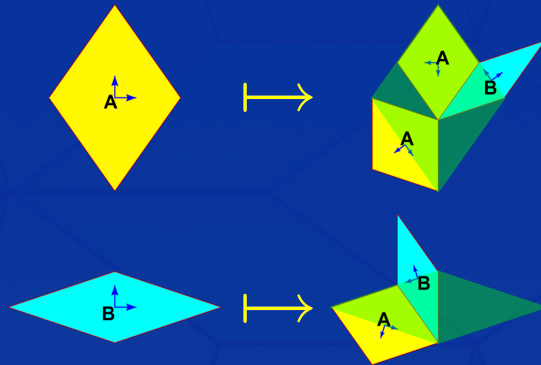
Penrose Tiling: Definition by Production Rules



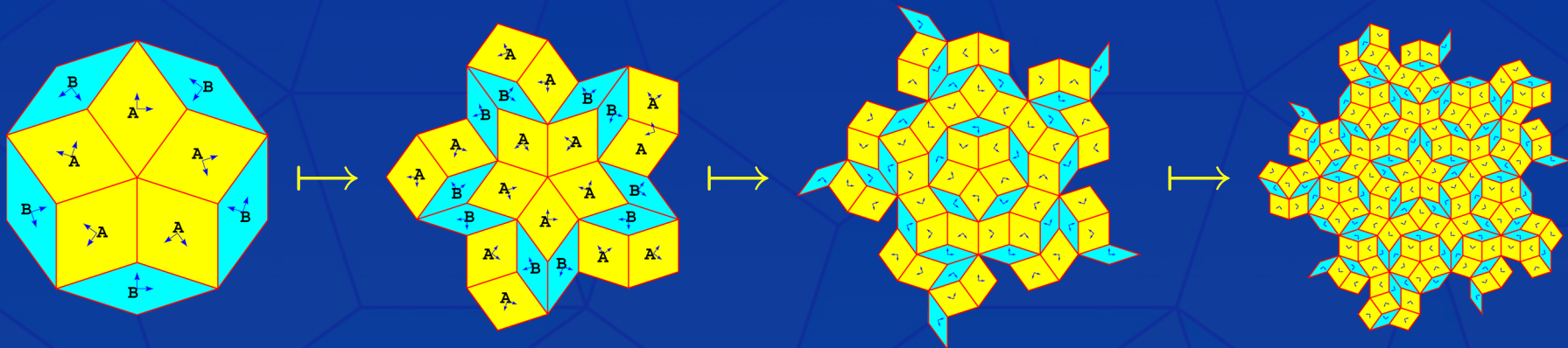
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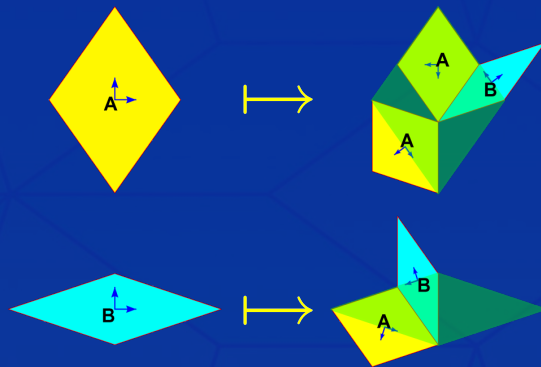
Penrose Tiling: Definition by Production Rules



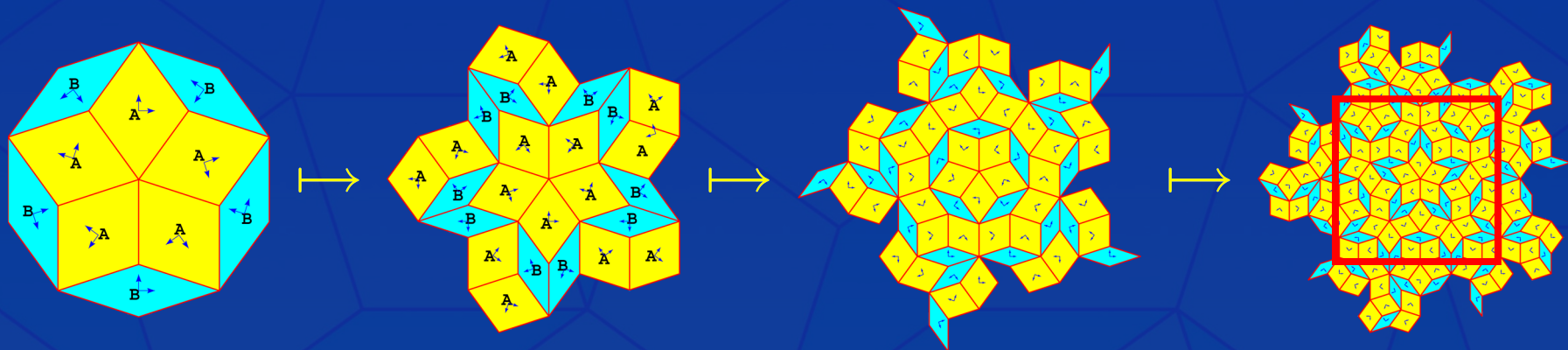
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Penrose Tiling: Definition by Production Rules

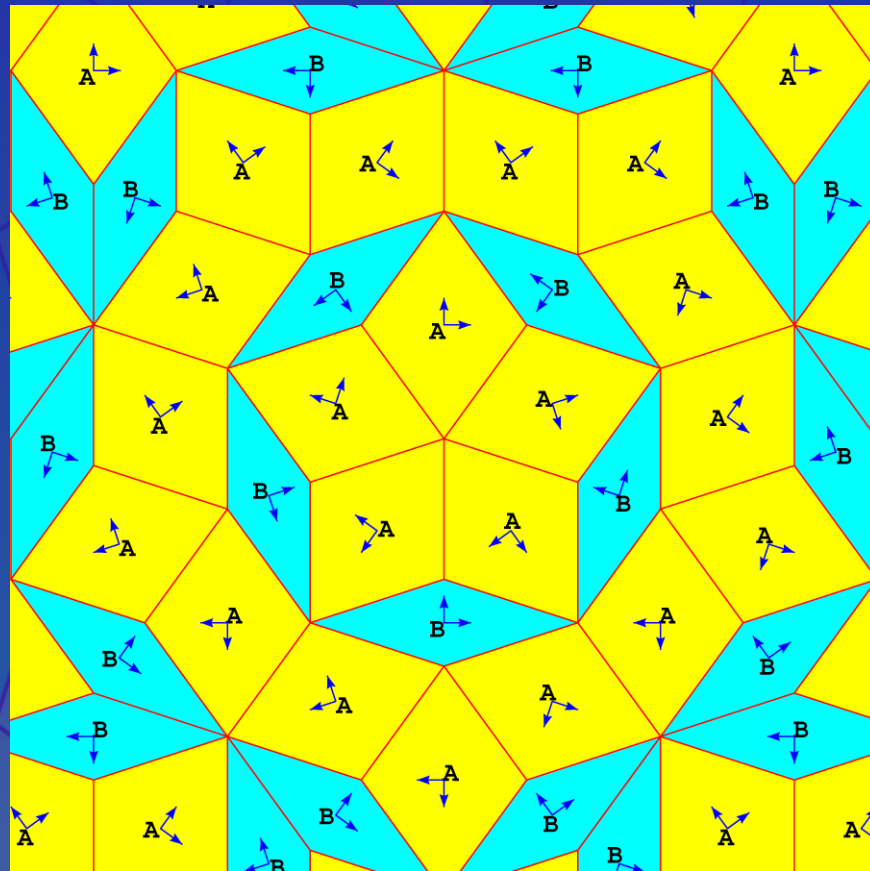


$$\mathcal{P}_{Penrose} := \begin{cases} A \mapsto \{A, B, A\} \\ B \mapsto \{A, B\} \end{cases}$$



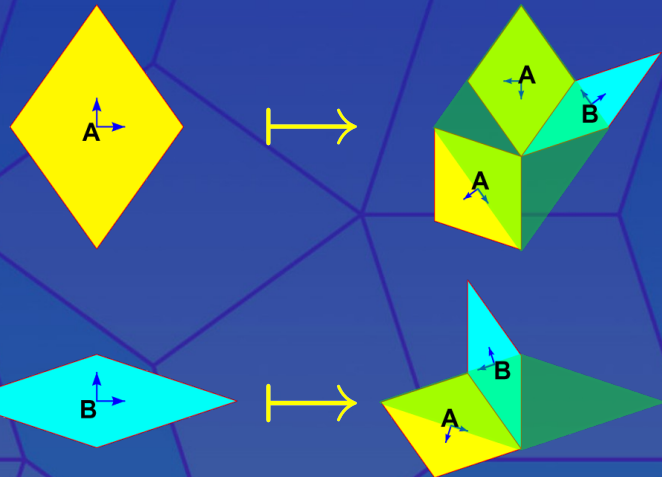
Penrose Tiling: Definition by Production Rules

Iteration N

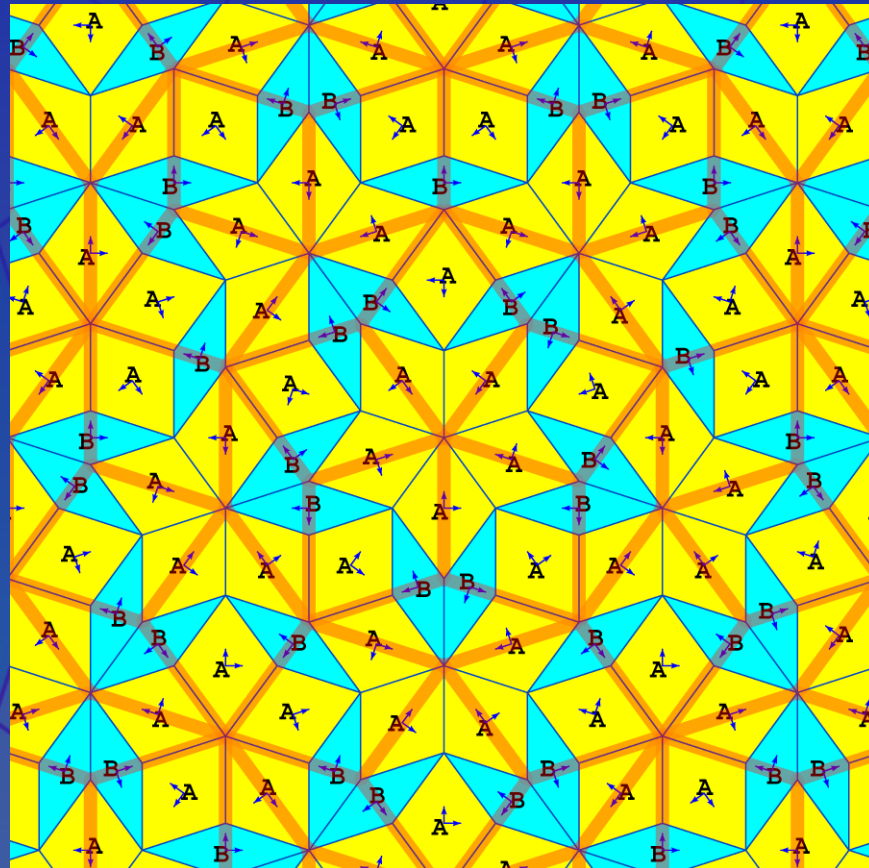


Penrose Tiling: Definition by Production Rules

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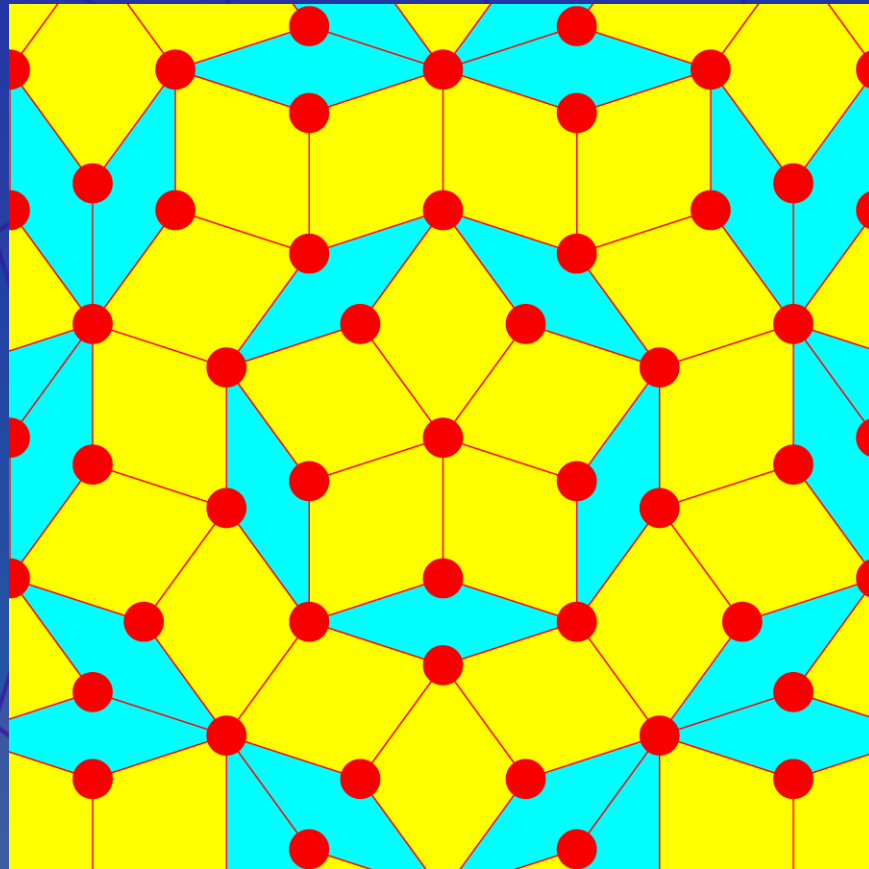
Iteration $N+1$



Ref: *"Tilings and Patterns"* by B. Grunbaum and G.C. Shephard

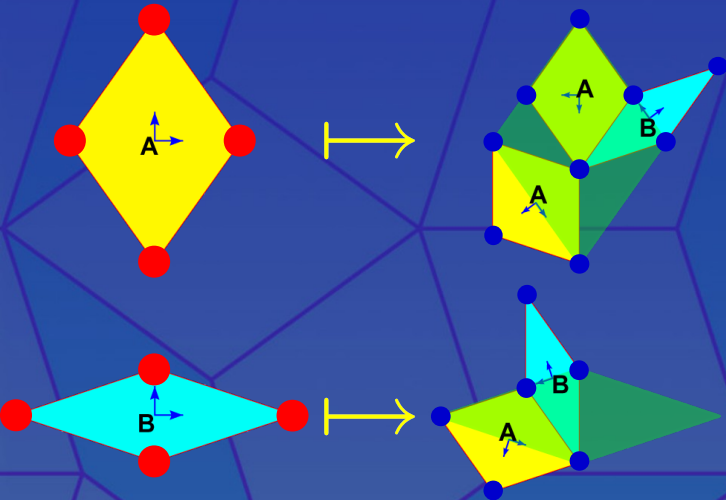
Penrose Tiling: Vertices

Iteration N

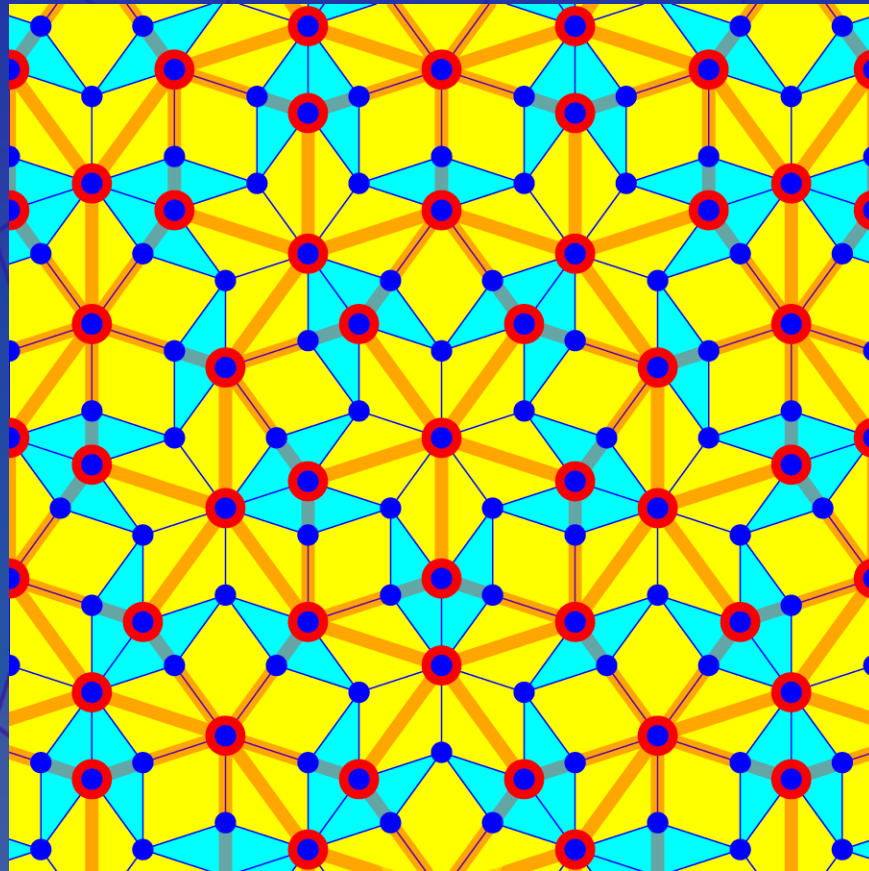


Penrose Tiling: Vertices

$$\mathcal{P}_{Penrose} := \begin{cases} A \mapsto \{A, B, A\} \\ B \mapsto \{A, B\} \end{cases}$$



Iteration $N+1$

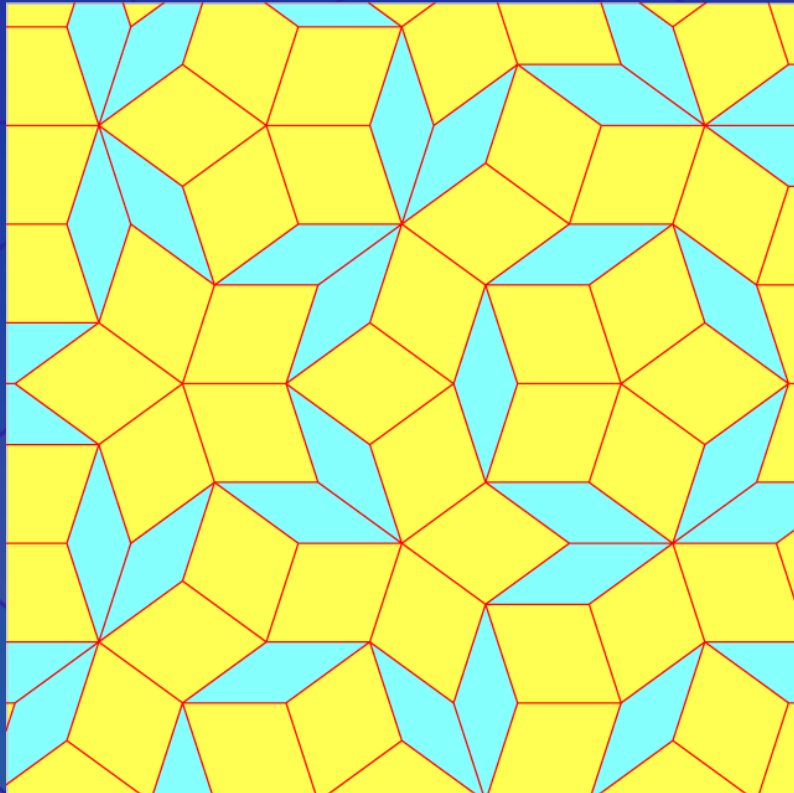


Sampling System Fundamentals

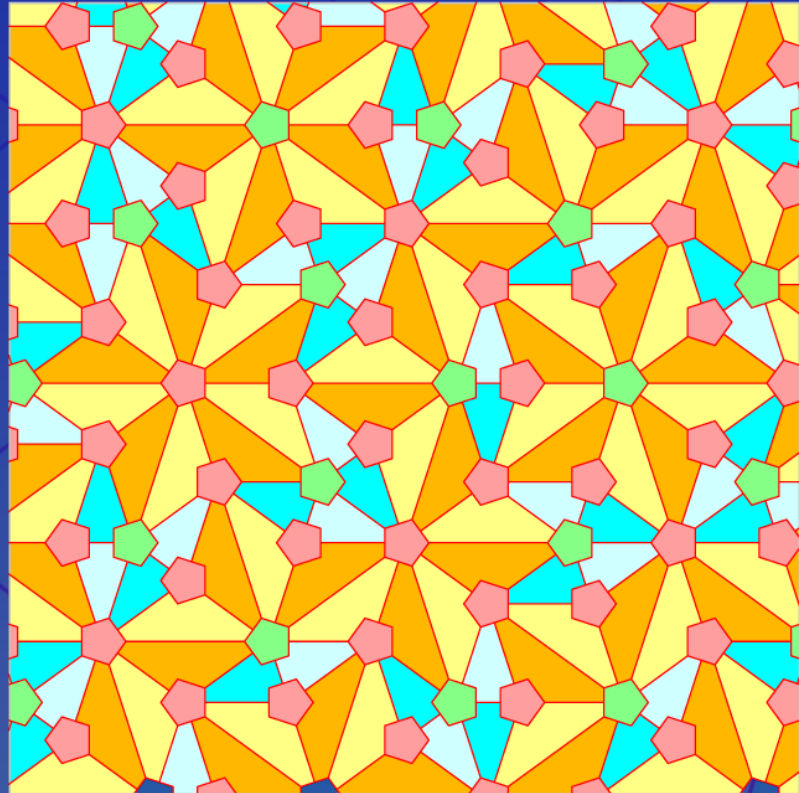
Sampling System Fundamentals

- Extension to Penrose Tiling
- Fibonacci Number System
- Adaptive Subdivision
- Corrective Vectors Lookup Table

Extension of Penrose Tiling



Original



Our Extension

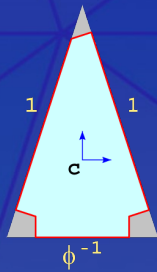
Extension of Penrose Tiling



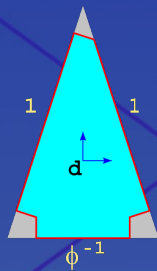
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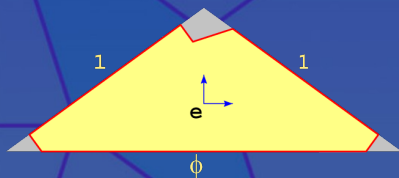
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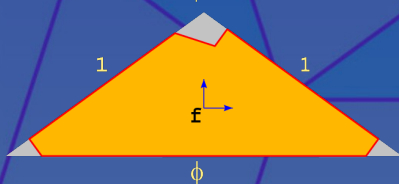
c



d

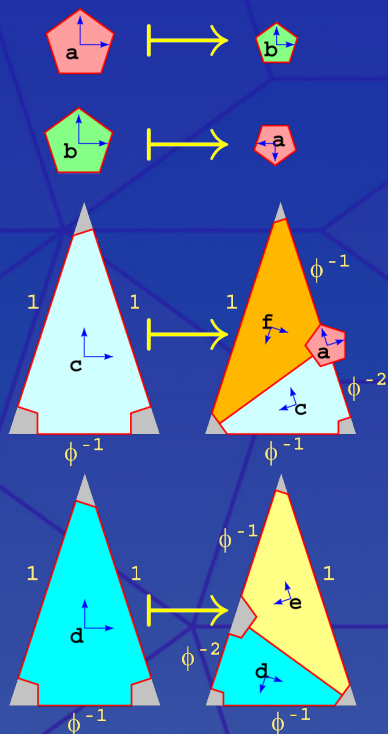


e



f

Extension of Penrose Tiling

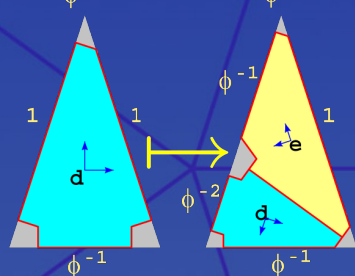
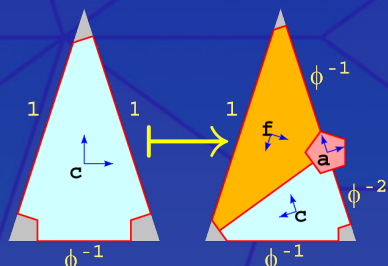
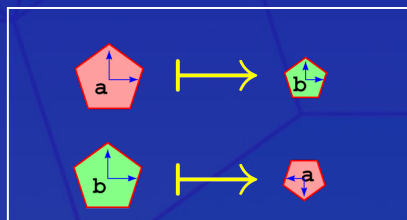


$$\mathcal{P} := \begin{cases} a \mapsto \{b\} \\ b \mapsto \{a\} \\ c \mapsto \{f, c, a\} \\ d \mapsto \{e, d\} \\ e \mapsto \{f, c, e, a\} \\ f \mapsto \{e, d, f, a\} \end{cases}$$

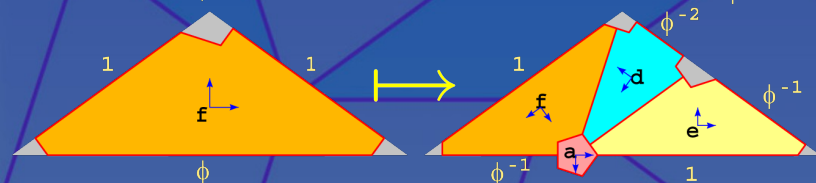
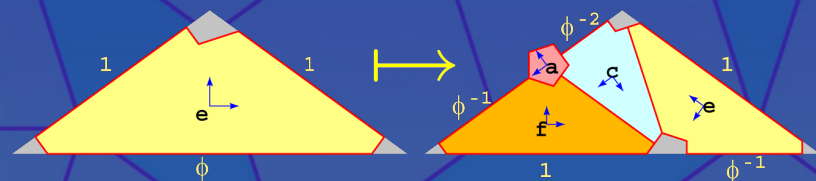
$$\frac{A_{new}}{A_{old}} = \phi^2$$

$$\phi = \frac{1+\sqrt{5}}{2} \text{ (Golden Ratio)}$$

Extension of Penrose Tiling



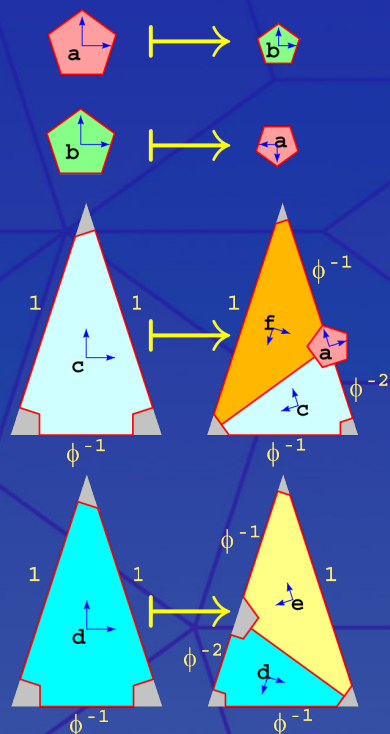
$$\mathcal{P} := \begin{cases} a \mapsto \{b\} \\ b \mapsto \{a\} \\ c \mapsto \{f, c, a\} \\ d \mapsto \{e, d\} \\ e \mapsto \{f, c, e, a\} \\ f \mapsto \{e, d, f, a\} \end{cases}$$



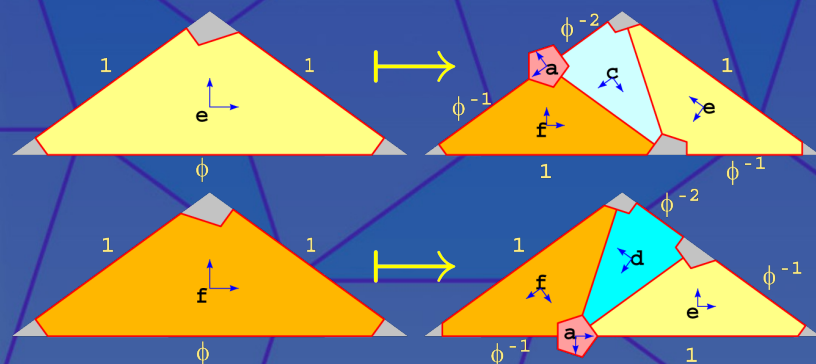
$$\frac{A_{new}}{A_{old}} = \phi^2$$

$$\phi = \frac{1+\sqrt{5}}{2} \text{ (Golden Ratio)}$$

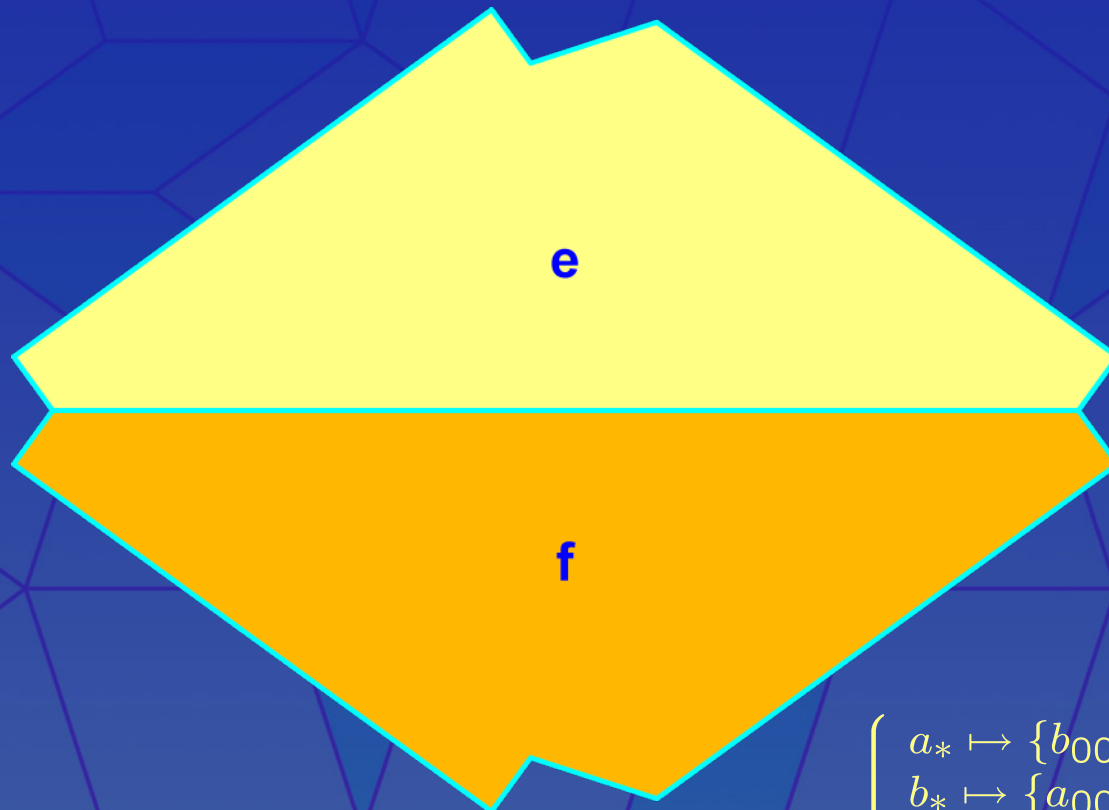
Fibonacci Number System



$$\mathcal{P} := \left\{ \begin{array}{l} a_* \mapsto \{b_{00*}\} \\ b_* \mapsto \{a_{00*}\} \\ c_* \mapsto \{f_{00*}, c_{10*}, a_{10*}\} \\ d_* \mapsto \{e_{00*}, d_{10*}\} \\ e_* \mapsto \{f_{00*}, c_{10*}, e_{01*}, a_{10*}\} \\ f_* \mapsto \{e_{00*}, d_{10*}, f_{01*}, a_{01*}\} \end{array} \right.$$

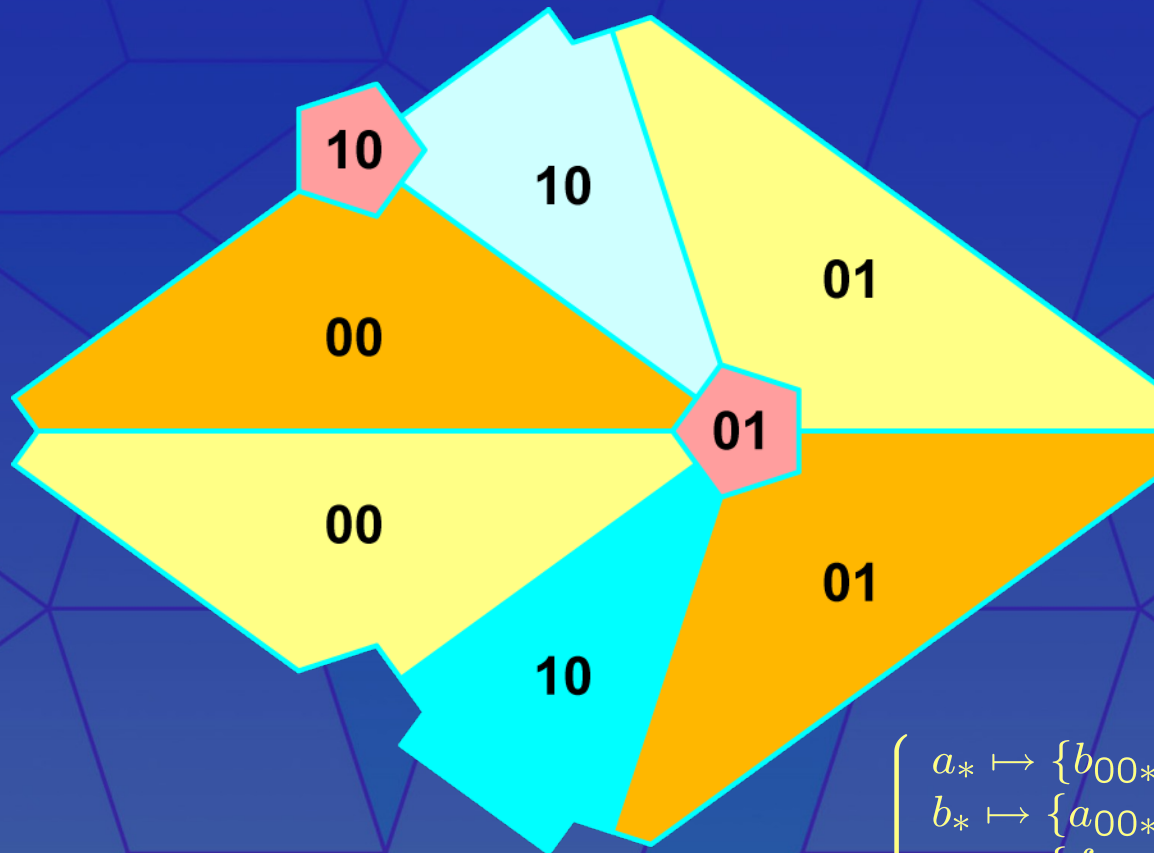


Fibonacci Number System



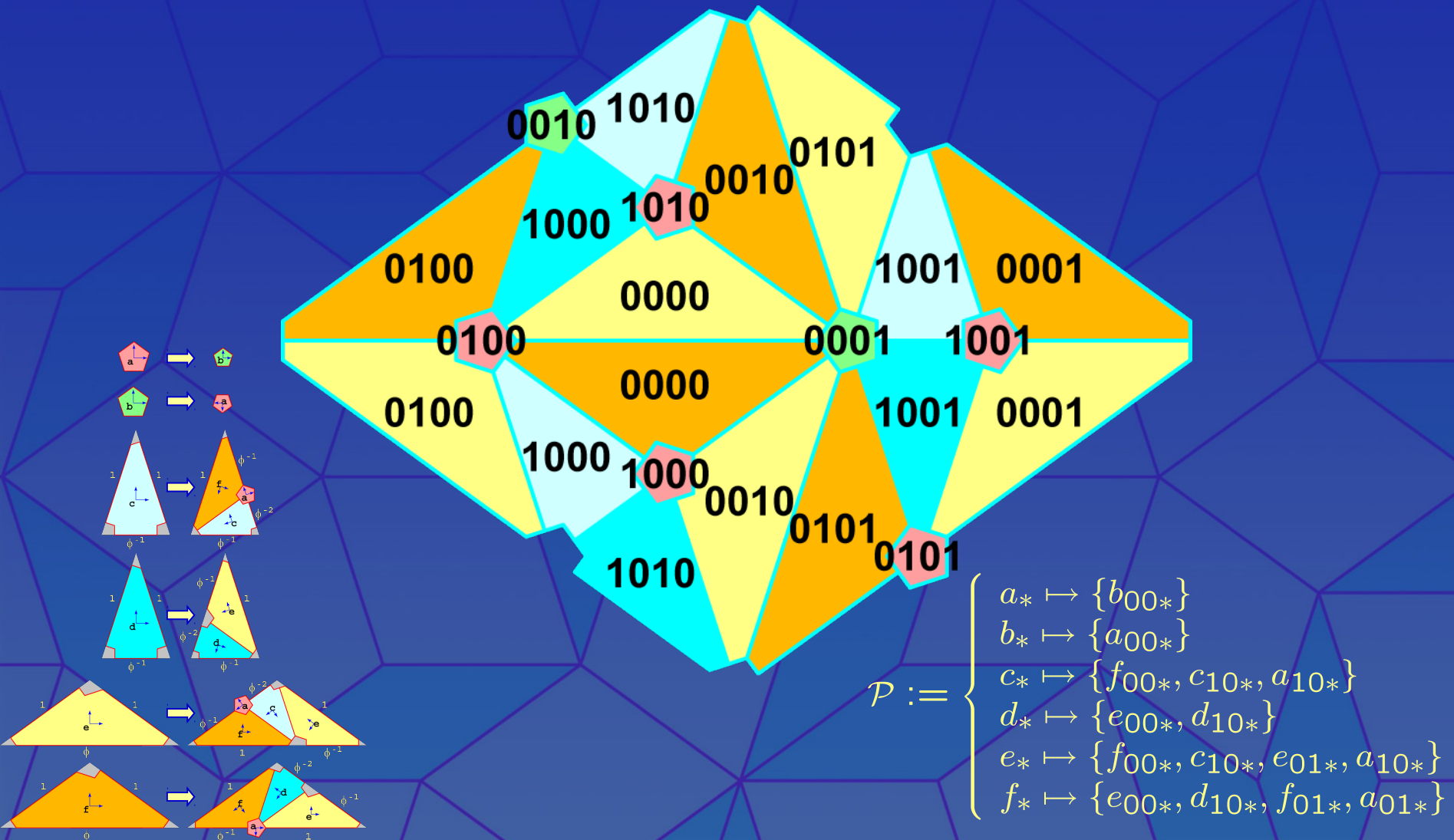
$$\mathcal{P} := \begin{cases} a_* \mapsto \{b_{00*}\} \\ b_* \mapsto \{a_{00*}\} \\ c_* \mapsto \{f_{00*}, c_{10*}, a_{10*}\} \\ d_* \mapsto \{e_{00*}, d_{10*}\} \\ e_* \mapsto \{f_{00*}, c_{10*}, e_{01*}, a_{10*}\} \\ f_* \mapsto \{e_{00*}, d_{10*}, f_{01*}, a_{01*}\} \end{cases}$$

Fibonacci Number System



$$\mathcal{P} := \begin{cases} a_* \mapsto \{b_{00*}\} \\ b_* \mapsto \{a_{00*}\} \\ c_* \mapsto \{f_{00*}, c_{10*}, a_{10*}\} \\ d_* \mapsto \{e_{00*}, d_{10*}\} \\ e_* \mapsto \{f_{00*}, c_{10*}, e_{01*}, a_{10*}\} \\ f_* \mapsto \{e_{00*}, d_{10*}, f_{01*}, a_{01*}\} \end{cases}$$

Fibonacci Number System



Fibonacci Number System

Ref: “*The Art of Computer Programming, Vol. 1,*” by D.E. Knuth

- Binary Number System:

$$n = (a_m a_{m-1} \dots a_1 a_0)_2 \iff n = \sum_{j=0}^m a_j 2^j.$$

- Fibonacci Number System:

$$n = (b_m b_{m-1} \dots b_3 b_2)_F \iff n = \sum_{j=2}^m b_j F_j.$$

- Fibonacci Numbers:

$$F_i = F_{i-1} + F_{i-2} \quad \{0, 1, 1, 2, 3, 5, 8, 13, 21, 34, \dots\}$$

Fibonacci Number System

Binary Number System

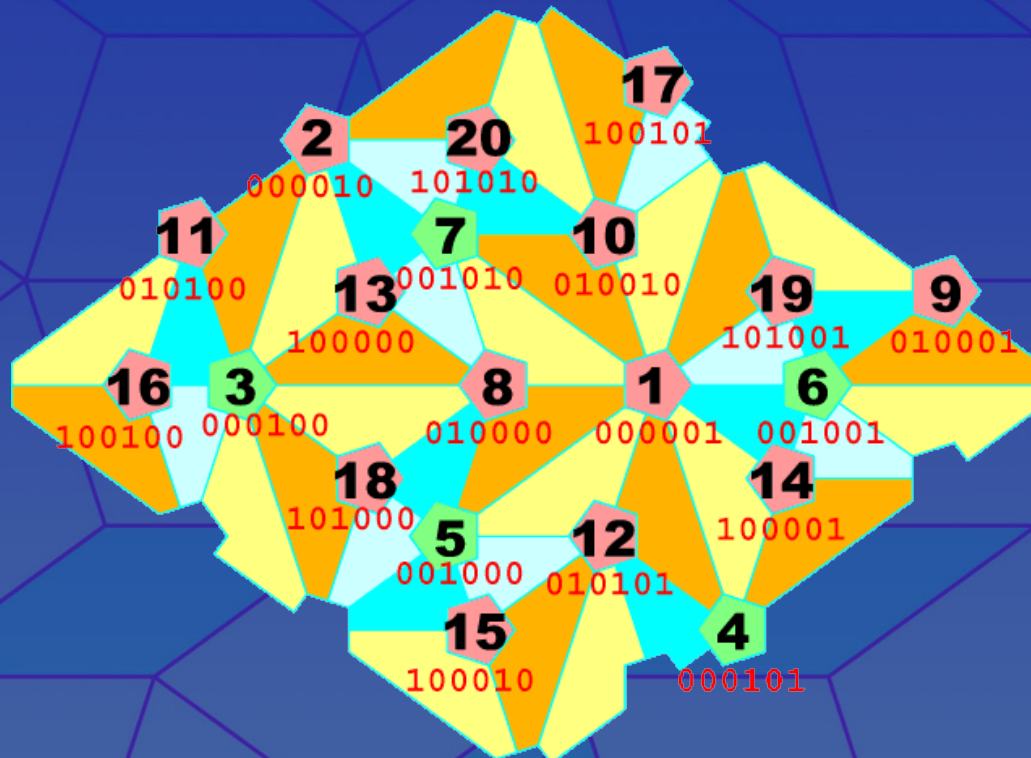
N_{10}	2^3 (8)	2^2 (4)	2^1 (2)	2^0 (1)
0	0	0	0	0
1	0	0	0	1
2	0	0	1	0
3	0	0	1	1
4	0	1	0	0
5	0	1	0	1
6	0	1	1	0
7	0	1	1	1
8	1	0	0	0
9	1	0	0	1
10	1	0	1	0
11	1	0	1	1
12	1	1	0	0

Fibonacci Number System

N_{10}	F_6 (8)	F_5 (5)	F_4 (3)	F_3 (2)	F_2 (1)
0	0	0	0	0	0
1	0	0	0	0	1
2	0	0	0	1	0
3	0	0	1	0	0
4	0	0	1	0	1
5	0	1	0	0	0
6	0	1	0	0	1
7	0	1	0	1	0
8	1	0	0	0	0
9	1	0	0	0	1
10	1	0	0	1	0
11	1	0	1	0	0
12	1	0	1	0	1

Fibonacci Number System

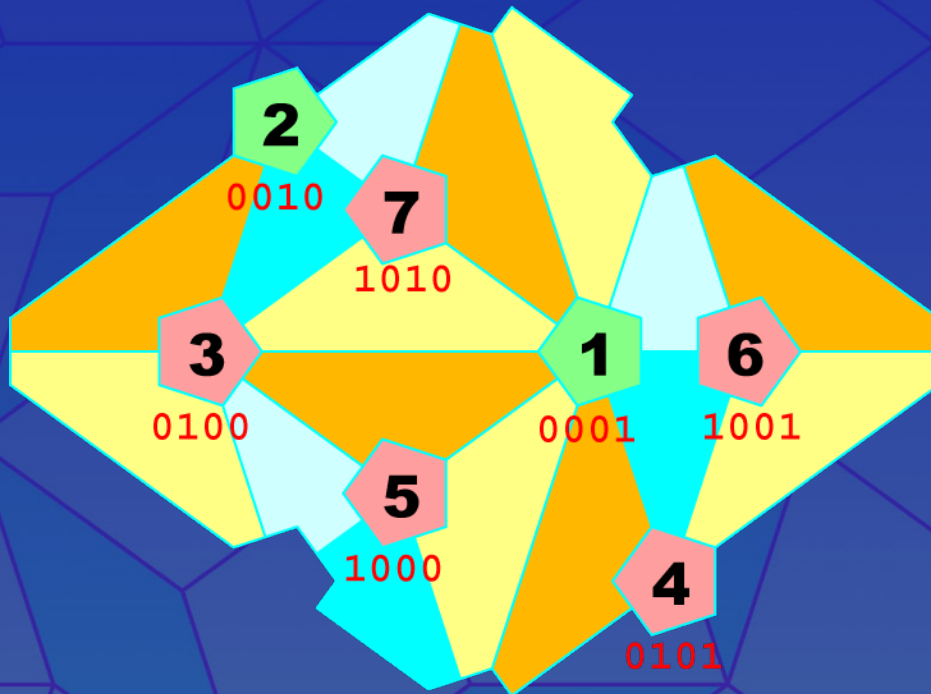
$(00001)_F = 1$, $(00010)_F = 2$, $(00100)_F = 3$,
 $(00101)_F = 4$, $(01000)_F = 5$, $(01001)_F = 6$,
 $(01010)_F = 7$, $(10000)_F = 8$, $(10001)_F = 9$,
 $(10010)_F = 10$, $(10100)_F = 11$, $(10101)_F = 12$.



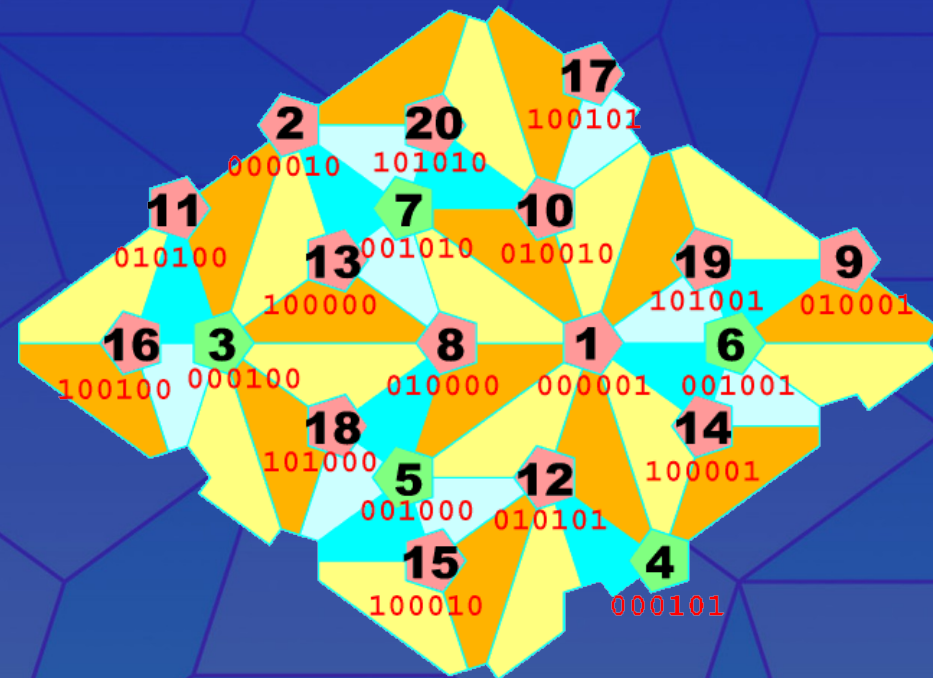
Pentagonal
Tiles Only

Fibonacci Number System

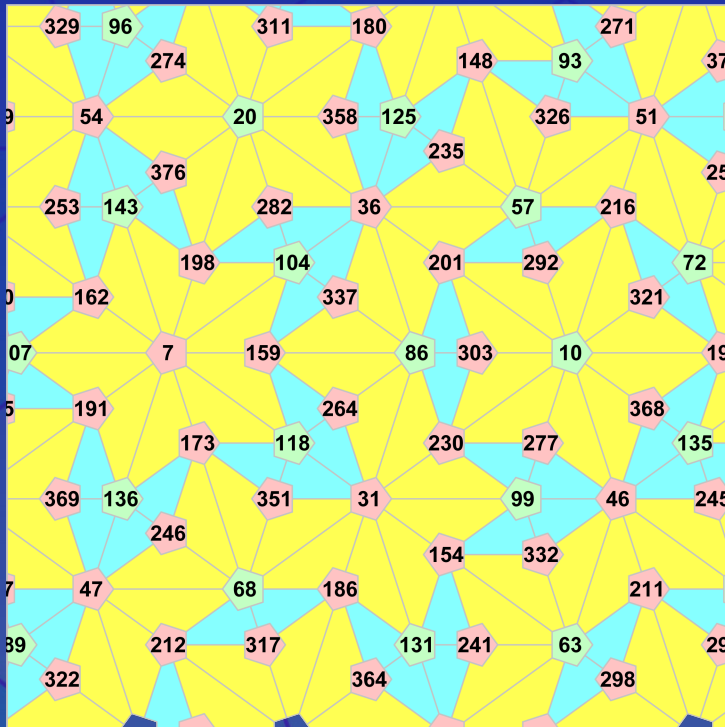
Iteration 4



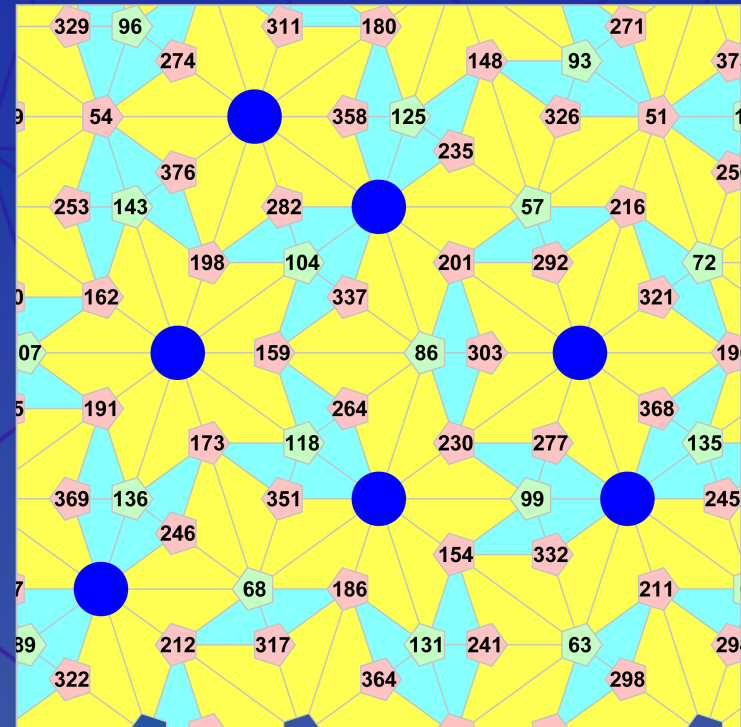
Iteration 5



Fibonacci Number System Thresholding

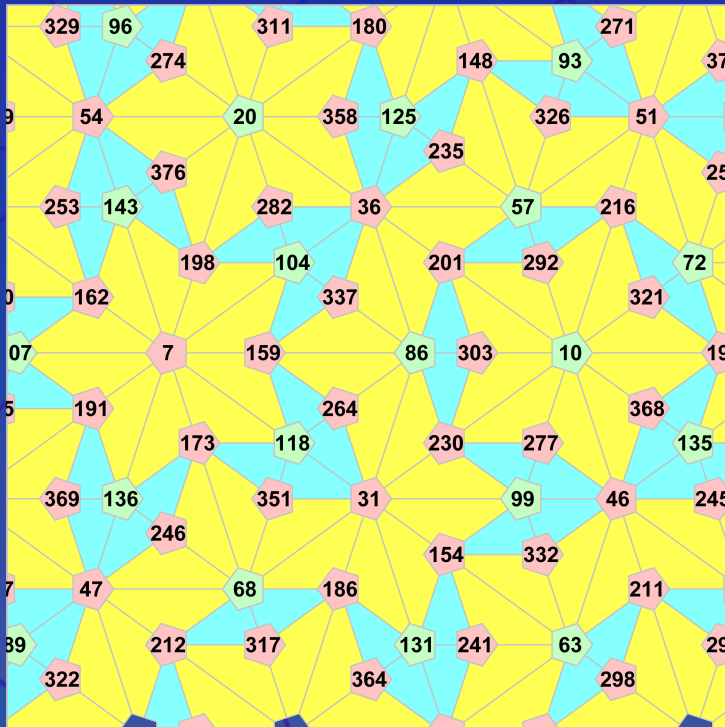


F-codes, in range [1..376]

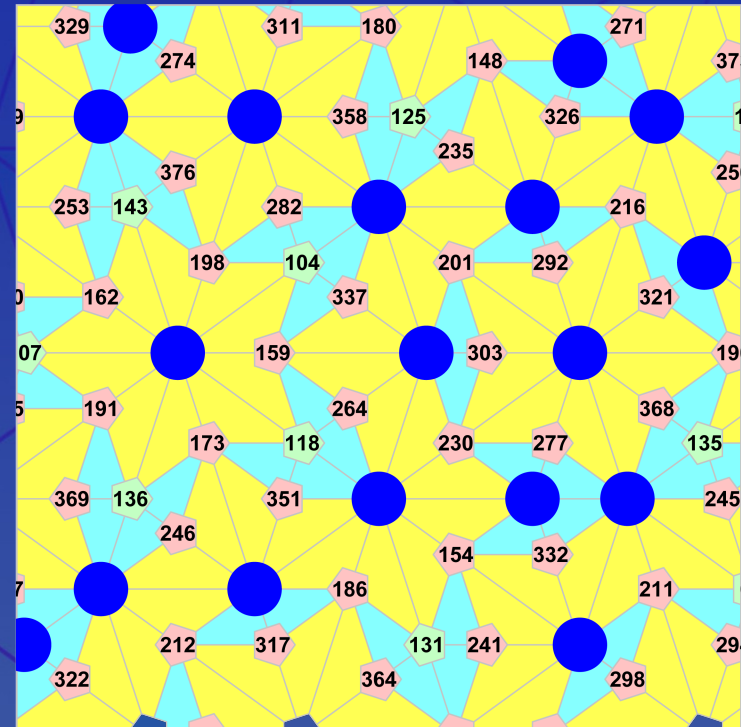


Thresholding at value = 50

Fibonacci Number System Thresholding

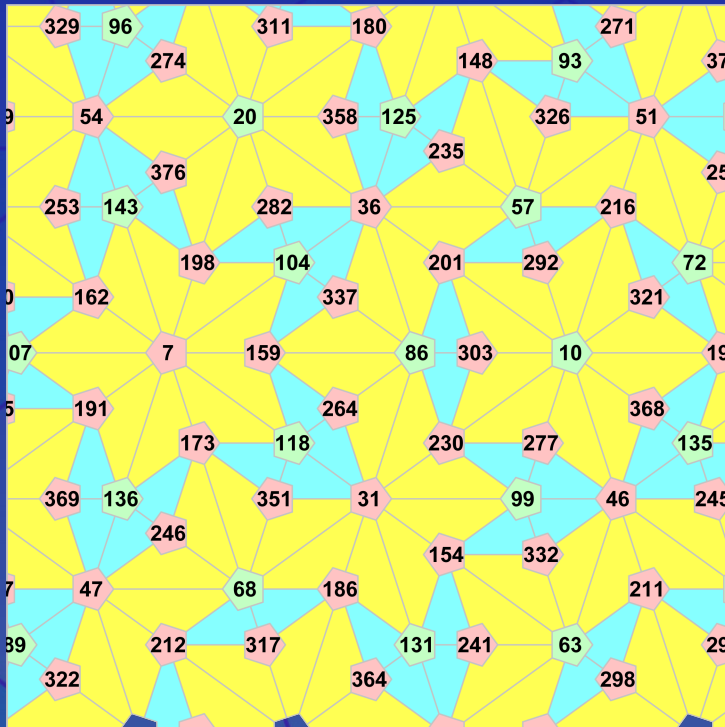


F-codes, in range [1..376]

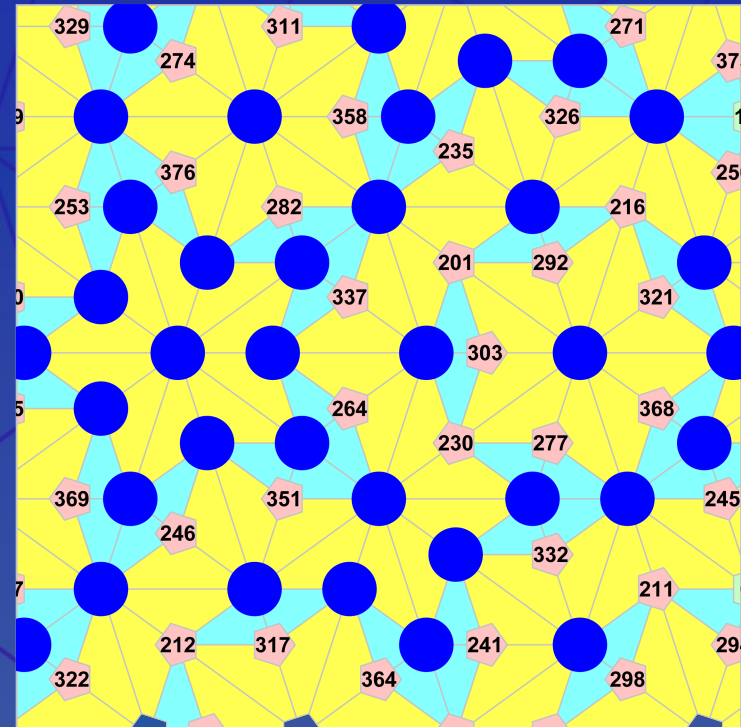


Thresholding at value = 100

Fibonacci Number System Thresholding



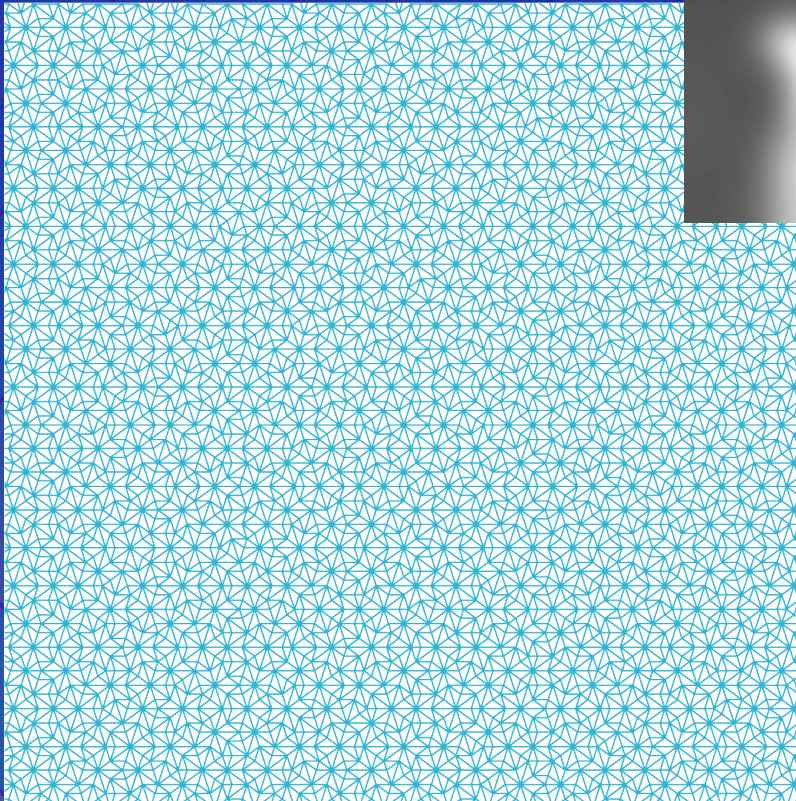
F-codes, in range [1..376]



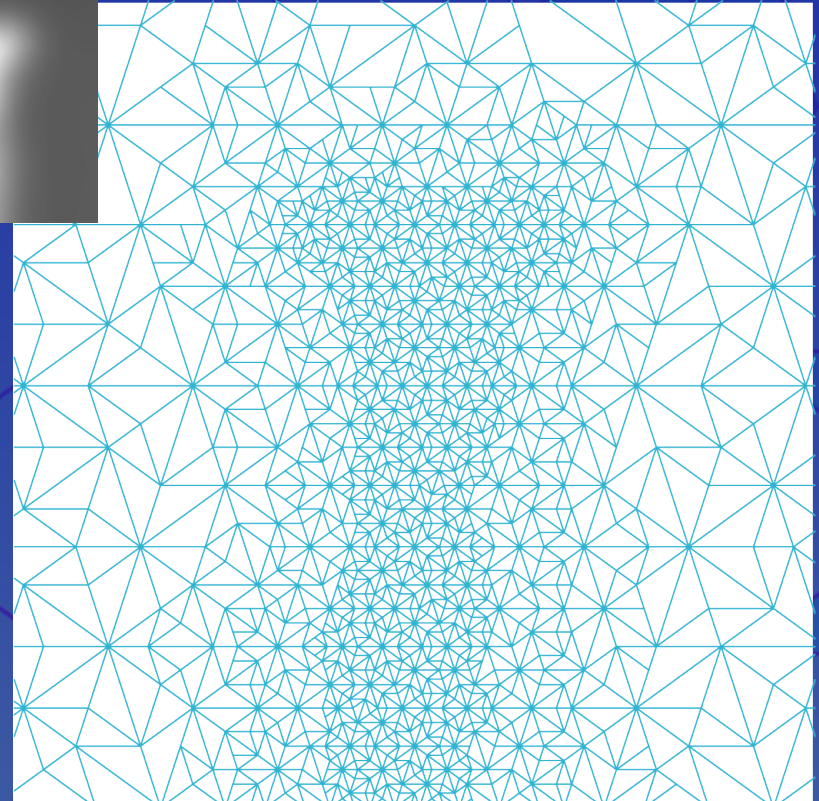
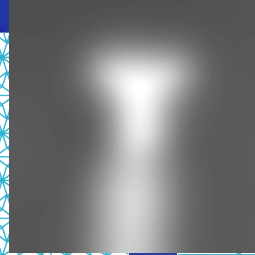
Thresholding at value = 200

Adaptive Subdivision

*Importance
Density Function*



Non-adaptive



Adaptive

Adaptive Subdivision

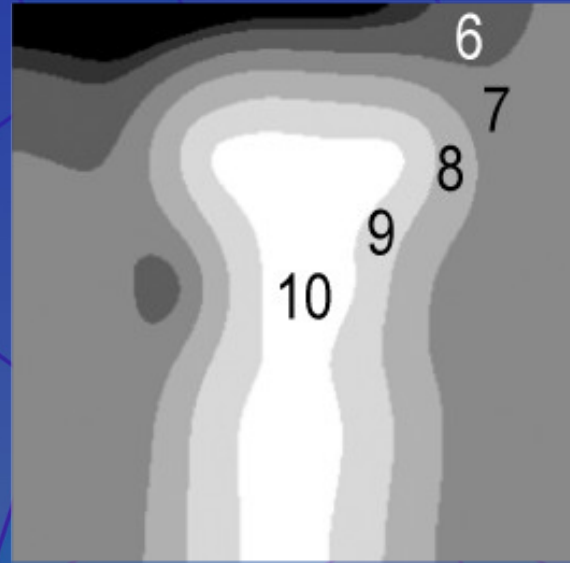
$$\lambda(x) = \log_{\phi^2}(x) + C$$

$$\kappa = \lceil \max_{tile} \lambda(mag \cdot I(x, y)) \rceil$$



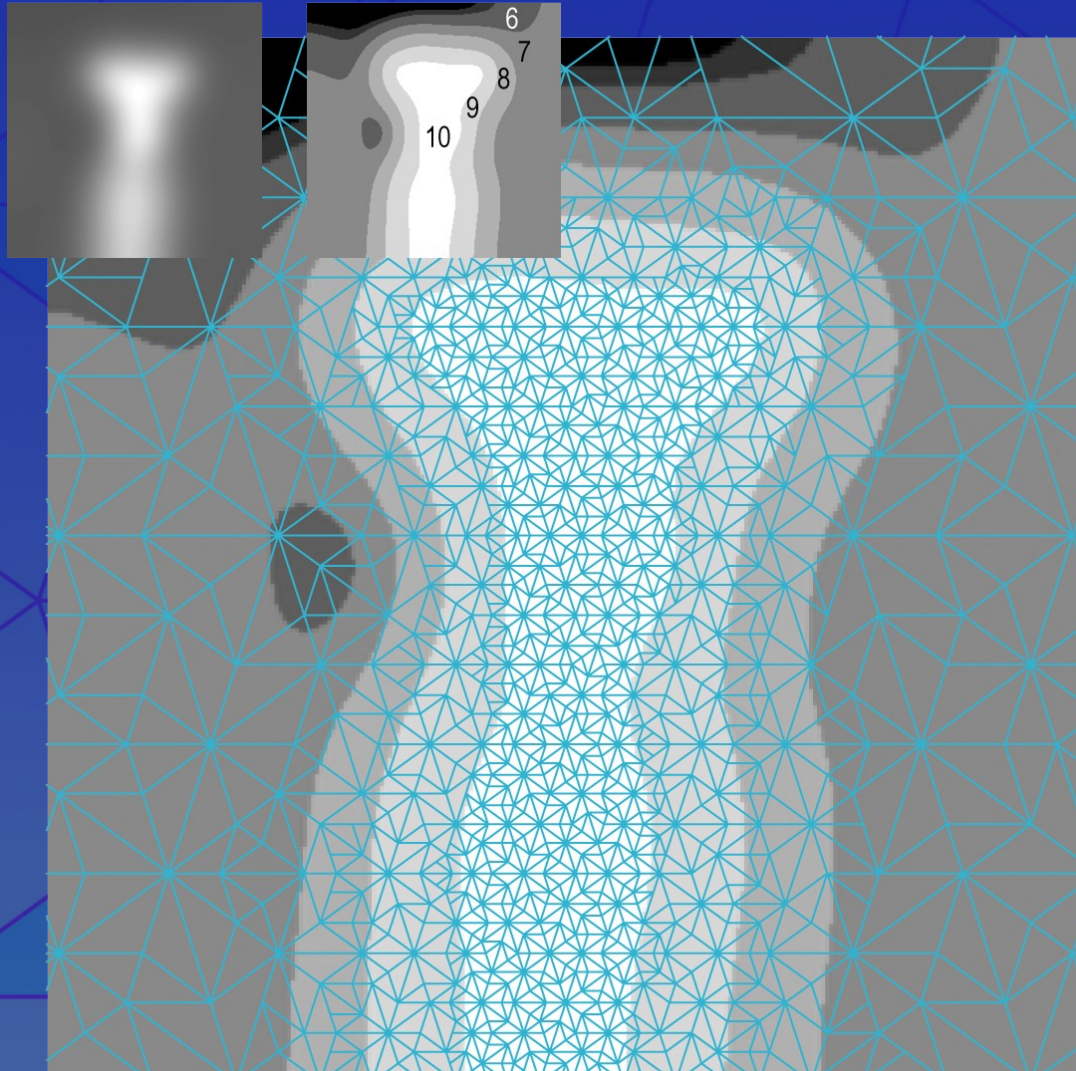
Importance Density Function

$I(x, y)$

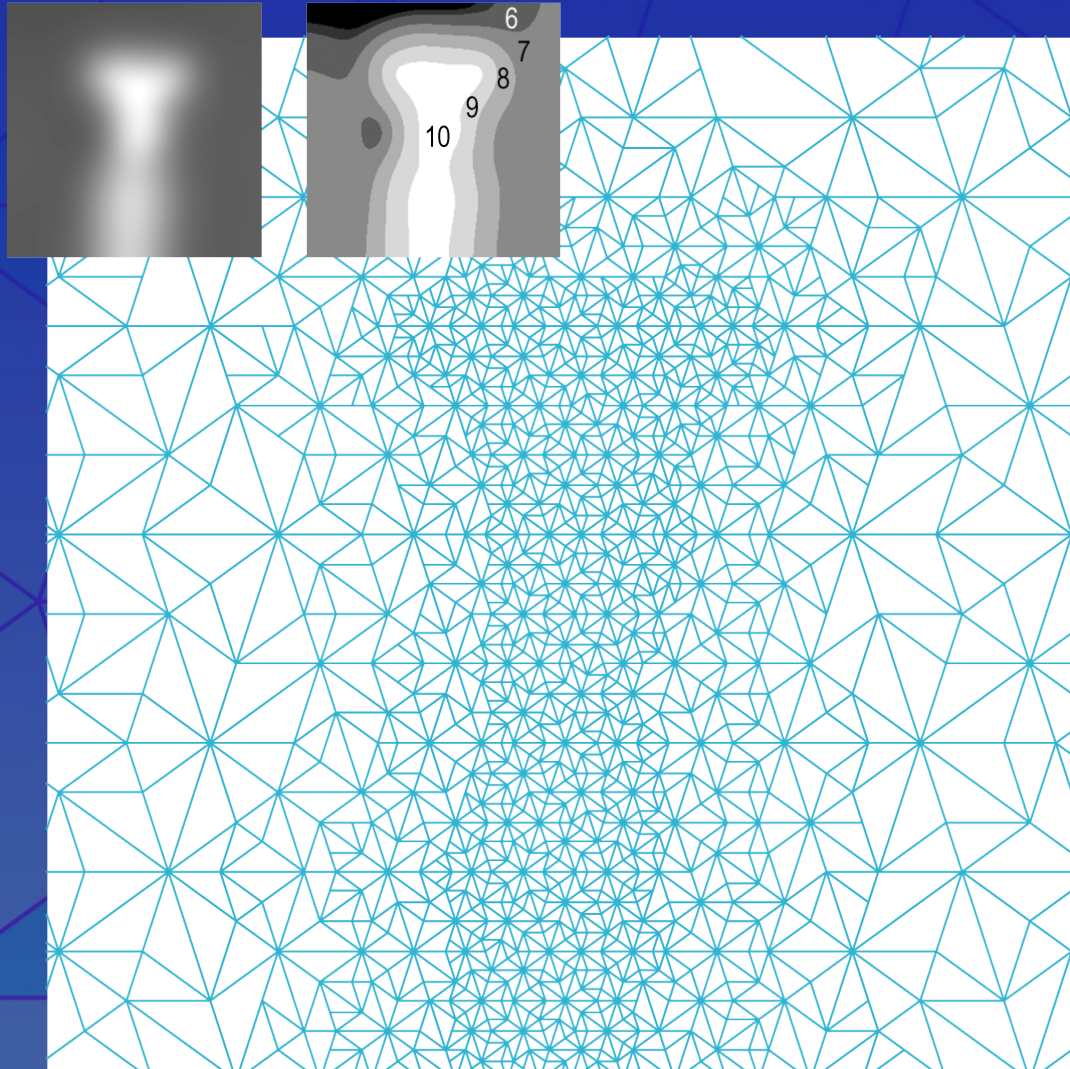


$\lceil \lambda(I(x, y)) \rceil$

Adaptive Subdivision

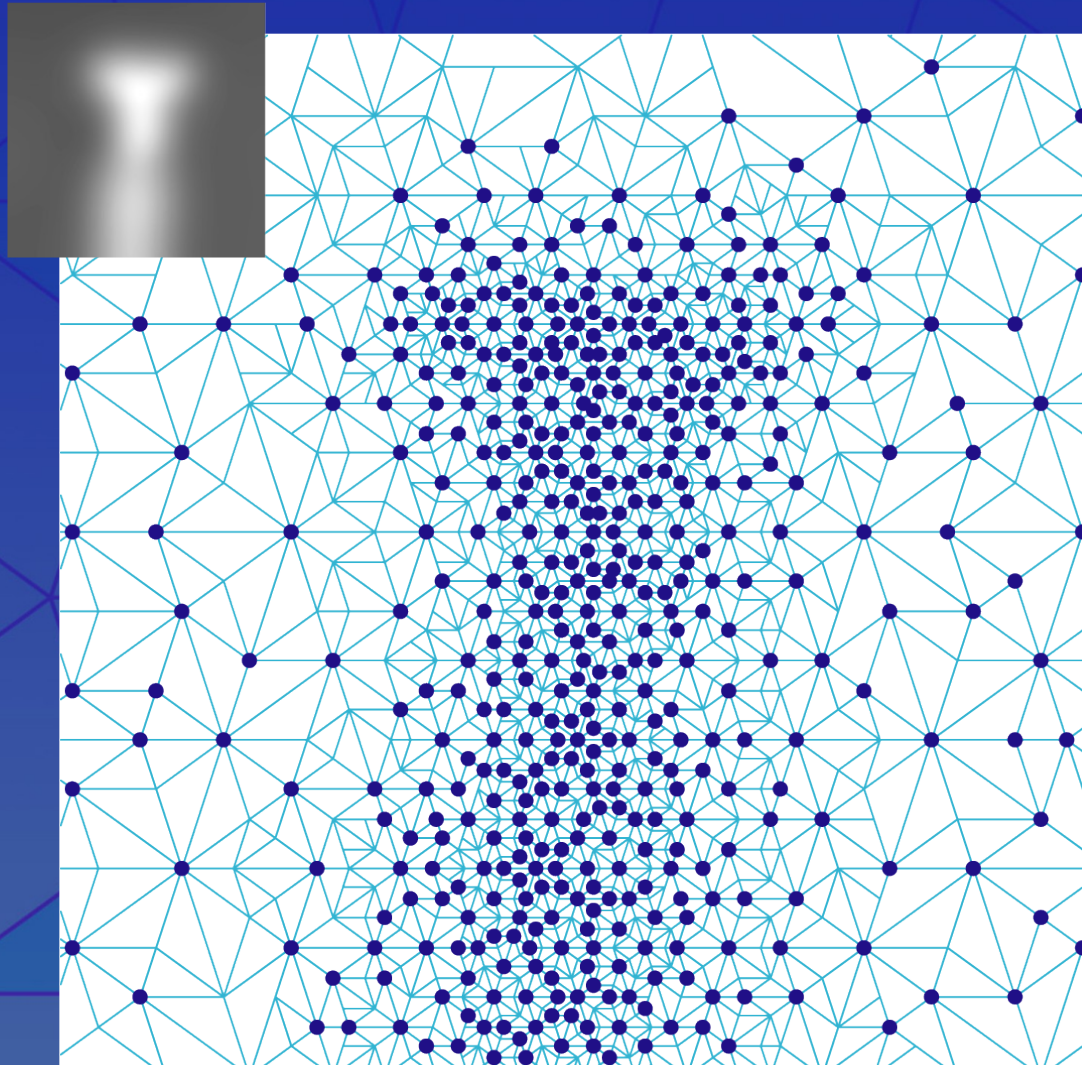


Adaptive Subdivision



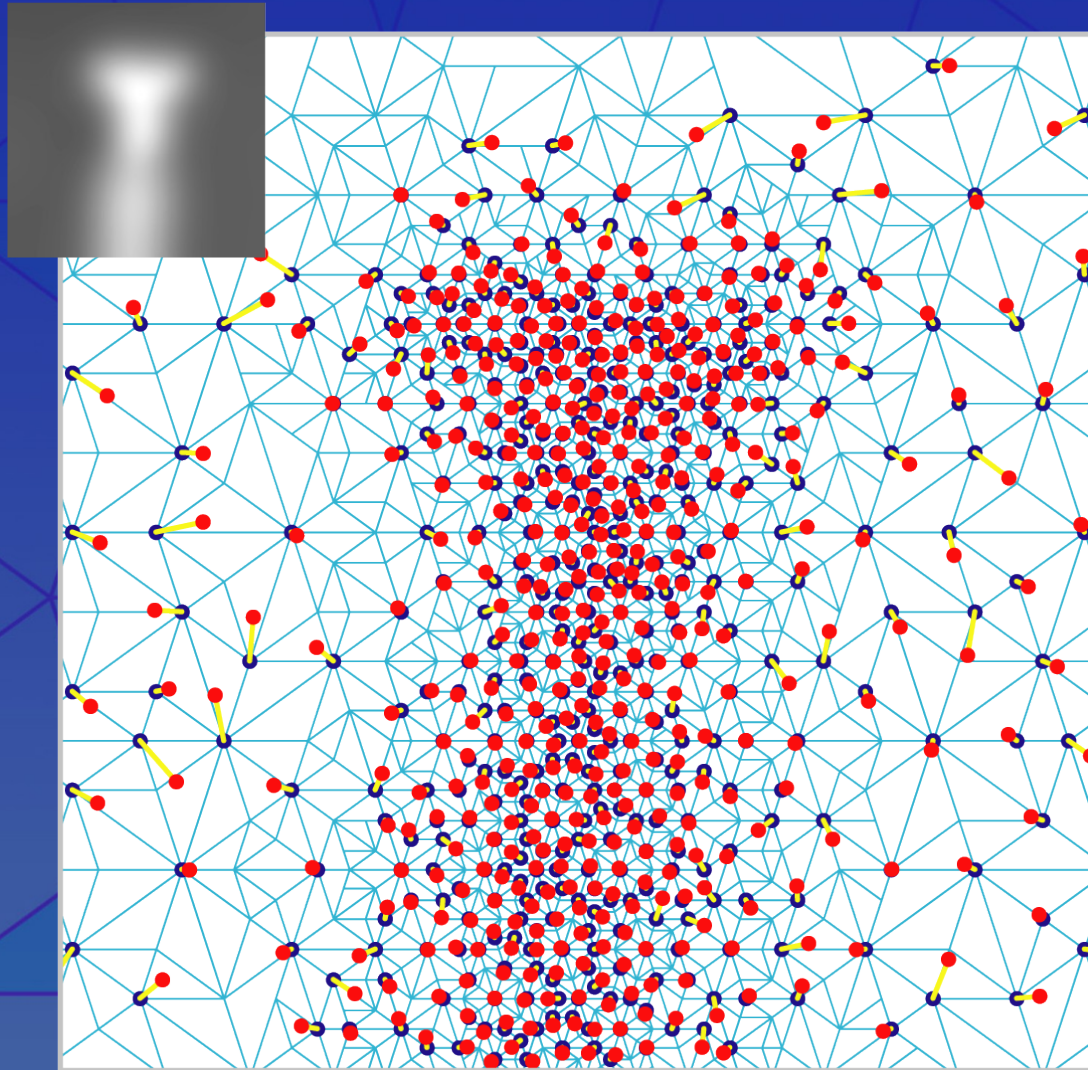
Corrective Vectors

Lookup table



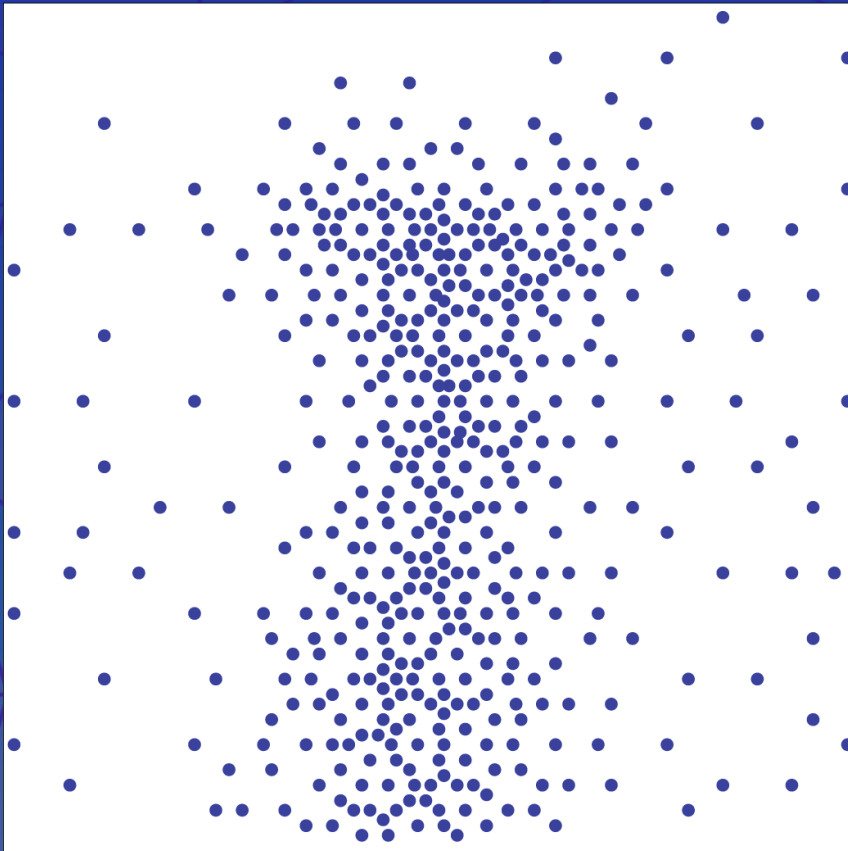
Corrective Vectors

Lookup table

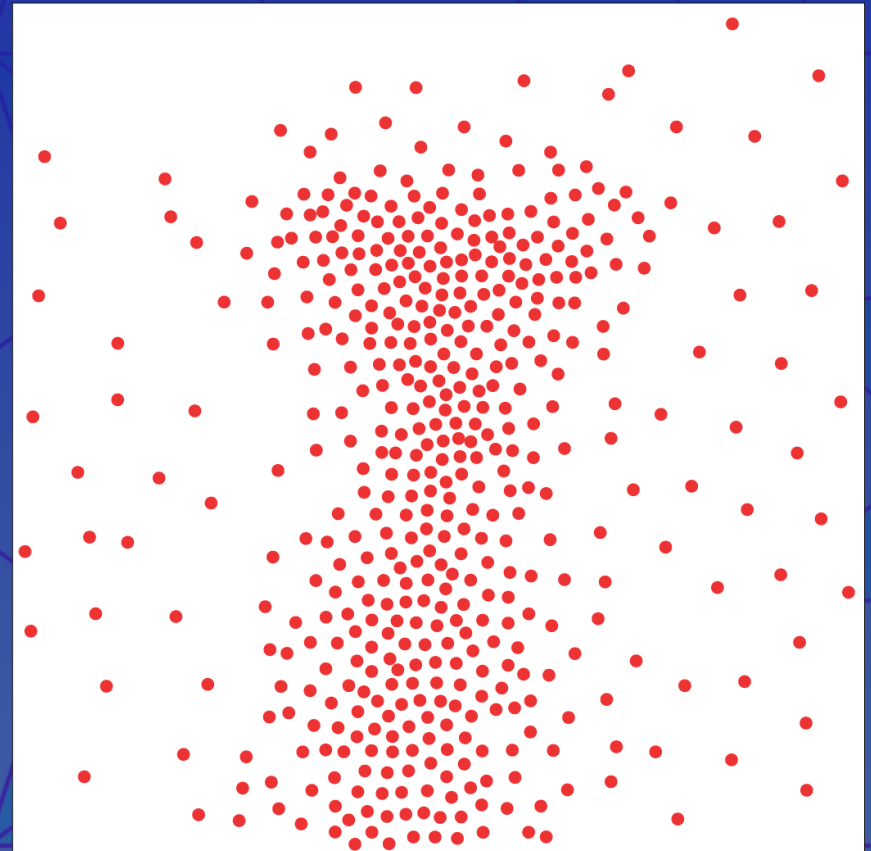


Corrective Vectors Lookup table

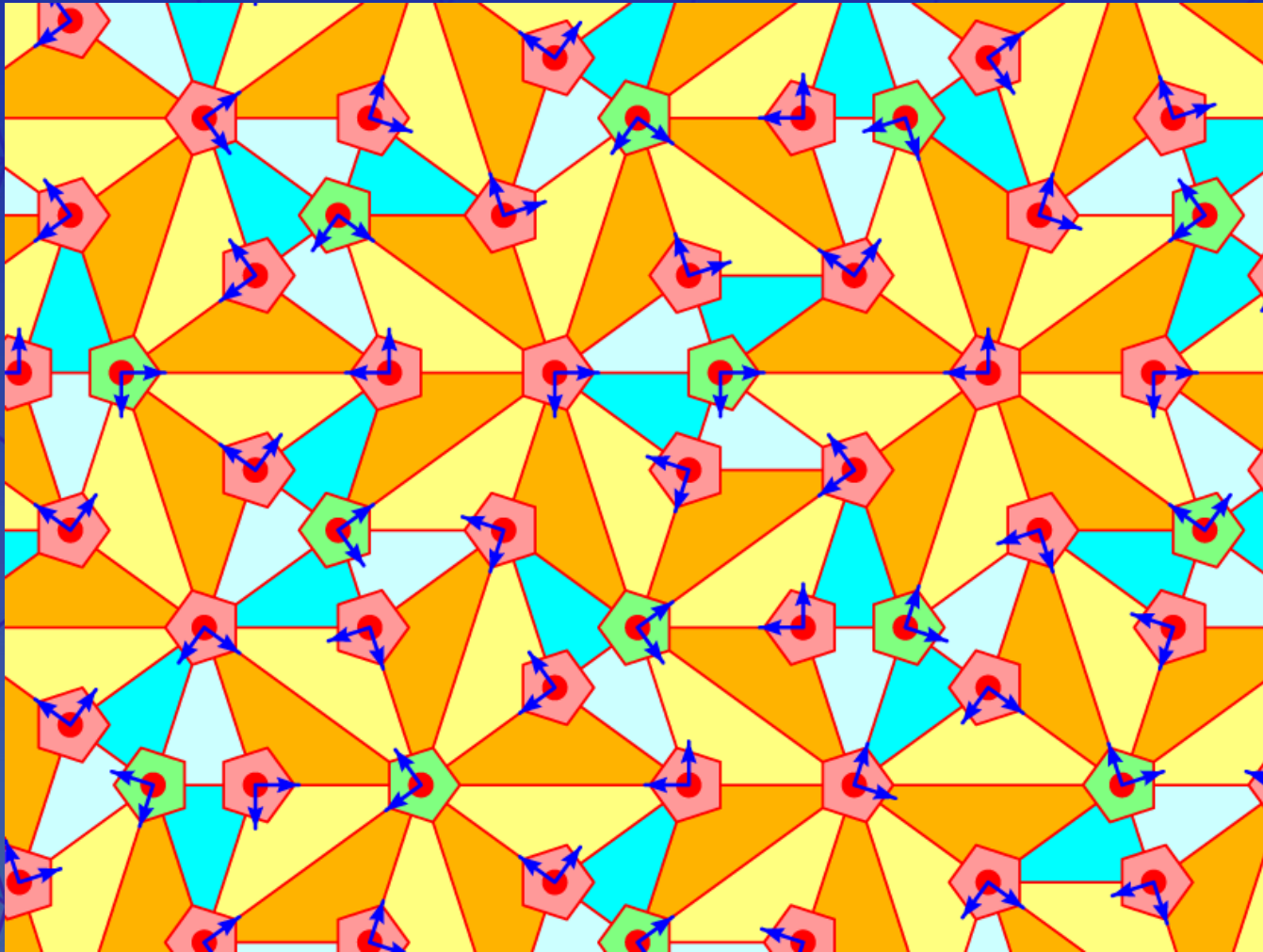
Before Correction



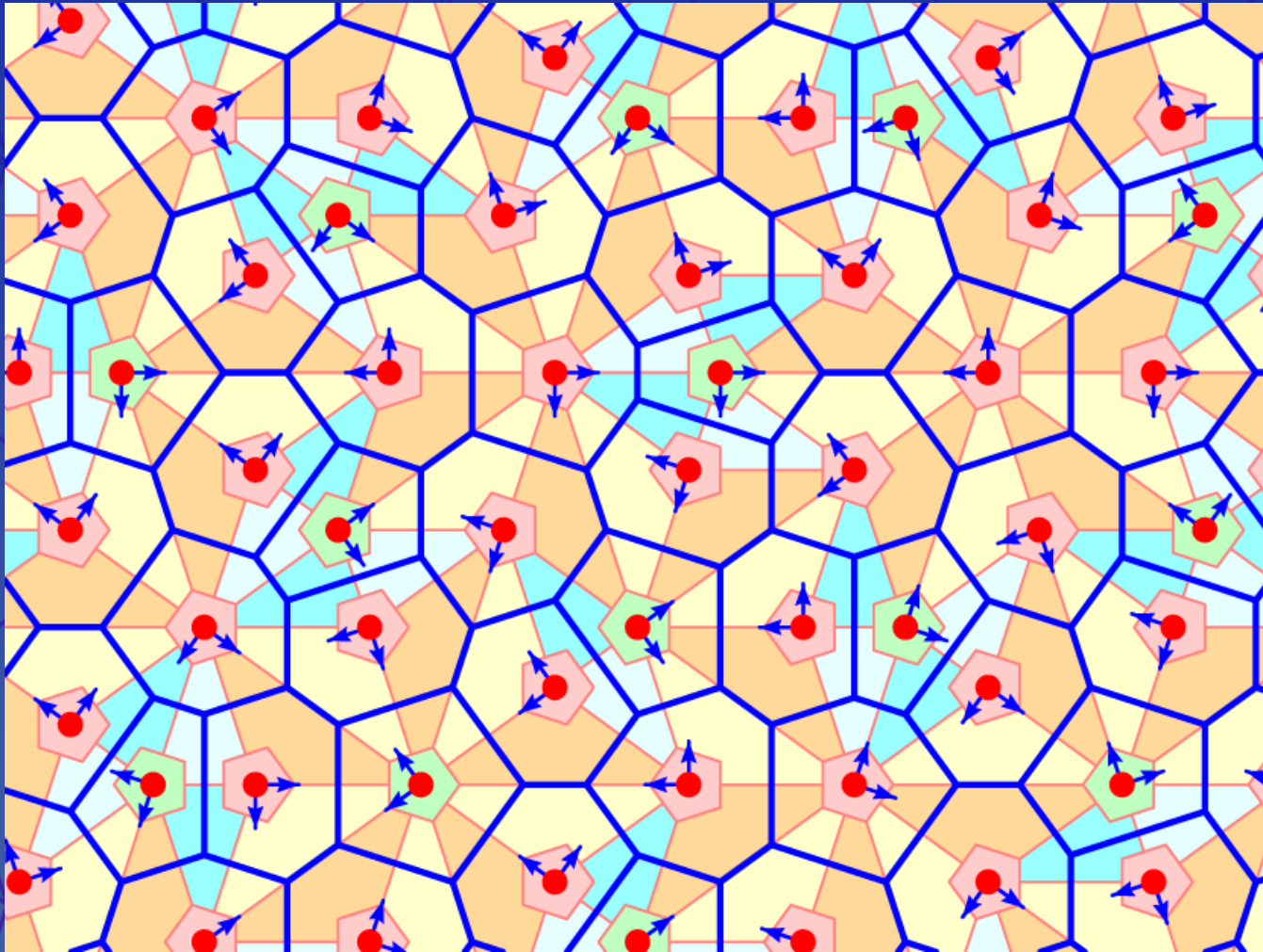
After Correction



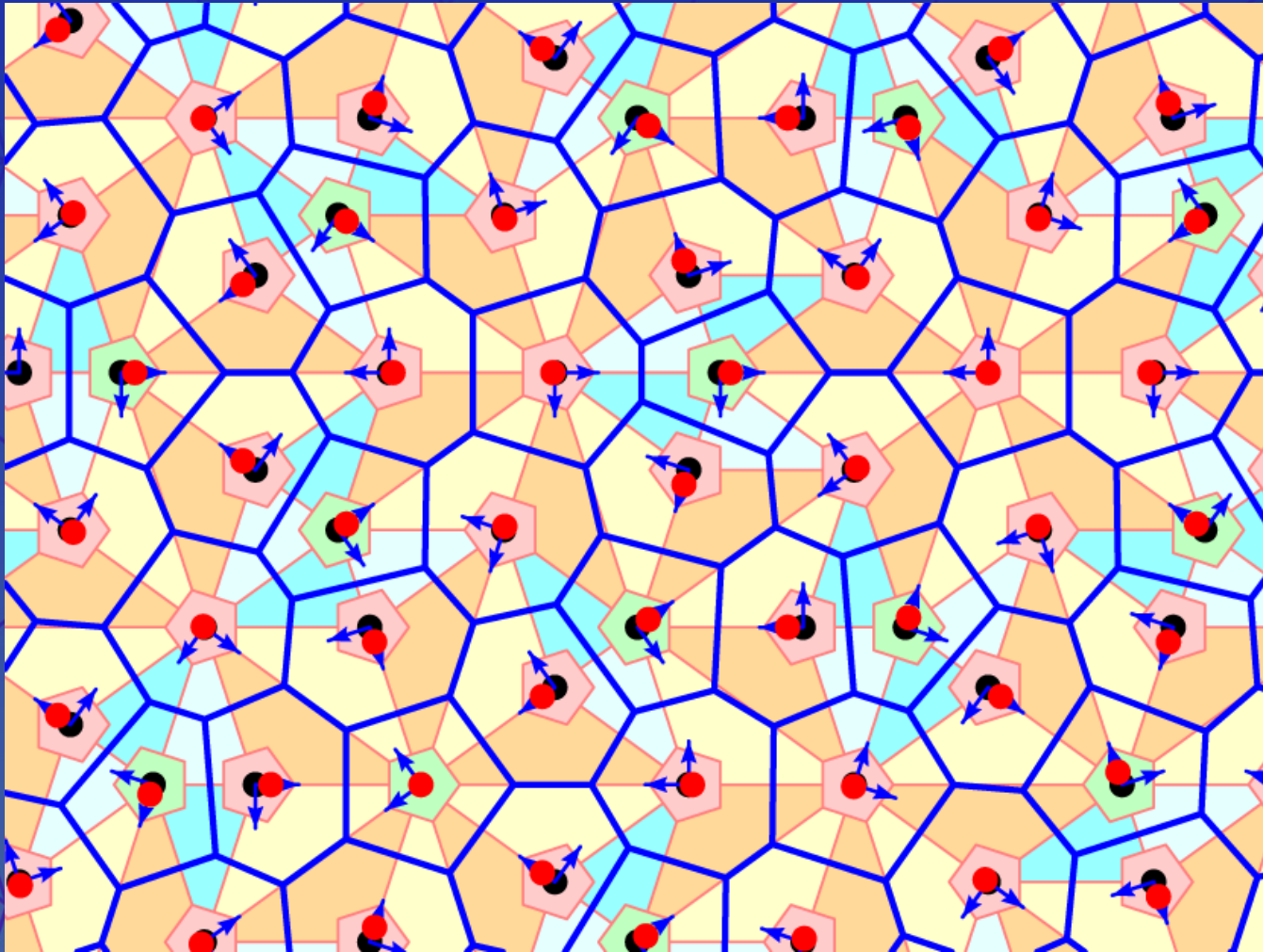
Offline Lloyd's Relaxation: init pts + basis frames



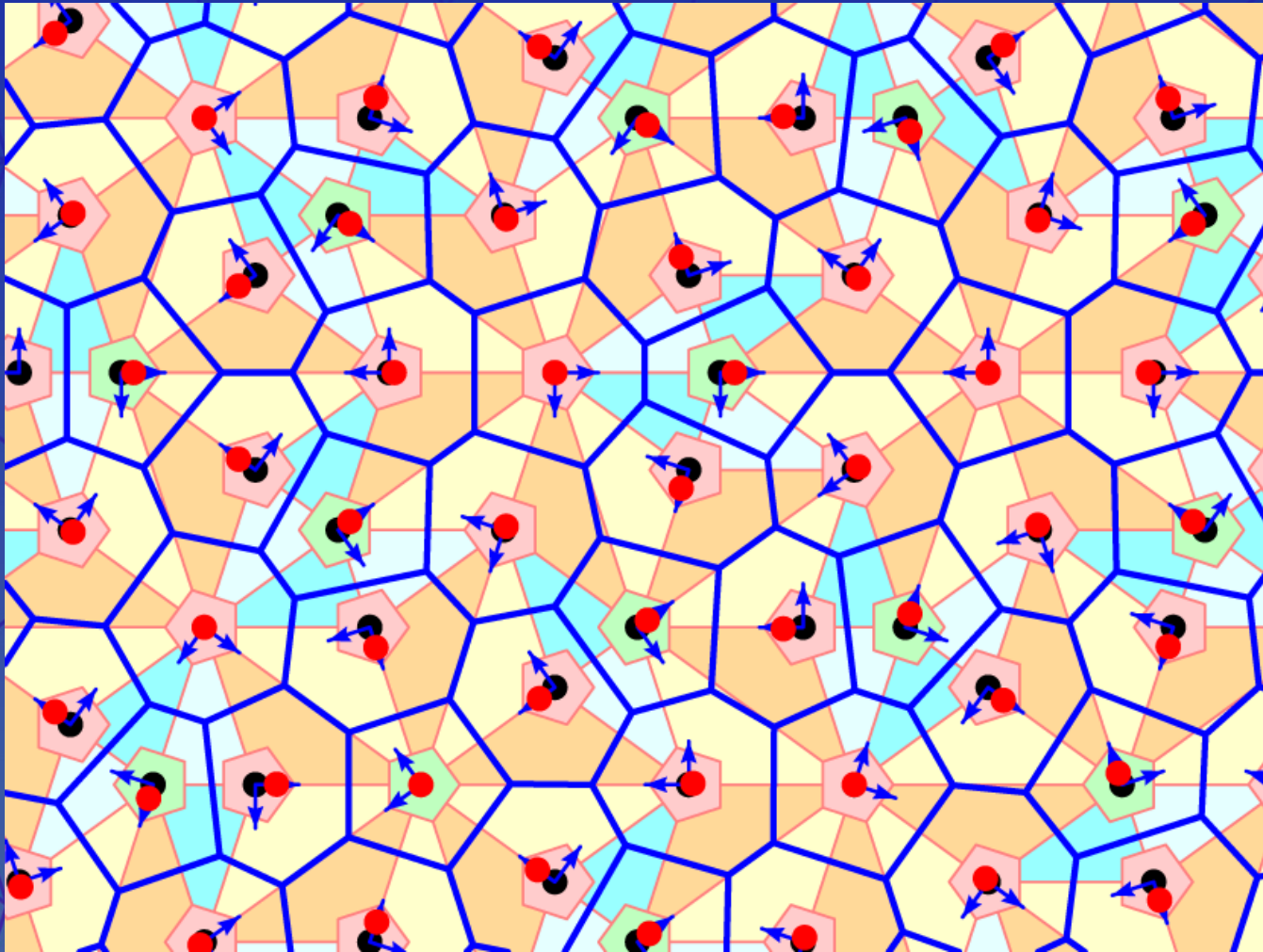
Offline Lloyd's Relaxation: init pts + basis frames



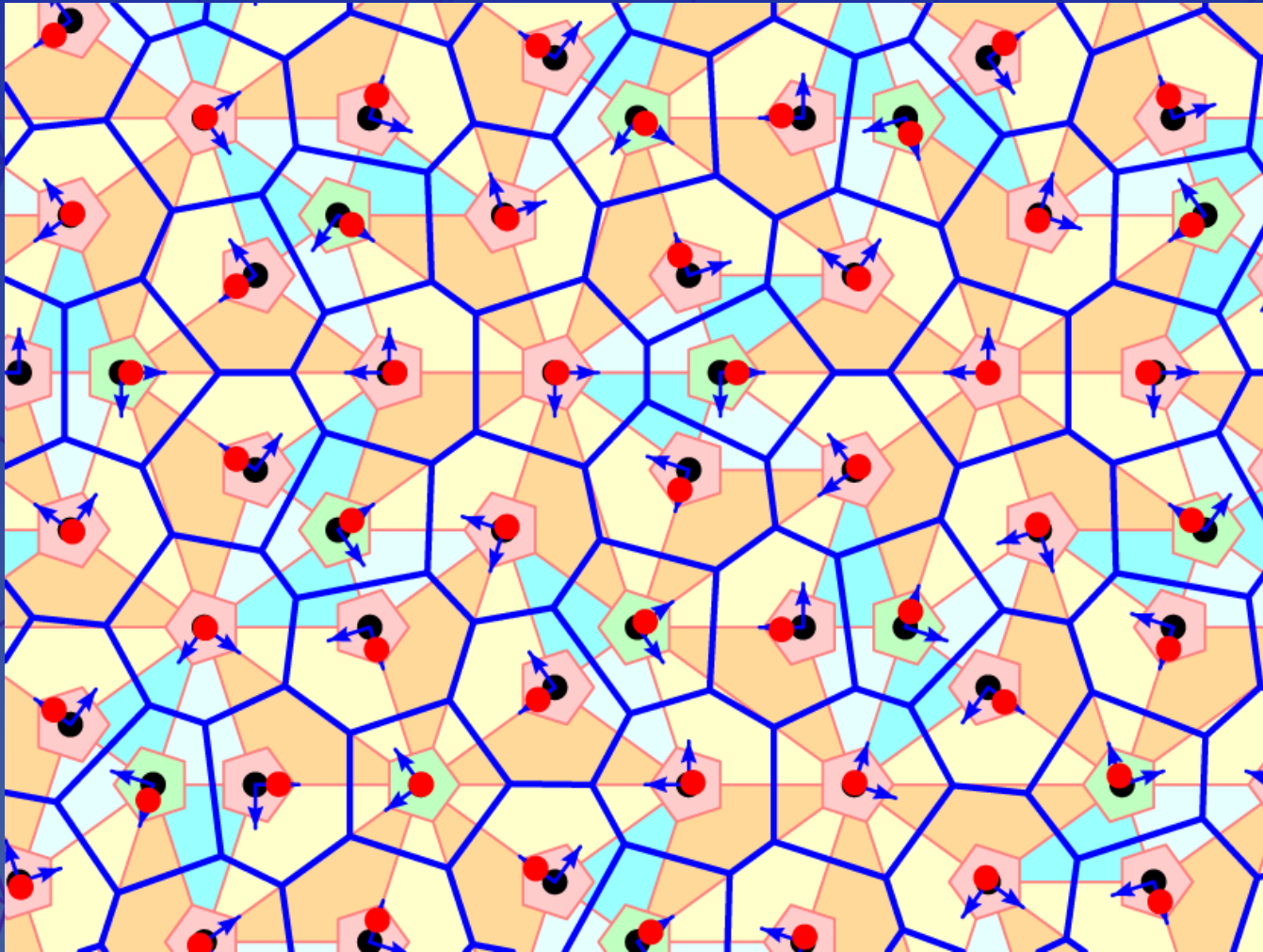
Offline Lloyd's Relaxation: iter 1



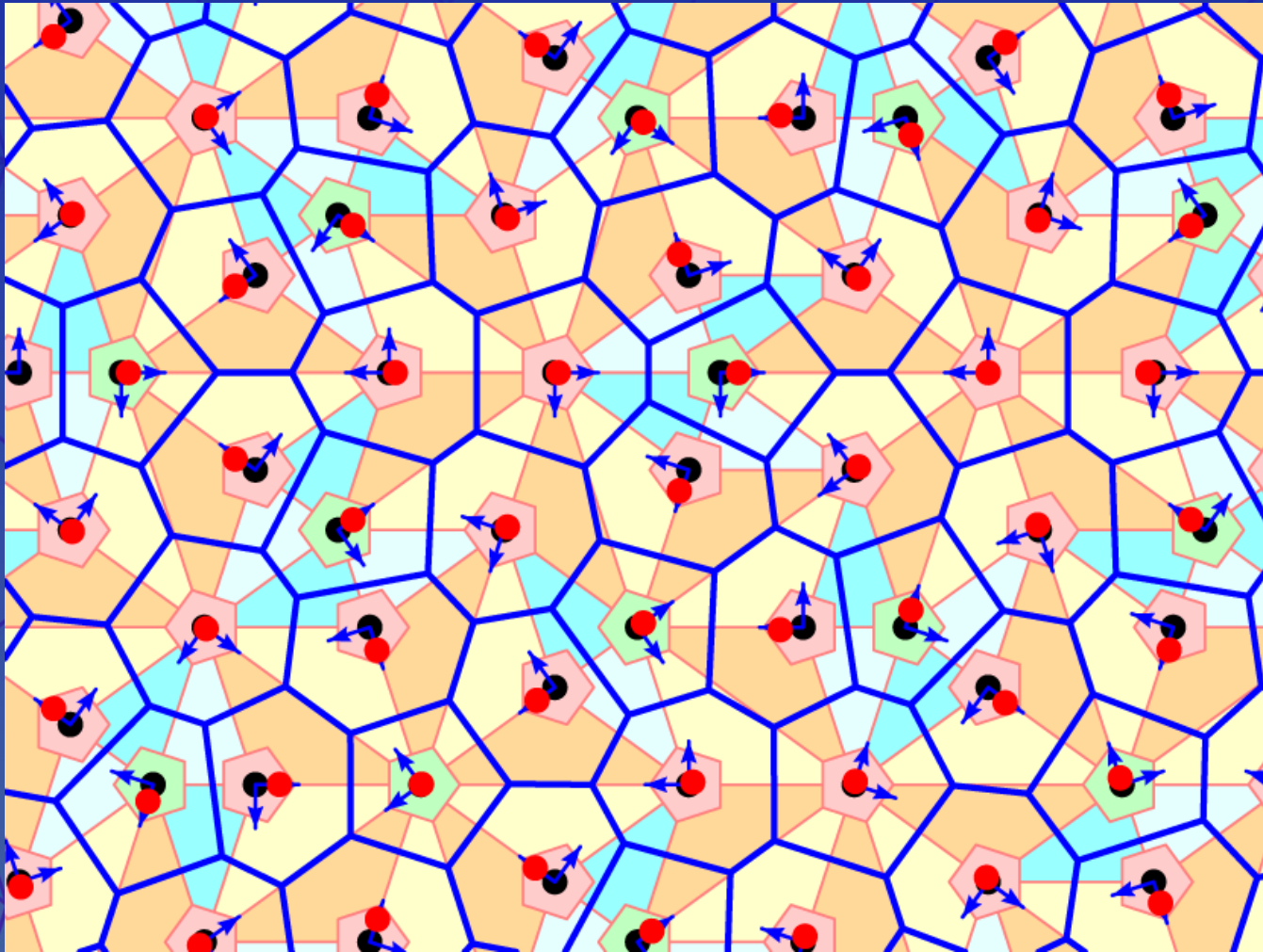
Offline Lloyd's Relaxation: iter 2



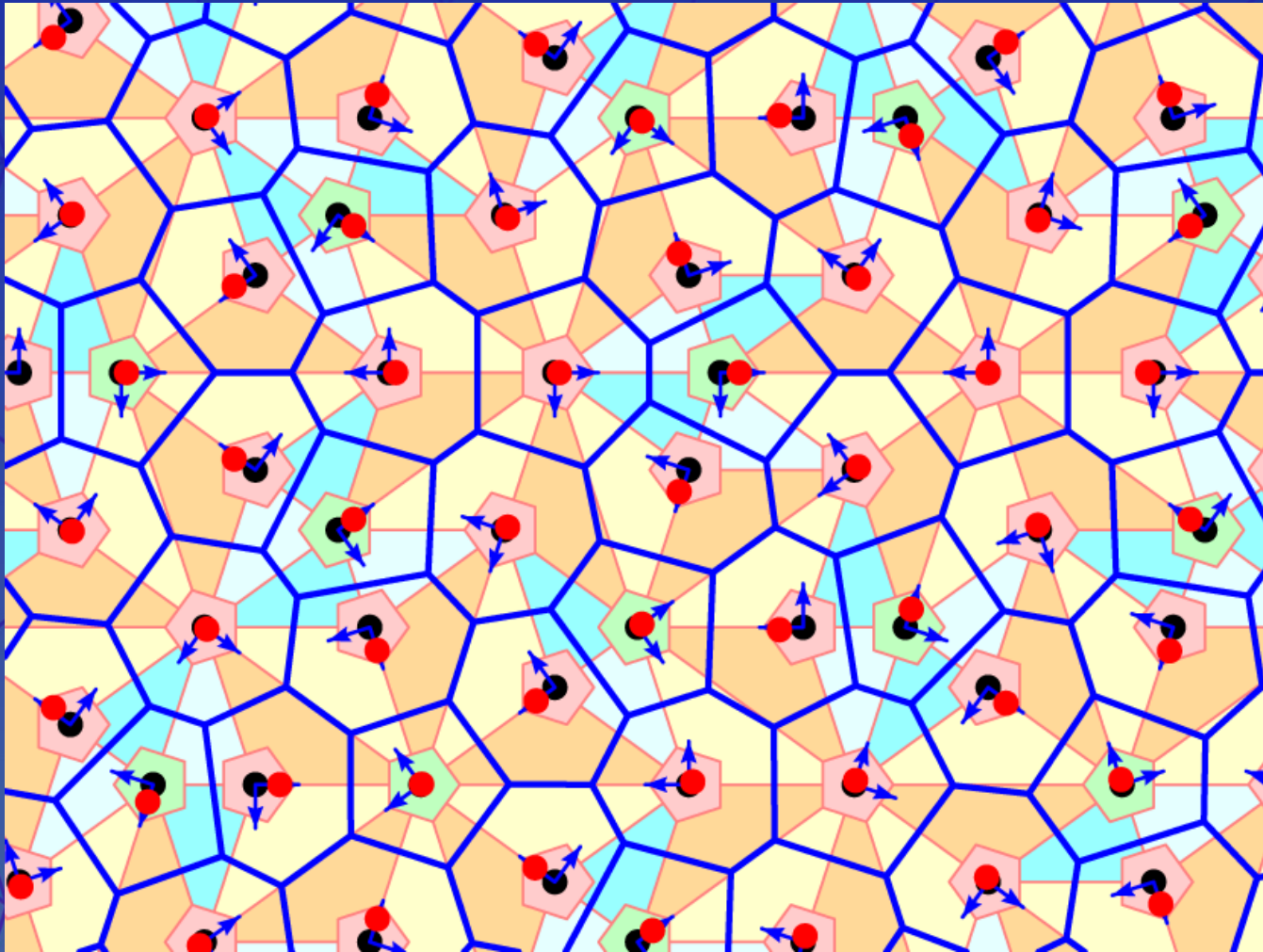
Offline Lloyd's Relaxation: iter 3



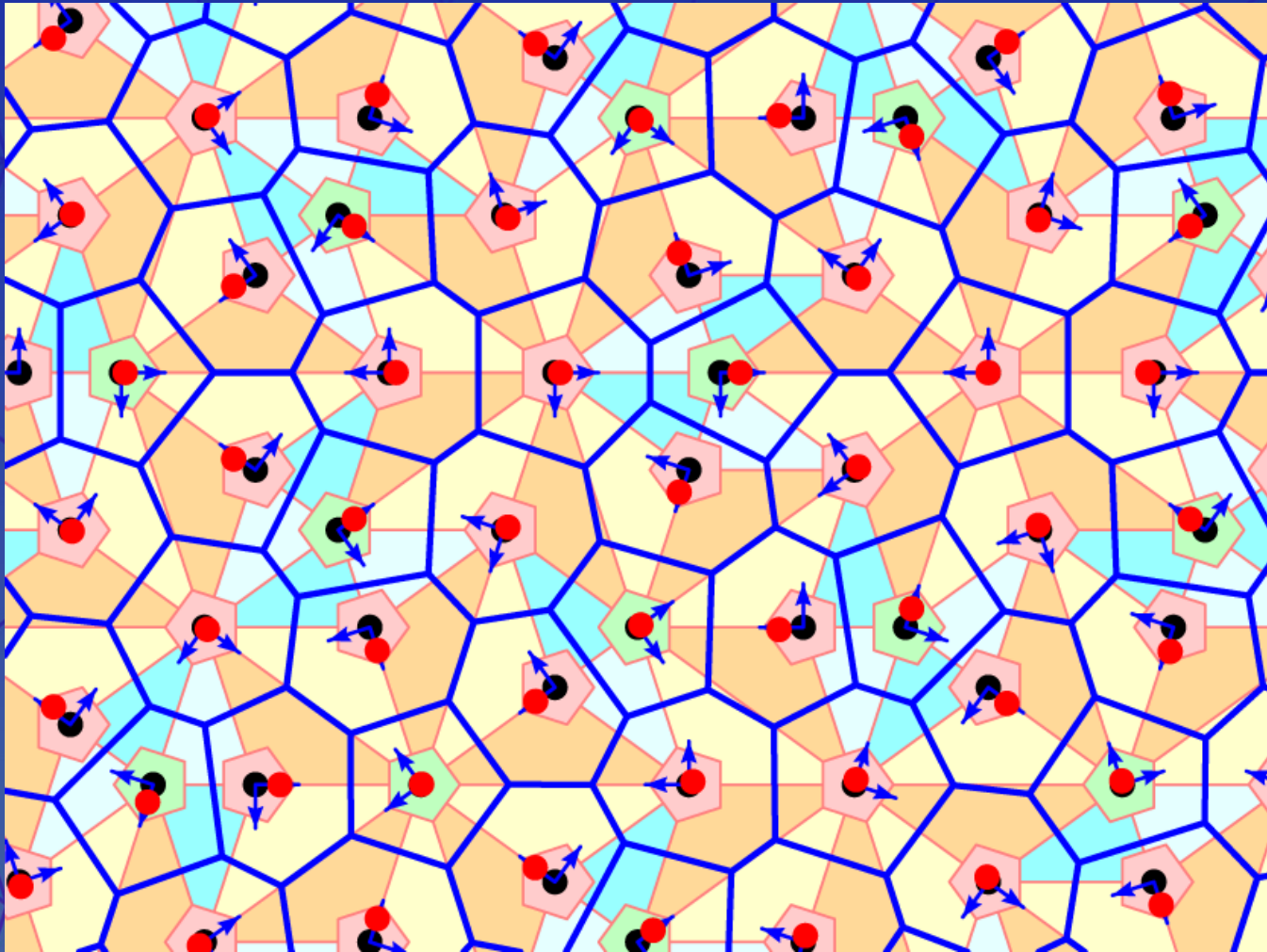
Offline Lloyd's Relaxation: iter 4



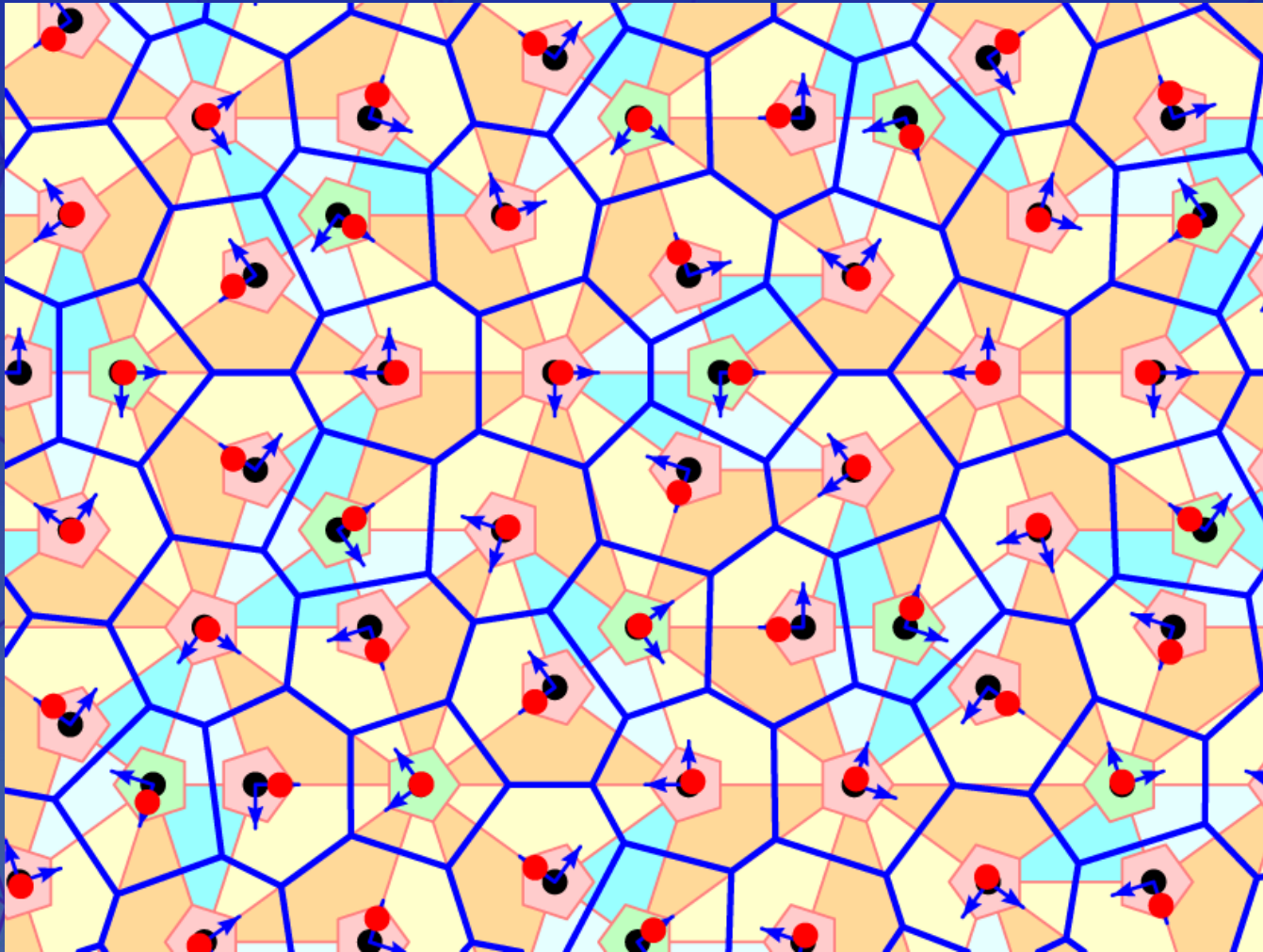
Offline Lloyd's Relaxation: iter 5



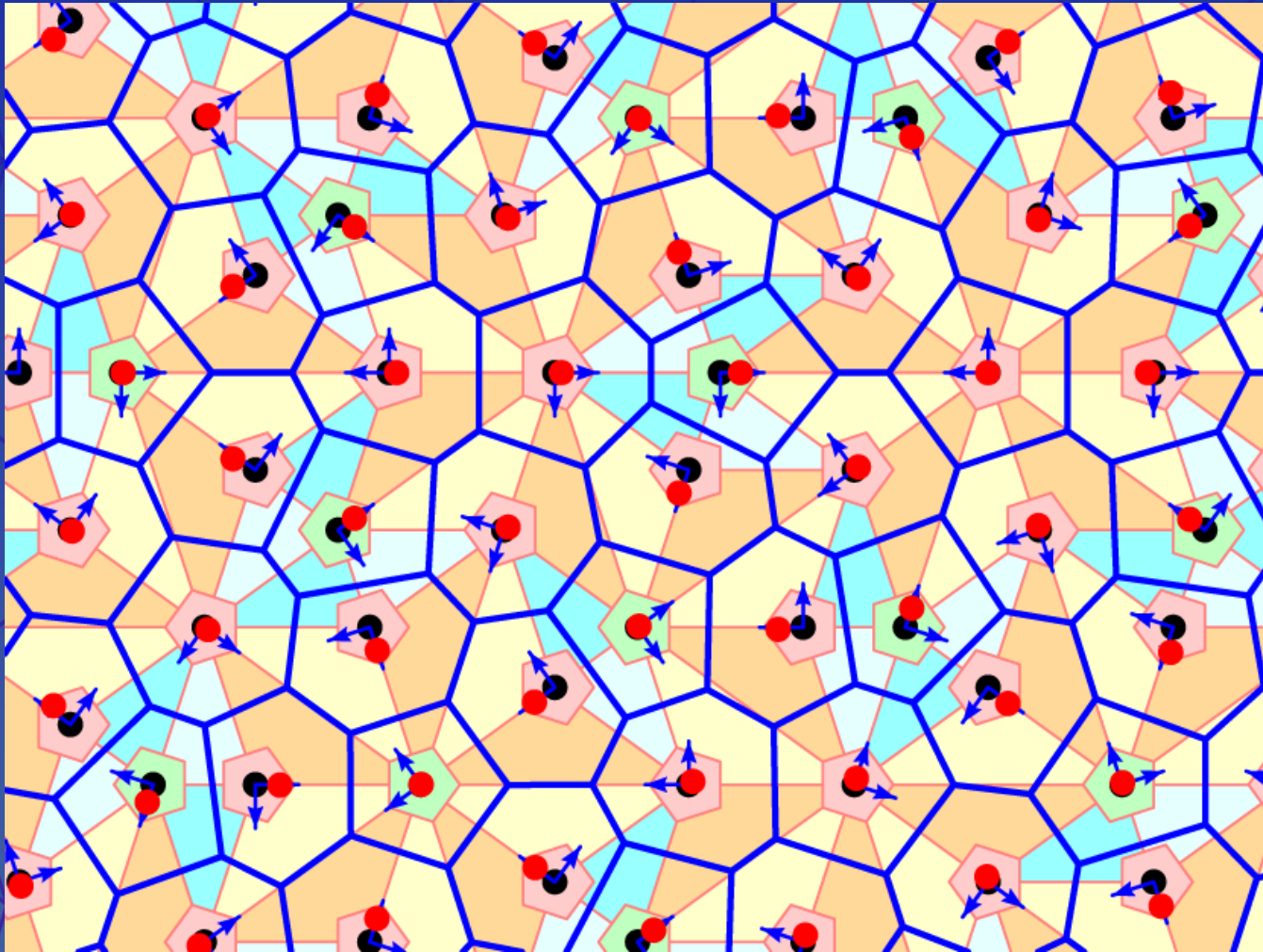
Offline Lloyd's Relaxation: iter 6



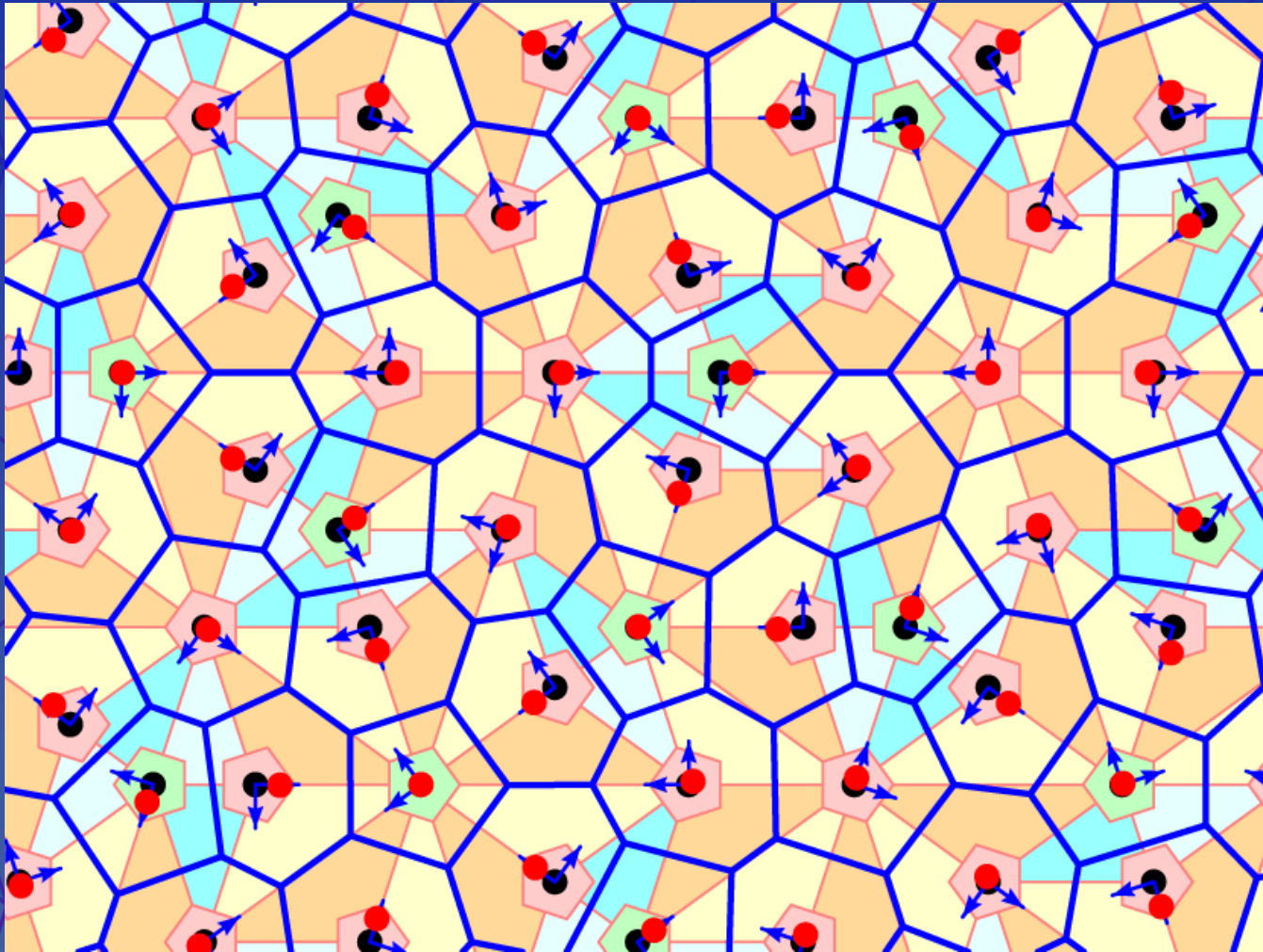
Offline Lloyd's Relaxation: iter 7



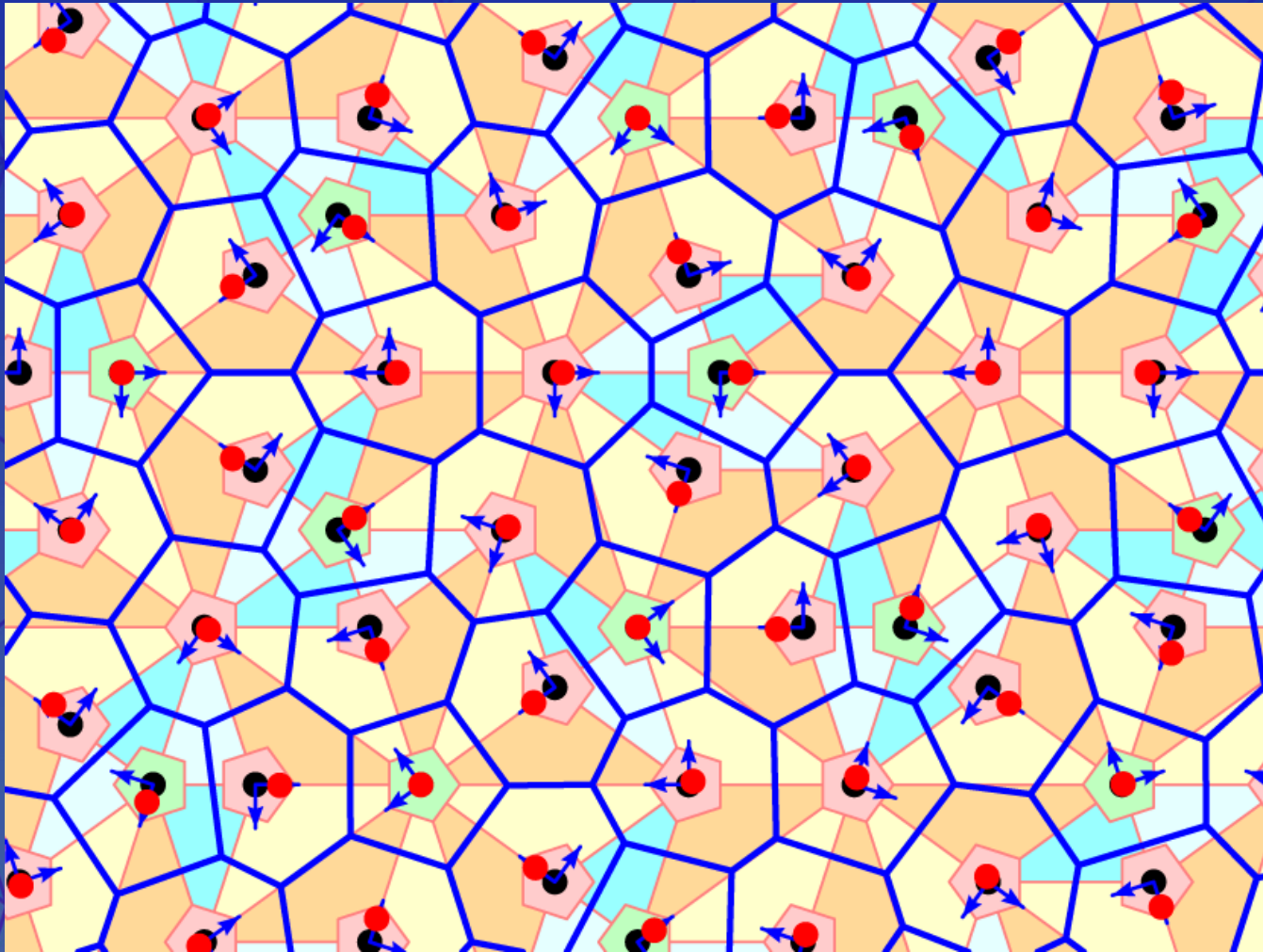
Offline Lloyd's Relaxation: iter 8



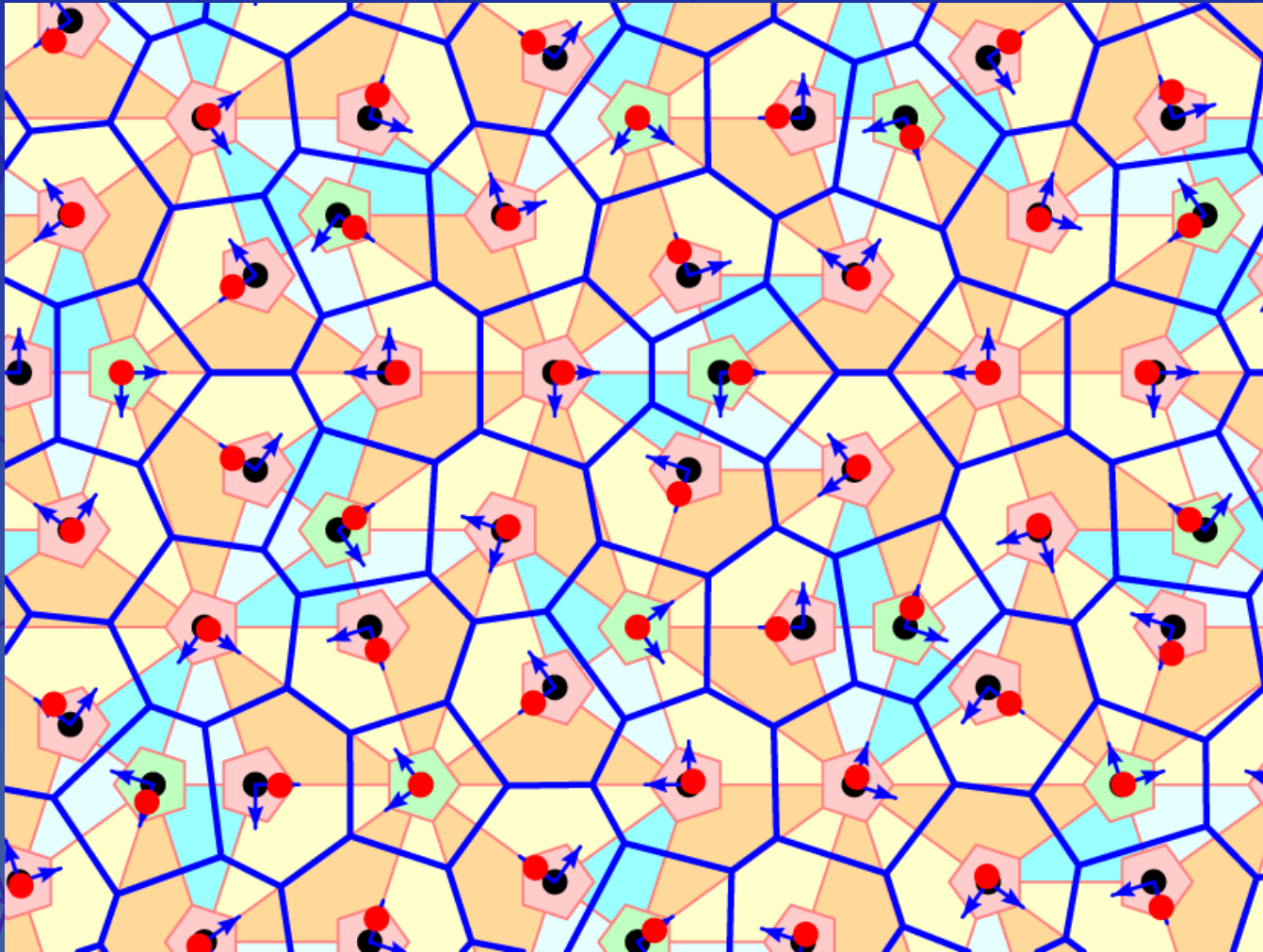
Offline Lloyd's Relaxation: iter 9



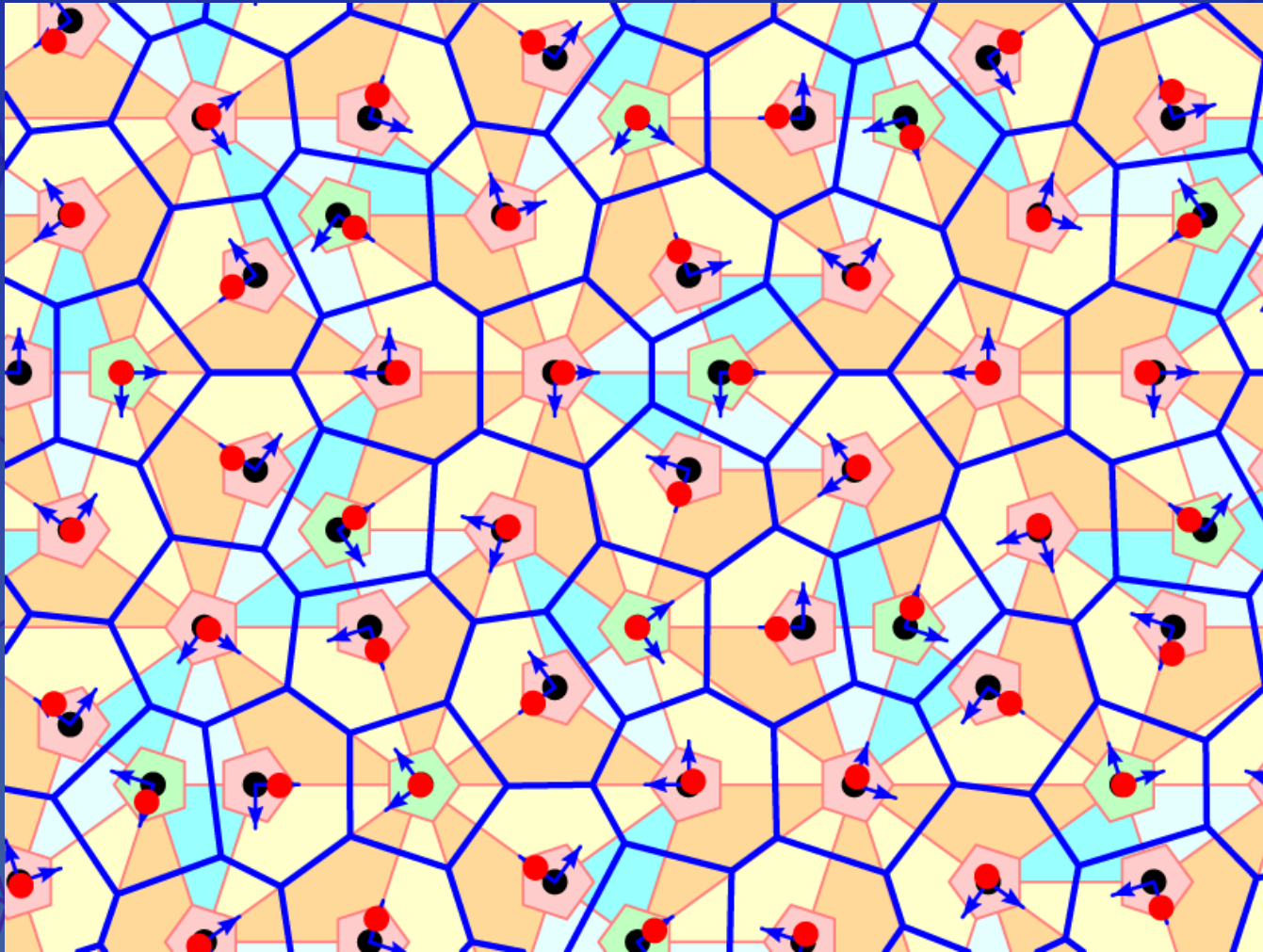
Offline Lloyd's Relaxation: iter 10



Offline Lloyd's Relaxation: iter 11

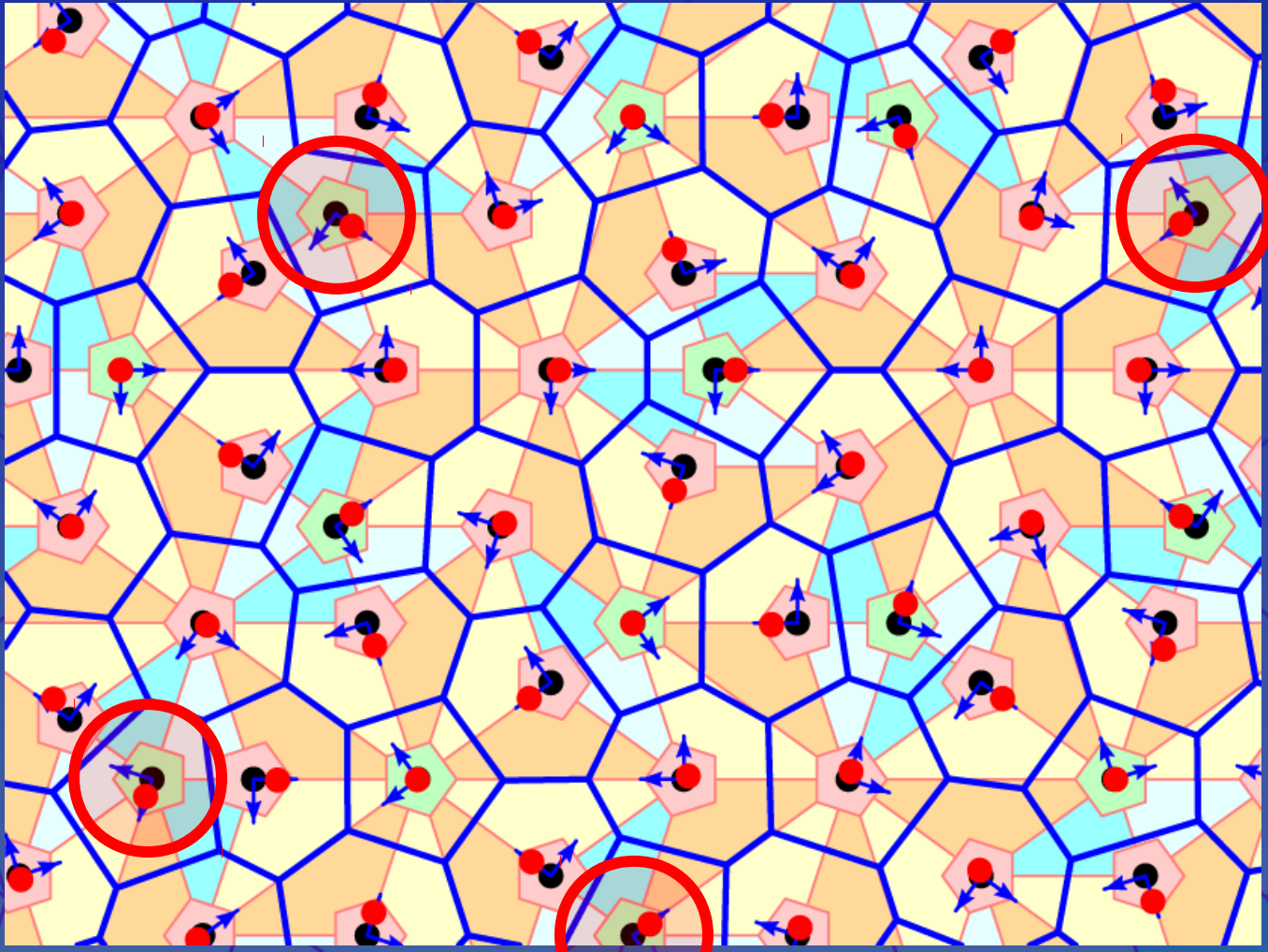


Offline Lloyd's Relaxation: iter 12



Corrective Vectors Lookup Table

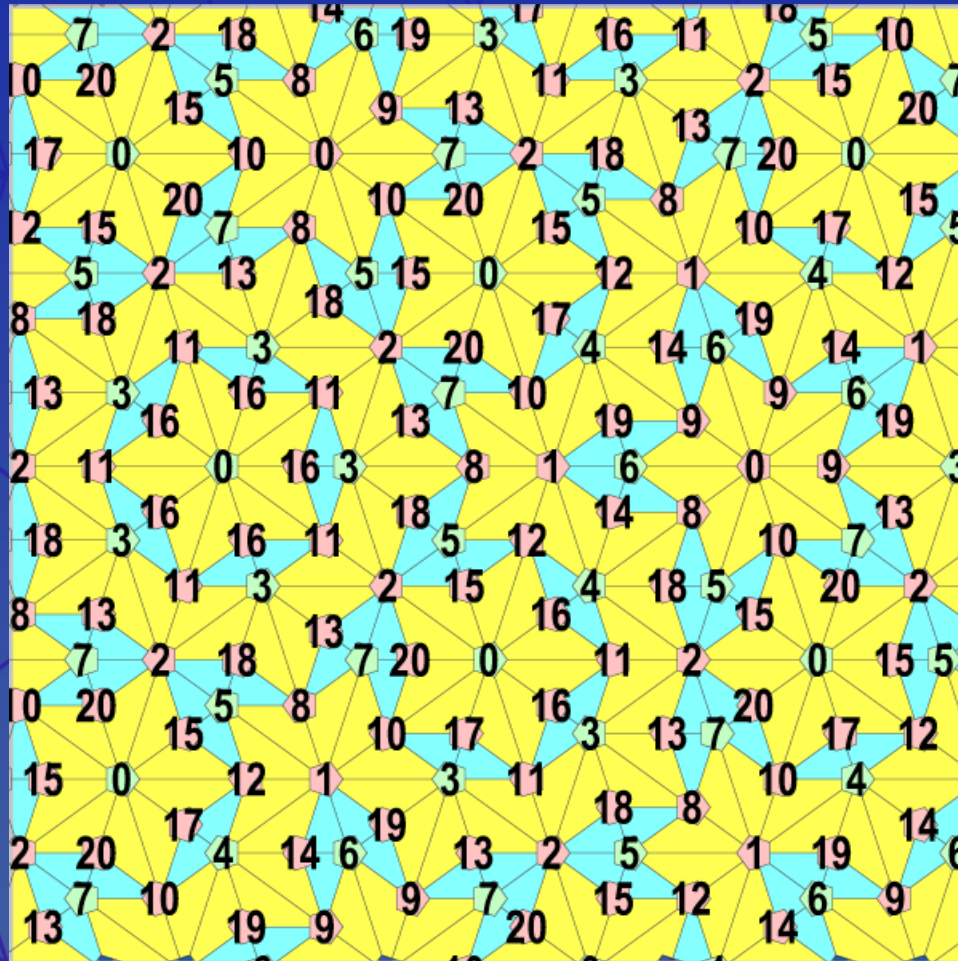
Lloyd's Relaxation: iter 12



Corrective Vectors Lookup Table

Structural Index i_s	Intensity Index i_v							
	0	1	2	3	4	5	6	7
0	(-.014,.018)	(.025,-.082)	(.049,-.097)	(.055,-.090)	(.038,-.074)	(-.006,-.058)	(-.027,-.028)	(-.018,.012)
1	(.013,-.007)	(-.133,-.292)	(-.208,-.358)	(-.197,-.380)	(-.129,-.323)	(-.066,-.266)	(-.029,-.218)	(-.004,-.204)
2	(.011,-.343)	(.079,-.563)	(.095,-.604)	(.072,-.583)	(-.053,-.504)	(-.144,-.383)	(-.170,-.265)	(-.124,-.172)
3	(.014,-.197)	(.085,-.233)	(.148,-.106)	(.236,.143)	(.317,.307)	(.345,.364)	(.296,.330)	(.198,.233)
4	(.036,.044)	(.180,-.014)	(.278,.078)	(.330,.222)	(.348,.358)	(.350,.386)	(.300,.367)	(.215,.301)
5	(.227,.365)	(.363,.061)	(.342,-.094)	(.268,-.156)	(.152,-.091)	(.004,.082)	(-.060,.256)	(-.039,.392)
6	(.171,.446)	(.301,-.089)	(.243,-.224)	(.132,-.299)	(-.052,-.258)	(-.228,-.155)	(-.266,-.013)	(-.200,.140)
7	(-.235,.391)	(-.580,.059)	(-.651,-.114)	(-.709,-.313)	(-.720,-.432)	(-.694,-.411)	(-.593,-.245)	(-.473,-.057)
8	(-.002,1.019)	(.081,.846)	(.118,.731)	(.168,.560)	(.182,.381)	(.221,.263)	(.216,.253)	(.197,.294)
9	(0,0)	(-.125,.256)	(-.108,.273)	(-.065,.249)	(.006,.171)	(.064,.086)	(.084,.026)	(.072,-.013)
10	(0,0)	(-.047,.030)	(-.104,.082)	(-.146,.159)	(-.173,.219)	(-.173,.221)	(-.151,.144)	(-.134,.077)
11	(0,0)	(-.070,.251)	(-.147,.532)	(-.177,.710)	(-.199,.735)	(-.187,.607)	(-.160,.366)	(-.114,.206)
12	(0,0)	(.001,.045)	(.003,.124)	(.009,.290)	(.012,.354)	(.011,.324)	(.010,.182)	(.019,.065)
13	(0,0)	(0,0)	(.020,.012)	(.063,.041)	(.148,.096)	(.198,.131)	(.218,.131)	(.212,.120)
14	(0,0)	(0,0)	(.017,.026)	(.048,.074)	(.107,.166)	(.124,.194)	(.117,.190)	(.099,.165)
15	(0,0)	(0,0)	(0,0)	(.012,.025)	(.038,.073)	(.092,.168)	(.124,.199)	(.138,.203)
16	(0,0)	(0,0)	(0,0)	(-.009,.009)	(-.024,.035)	(-.051,.094)	(-.044,.158)	(-.033,.203)
17	(0,0)	(0,0)	(0,0)	(0,0)	(-.008,.018)	(-.025,.058)	(-.051,.136)	(-.059,.184)
18	(0,0)	(0,0)	(0,0)	(0,0)	(.004,.010)	(.003,.031)	(-.011,.069)	(-.049,.090)
19	(0,0)	(0,0)	(0,0)	(0,0)	(0,0)	(-.016,.013)	(-.045,.030)	(-.100,.067)
20	(0,0)	(0,0)	(0,0)	(0,0)	(0,0)	(-.010,.034)	(-.029,.078)	(-.064,.174)

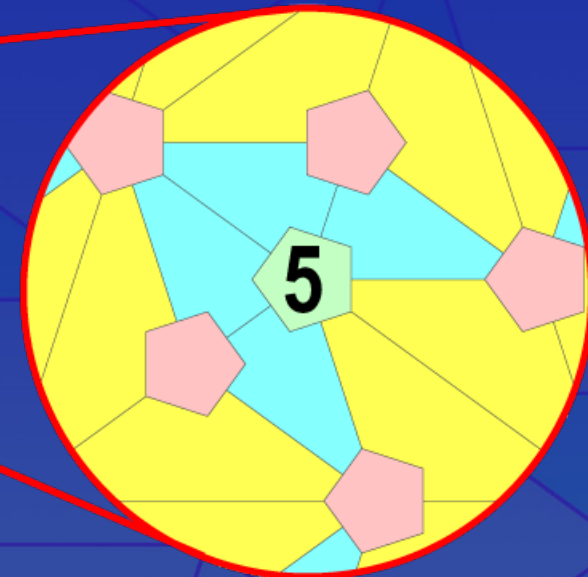
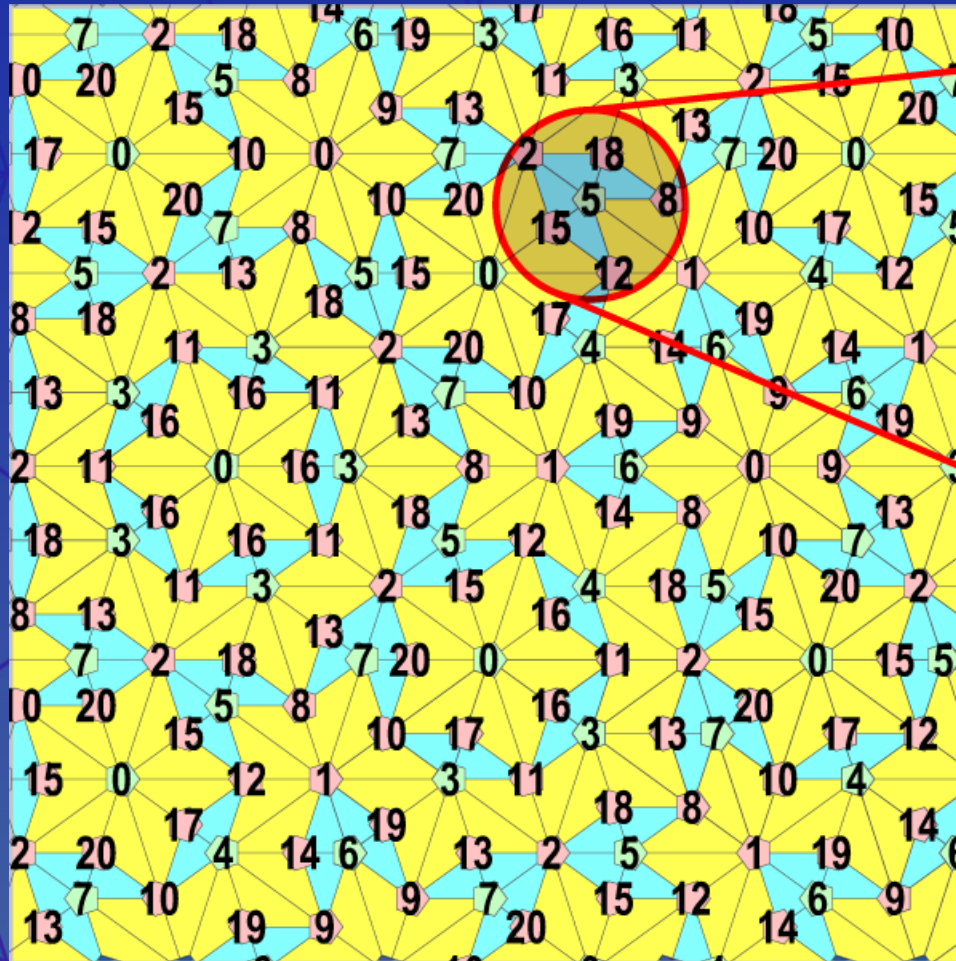
Corrective Vectors Lookup Table Structural Indices (i_s)



6 most-significant bits in F-code

Corrective Vectors Lookup Table

Structural Indices (i_s)



F-code:

00100000000010

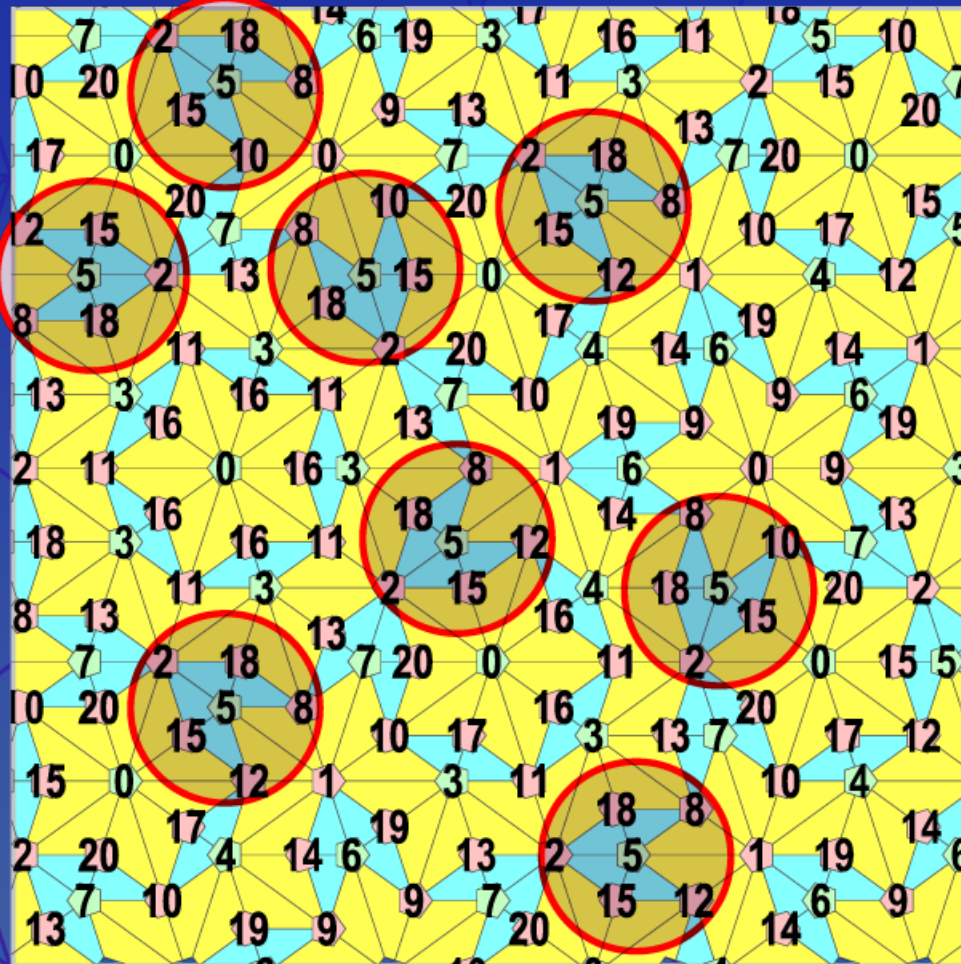
6 most-significant bits:

00100000000010

Decimal value: 5

Corrective Vectors Lookup Table

Structural Indices (i_s)



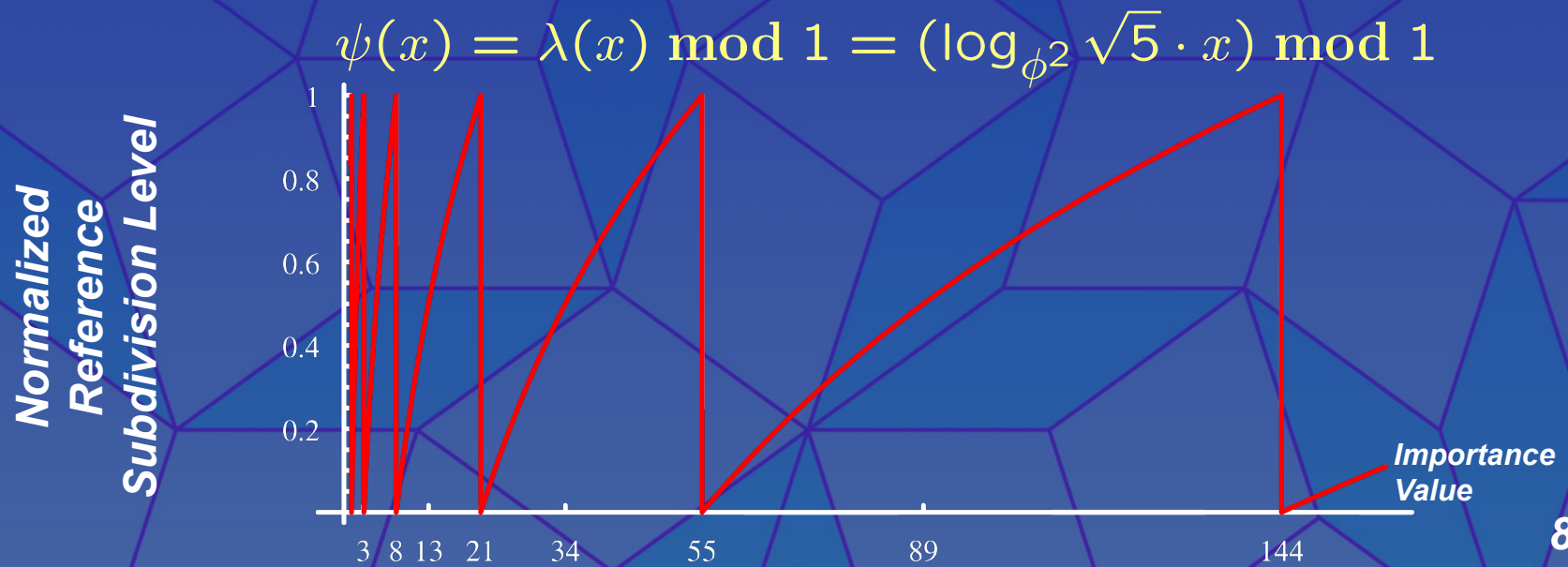
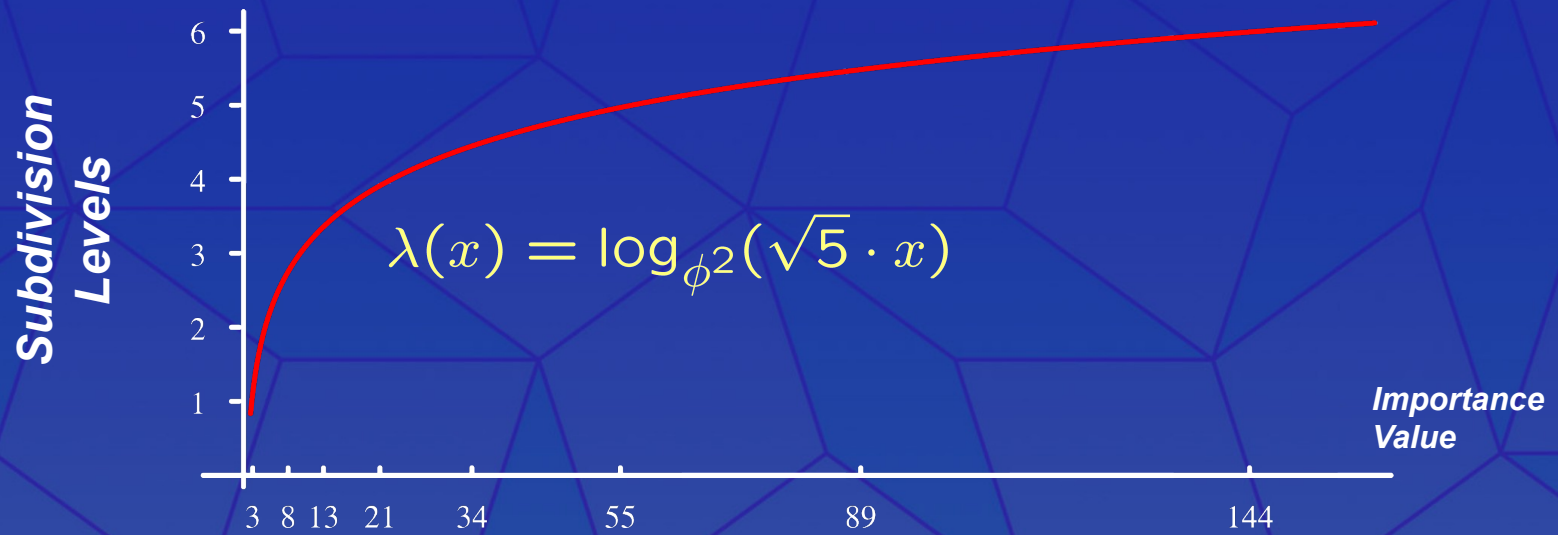
$$i_s = 5$$

Corrective Vectors Lookup Table

Structural Index i_s	Intensity Index i_v							
	0	1	2	3	4	5	6	7
0	(-.014,.018)	(.025,-.082)	(.049,-.097)	(.055,-.090)	(.038,-.074)	(-.006,-.058)	(-.027,-.028)	(-.018,.012)
1	(.013,-.007)	(-.133,-.292)	(-.208,-.358)	(-.197,-.380)	(-.129,-.323)	(-.066,-.266)	(-.029,-.218)	(-.004,-.204)
2	(.011,-.343)	(.079,-.563)	(.095,-.604)	(.072,-.583)	(-.053,-.504)	(-.144,-.383)	(-.170,-.265)	(-.124,-.172)
3	(.014,-.197)	(.085,-.233)	(.148,-.106)	(.236,.143)	(.317,.307)	(.345,.364)	(.296,.330)	(.198,.233)
4	(.036,.044)	(.180,-.014)	(.278,.078)	(.330,.222)	(.348,.358)	(.350,.386)	(.300,.367)	(.215,.301)
5	(.227,.365)	(.363,.061)	(.342,-.094)	(.268,-.156)	(.152,-.091)	(.004,.082)	(-.060,.256)	(-.039,.392)
6	(.171,.446)	(.301,-.089)	(.243,-.224)	(.132,-.299)	(-.052,-.258)	(-.228,-.155)	(-.266,-.013)	(-.200,.140)
7	(-.235,.391)	(-.580,.059)	(-.651,-.114)	(-.709,-.313)	(-.720,-.432)	(-.694,-.411)	(-.593,-.245)	(-.473,-.057)
8	(-.002,1.019)	(.081,.846)	(.118,.731)	(.168,.560)	(.182,.381)	(.221,.263)	(.216,.253)	(.197,.294)
9	(0,0)	(-.125,.256)	(-.108,.273)	(-.065,.249)	(.006,.171)	(.064,.086)	(.084,.026)	(.072,-.013)
10	(0,0)	(-.047,.030)	(-.104,.082)	(-.146,.159)	(-.173,.219)	(-.173,.221)	(-.151,.144)	(-.134,.077)
11	(0,0)	(-.070,.251)	(-.147,.532)	(-.177,.710)	(-.199,.735)	(-.187,.607)	(-.160,.366)	(-.114,.206)
12	(0,0)	(.001,.045)	(.003,.124)	(.009,.290)	(.012,.354)	(.011,.324)	(.010,.182)	(.019,.065)
13	(0,0)	(0,0)	(.020,.012)	(.063,.041)	(.148,.096)	(.198,.131)	(.218,.131)	(.212,.120)
14	(0,0)	(0,0)	(.017,.026)	(.048,.074)	(.107,.166)	(.124,.194)	(.117,.190)	(.099,.165)
15	(0,0)	(0,0)	(0,0)	(.012,.025)	(.038,.073)	(.092,.168)	(.124,.199)	(.138,.203)
16	(0,0)	(0,0)	(0,0)	(-.009,.009)	(-.024,.035)	(-.051,.094)	(-.044,.158)	(-.033,.203)
17	(0,0)	(0,0)	(0,0)	(0,0)	(-.008,.018)	(-.025,.058)	(-.051,.136)	(-.059,.184)
18	(0,0)	(0,0)	(0,0)	(0,0)	(.004,.010)	(.003,.031)	(-.011,.069)	(-.049,.090)
19	(0,0)	(0,0)	(0,0)	(0,0)	(0,0)	(-.016,.013)	(-.045,.030)	(-.100,.067)
20	(0,0)	(0,0)	(0,0)	(0,0)	(0,0)	(-.010,.034)	(-.029,.078)	(-.064,.174)

Corrective Vectors Lookup Table

Importance Indices (i_v)

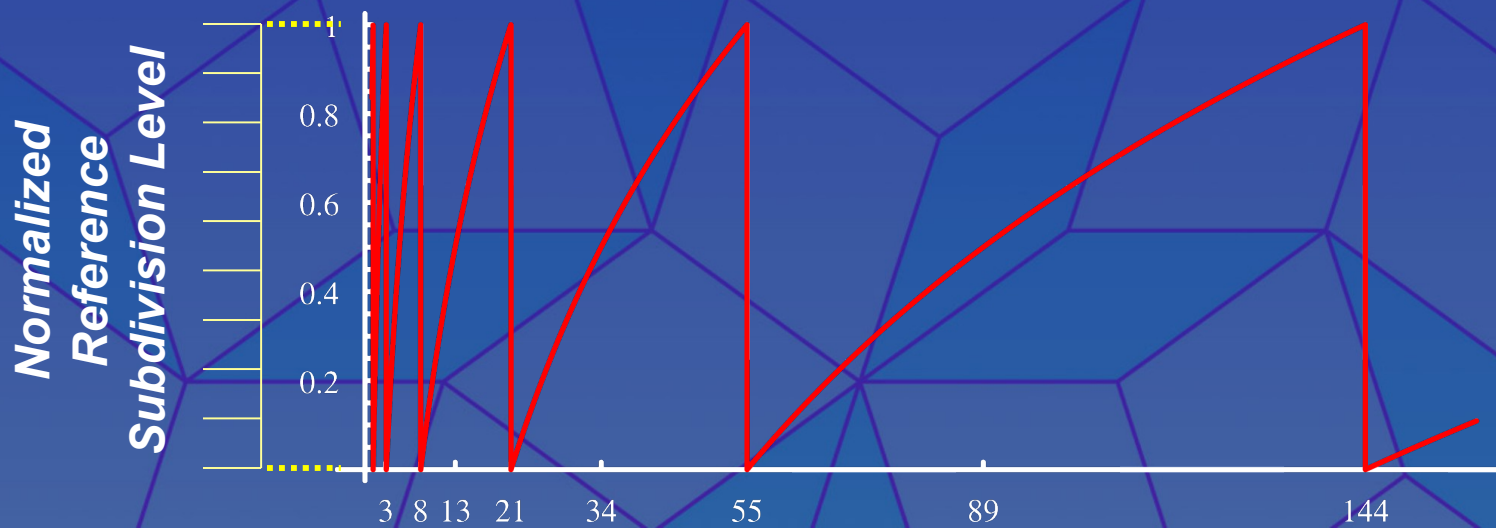


Corrective Vectors Lookup Table

Importance Indices (i_v)

$$i_v = \lfloor n \cdot \psi(\text{mag} \cdot I(x, y)) \rfloor$$

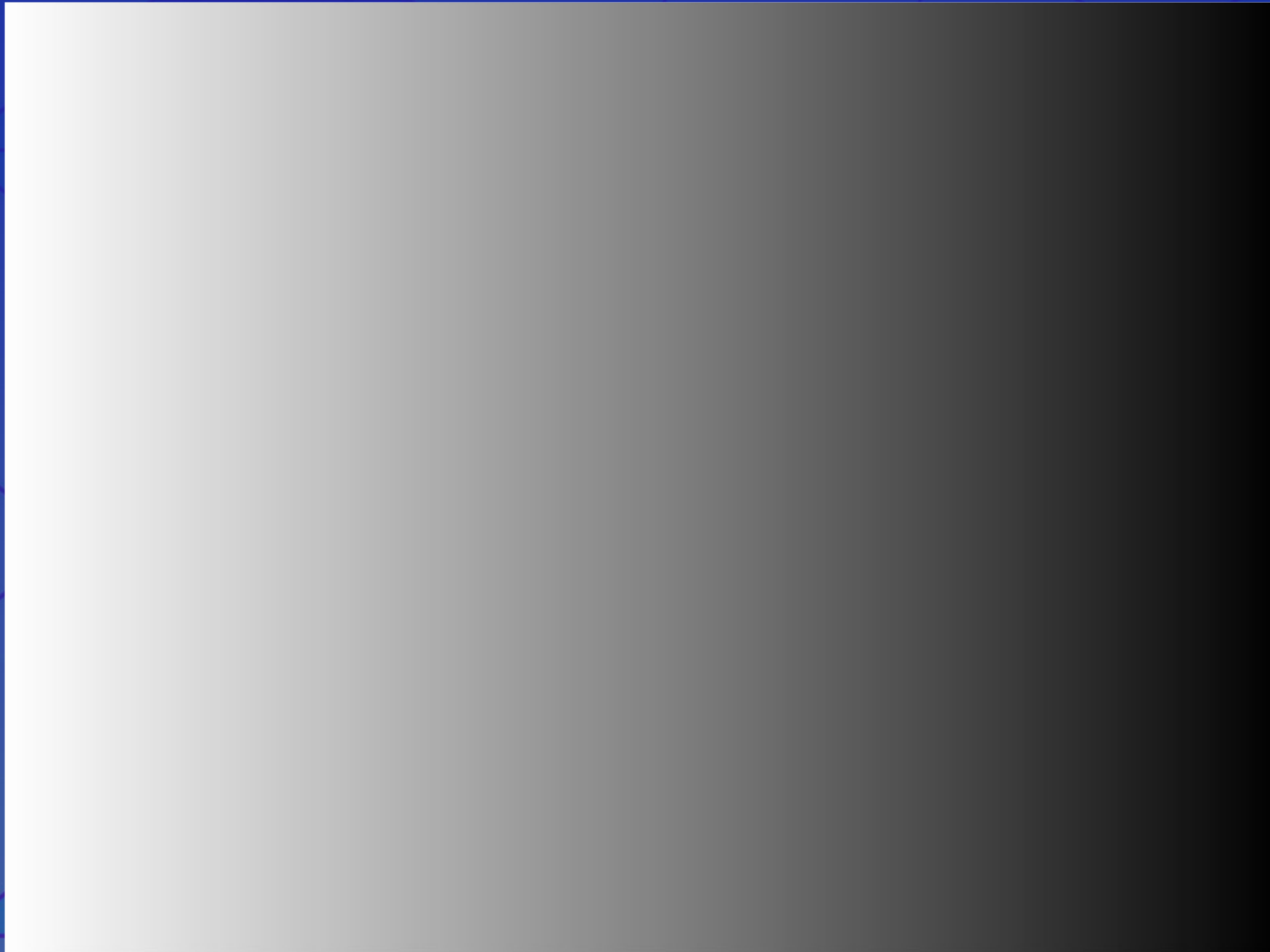
$$\psi(x) = (\log_{\phi^2} \sqrt{5} \cdot x) \bmod 1$$



System Summary

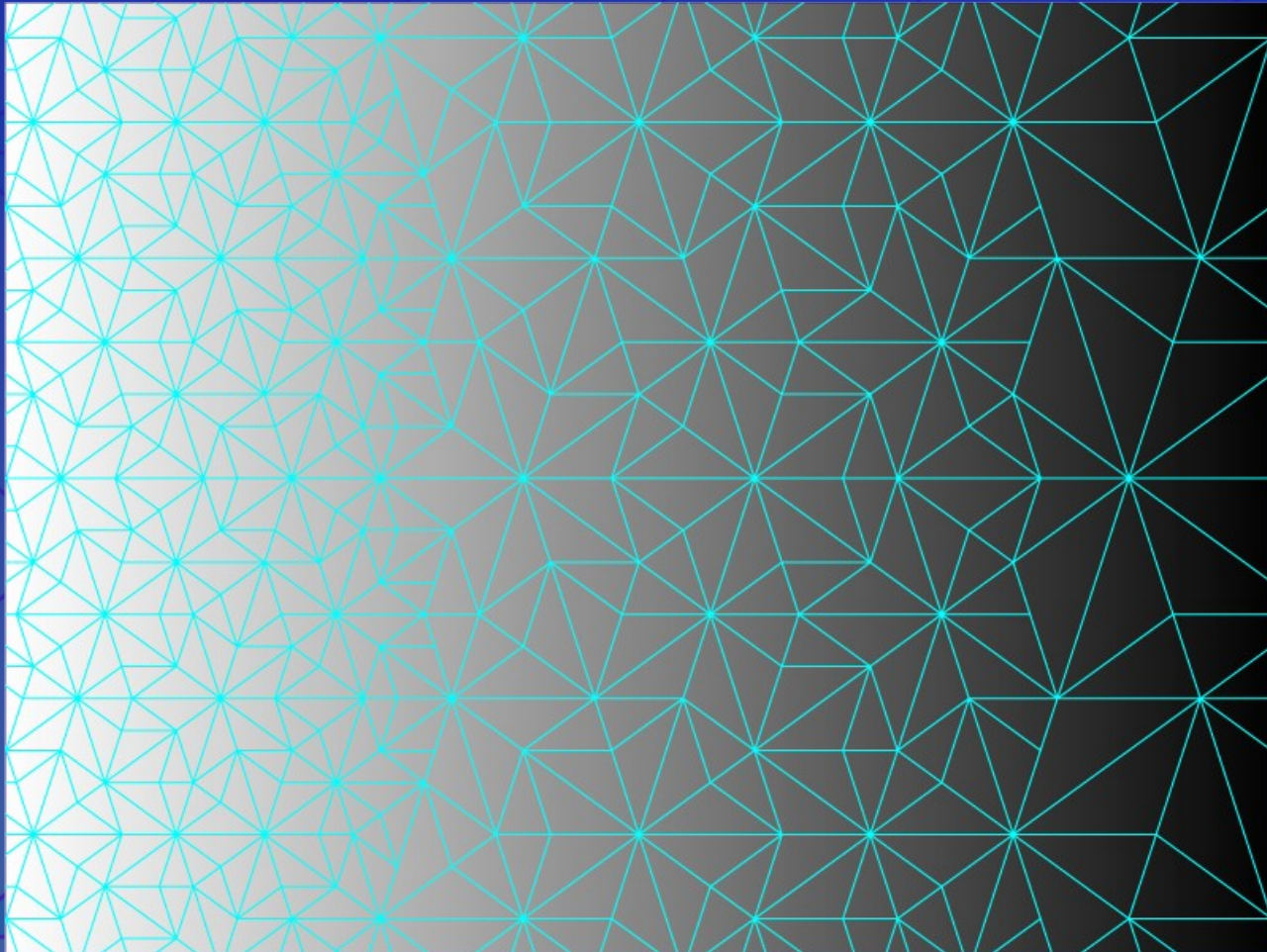
System Summary

Sampled density function



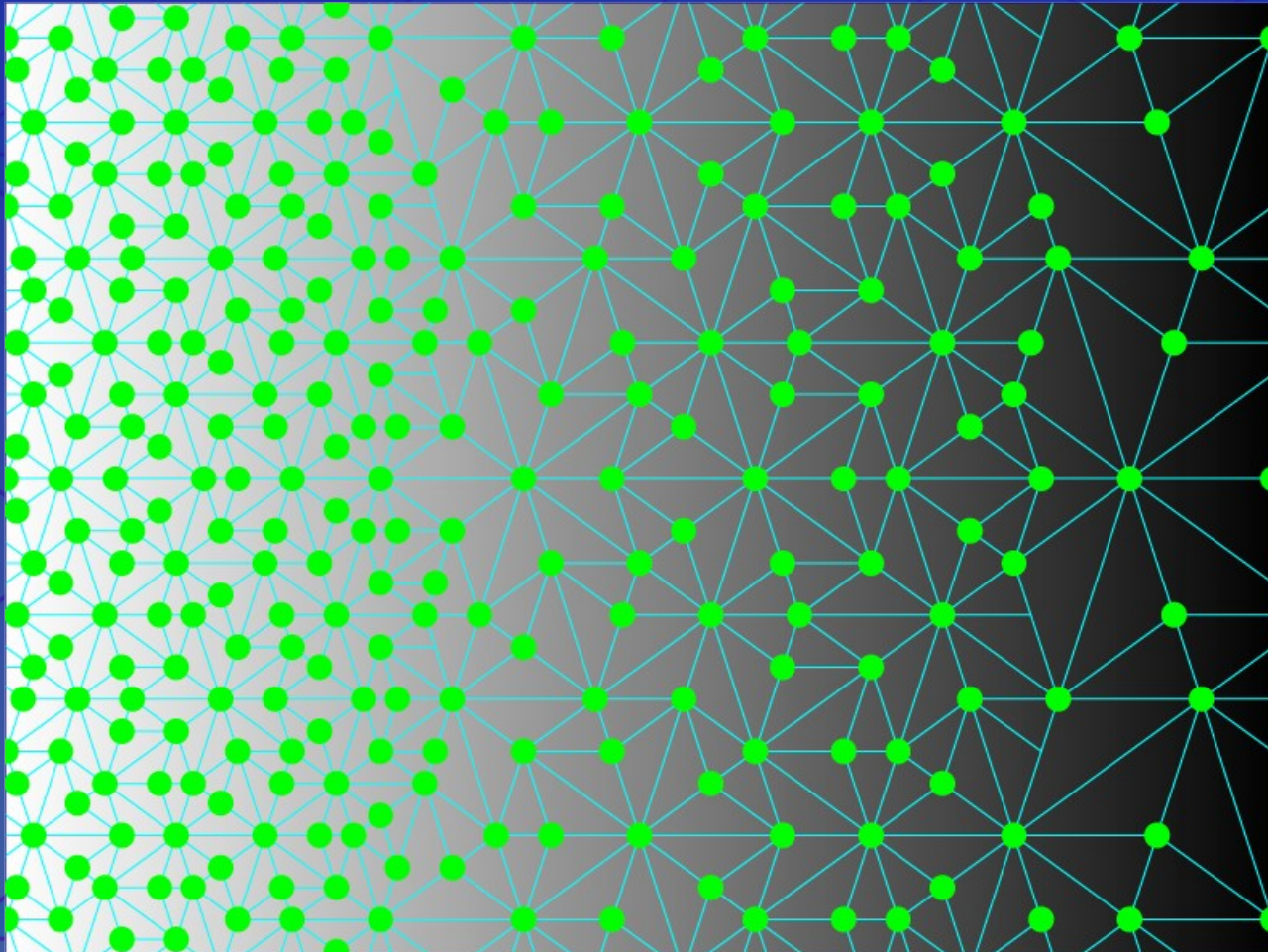
System Summary

Adaptive subdivision



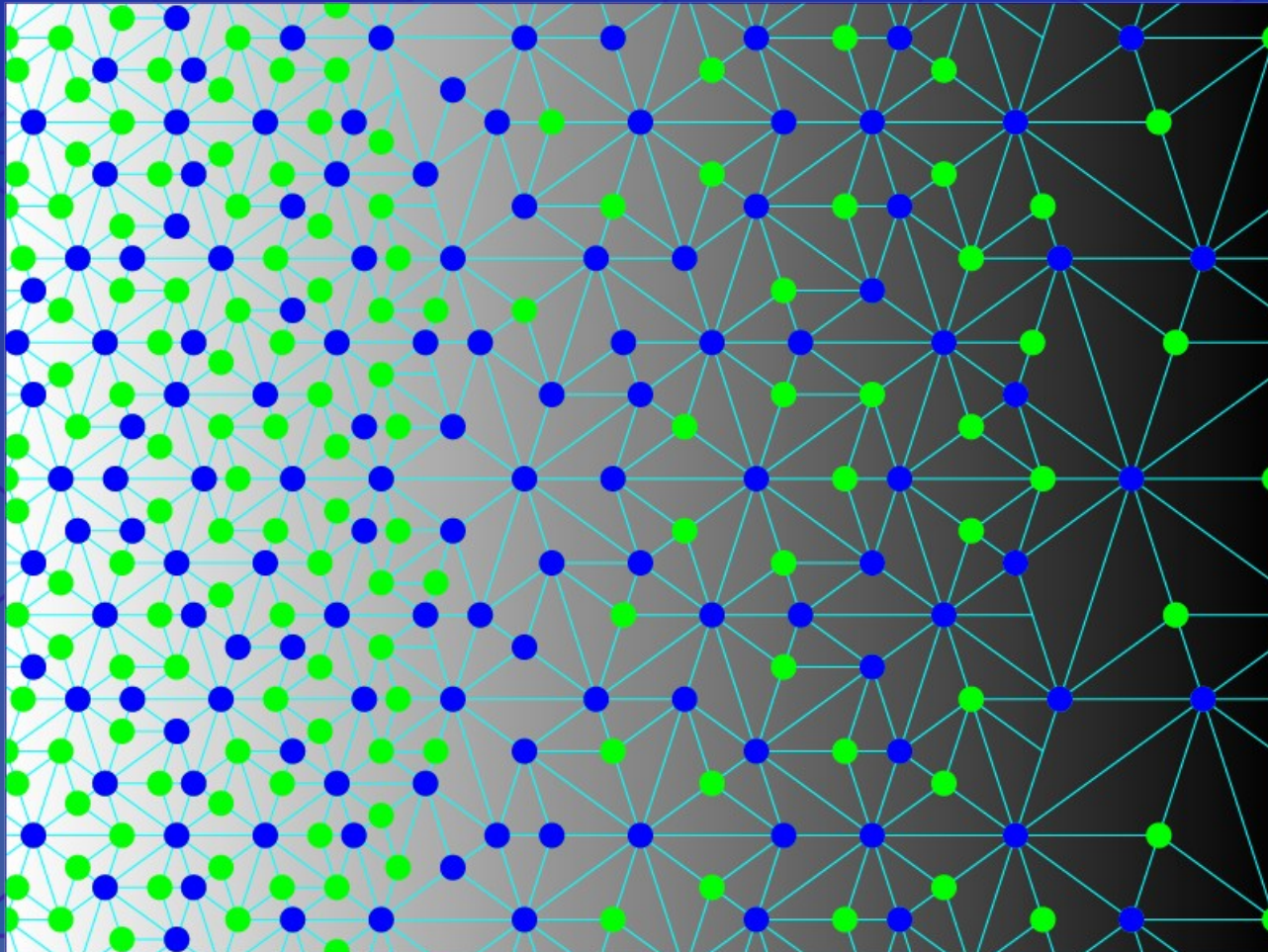
System Summary

'Sampling' tiles



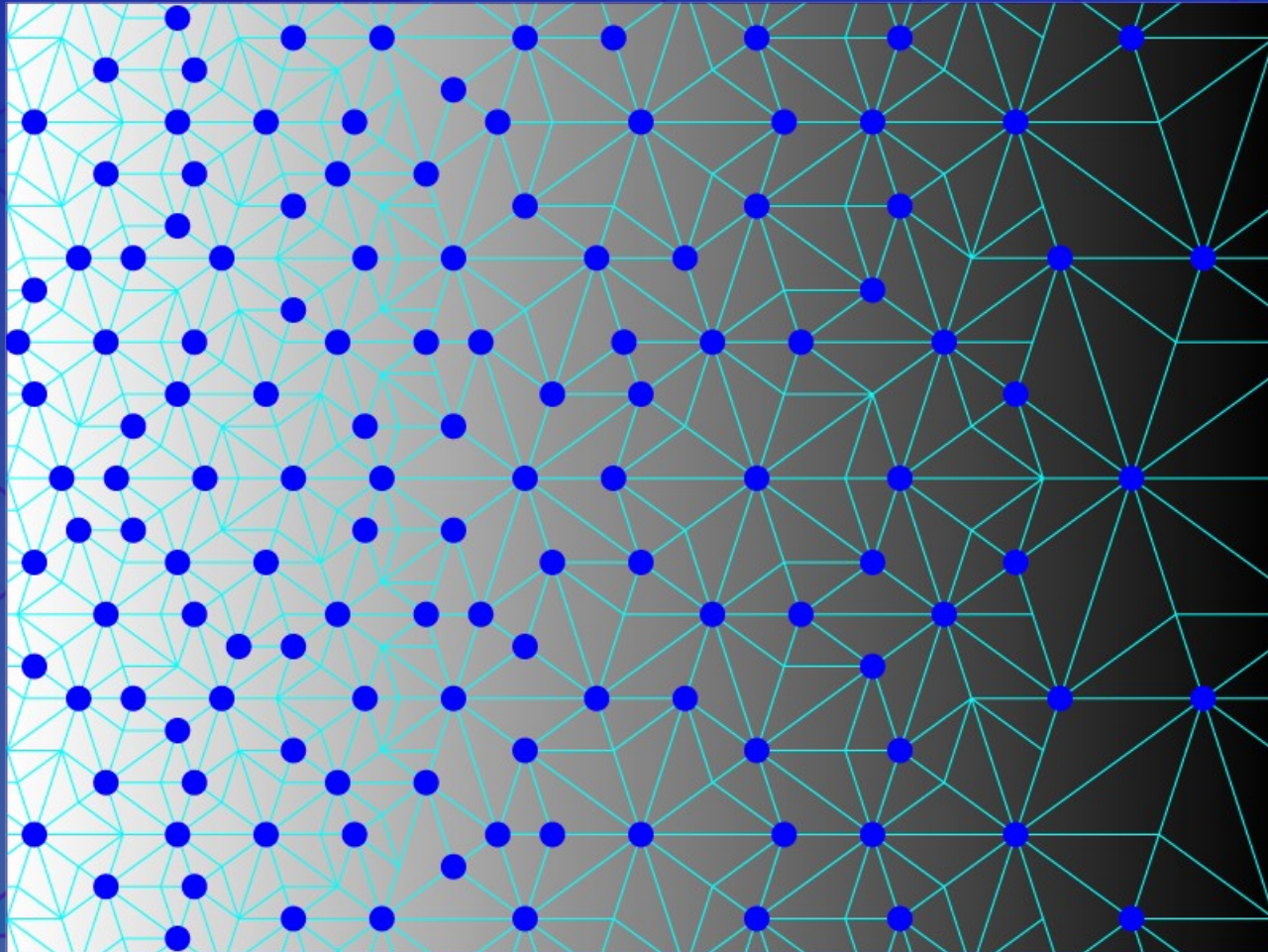
System Summary

Threshold with F-codes



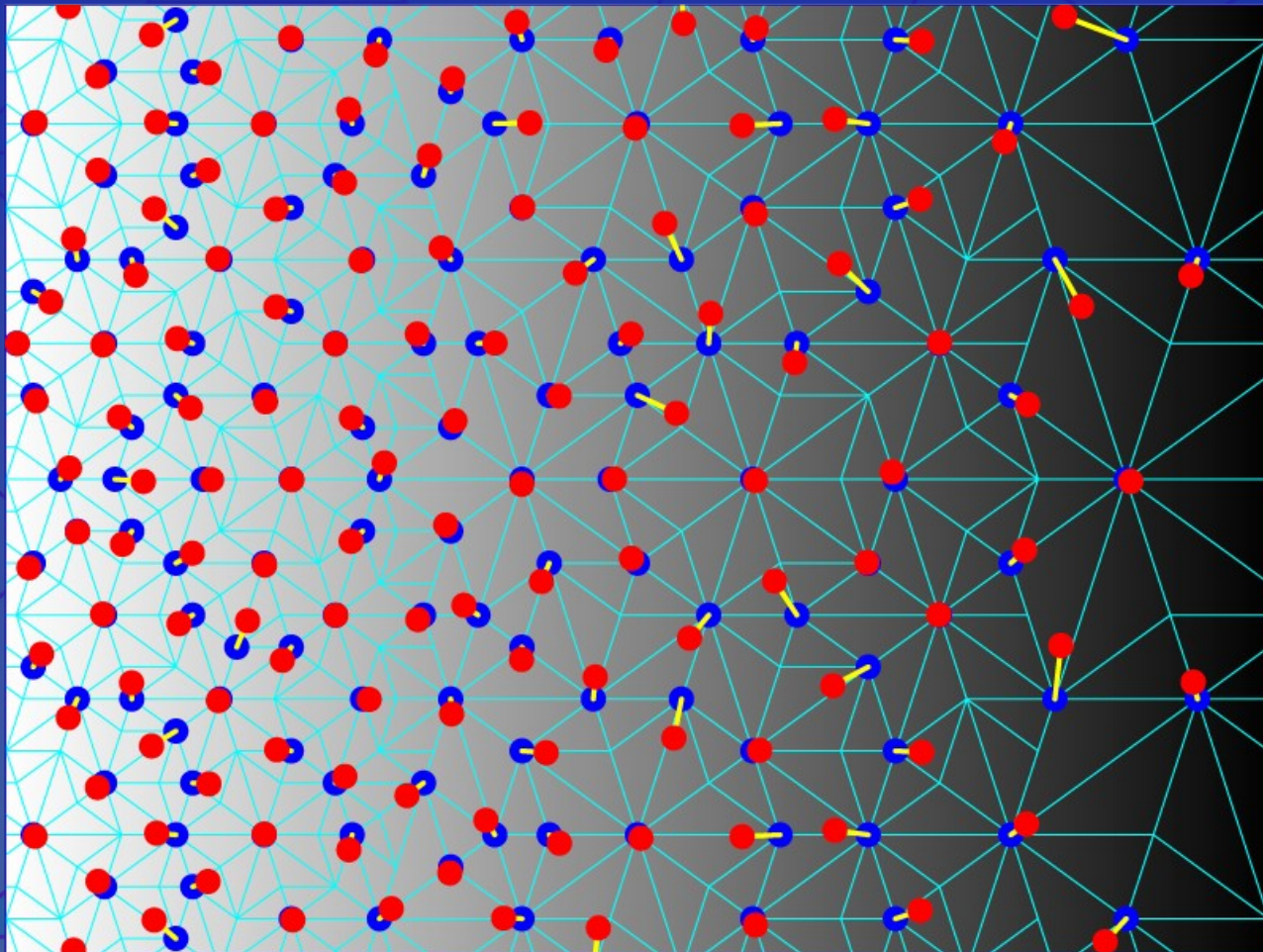
System Summary

Threshold with F-codes



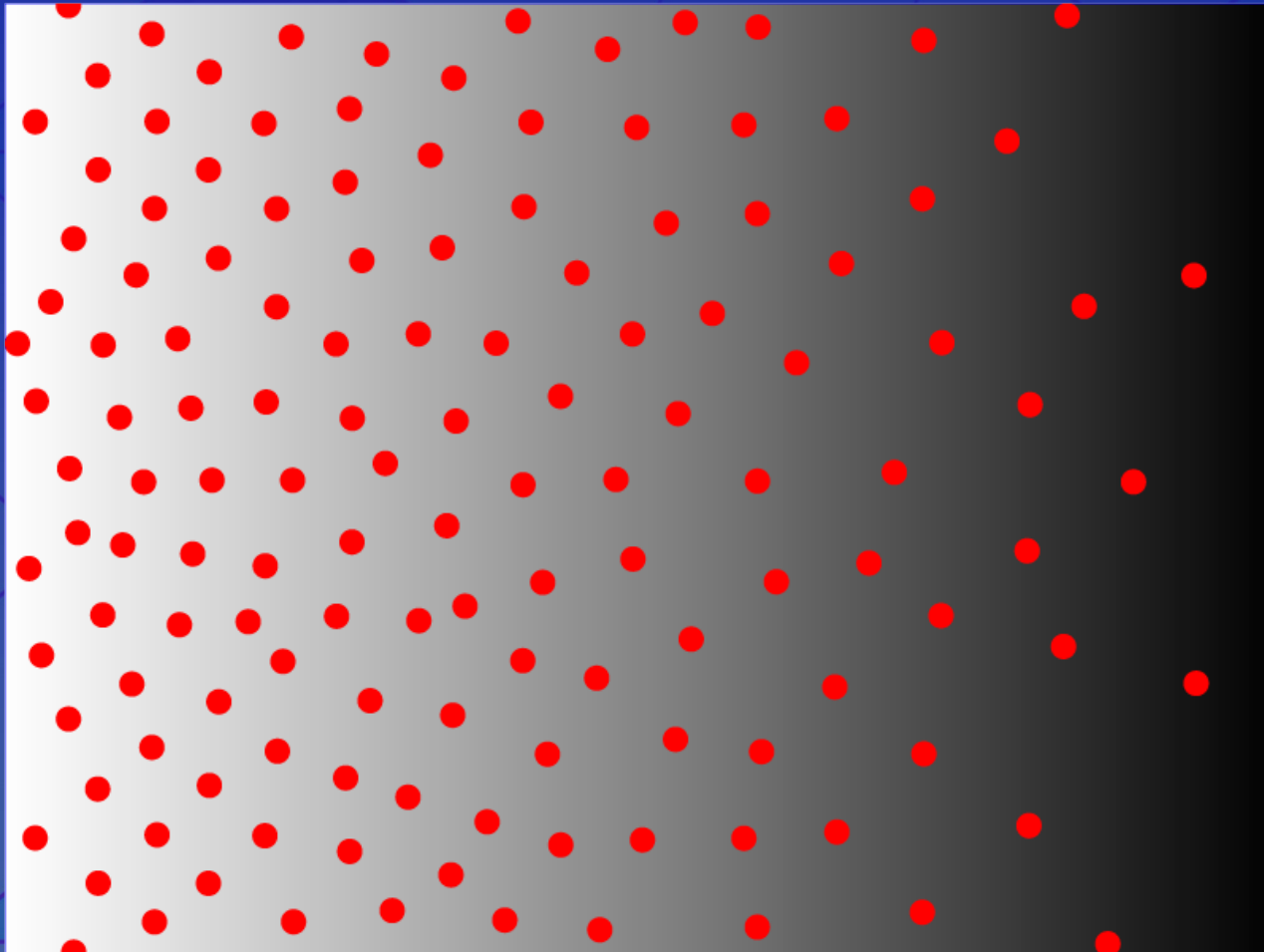
System Summary

Corrective Vectors



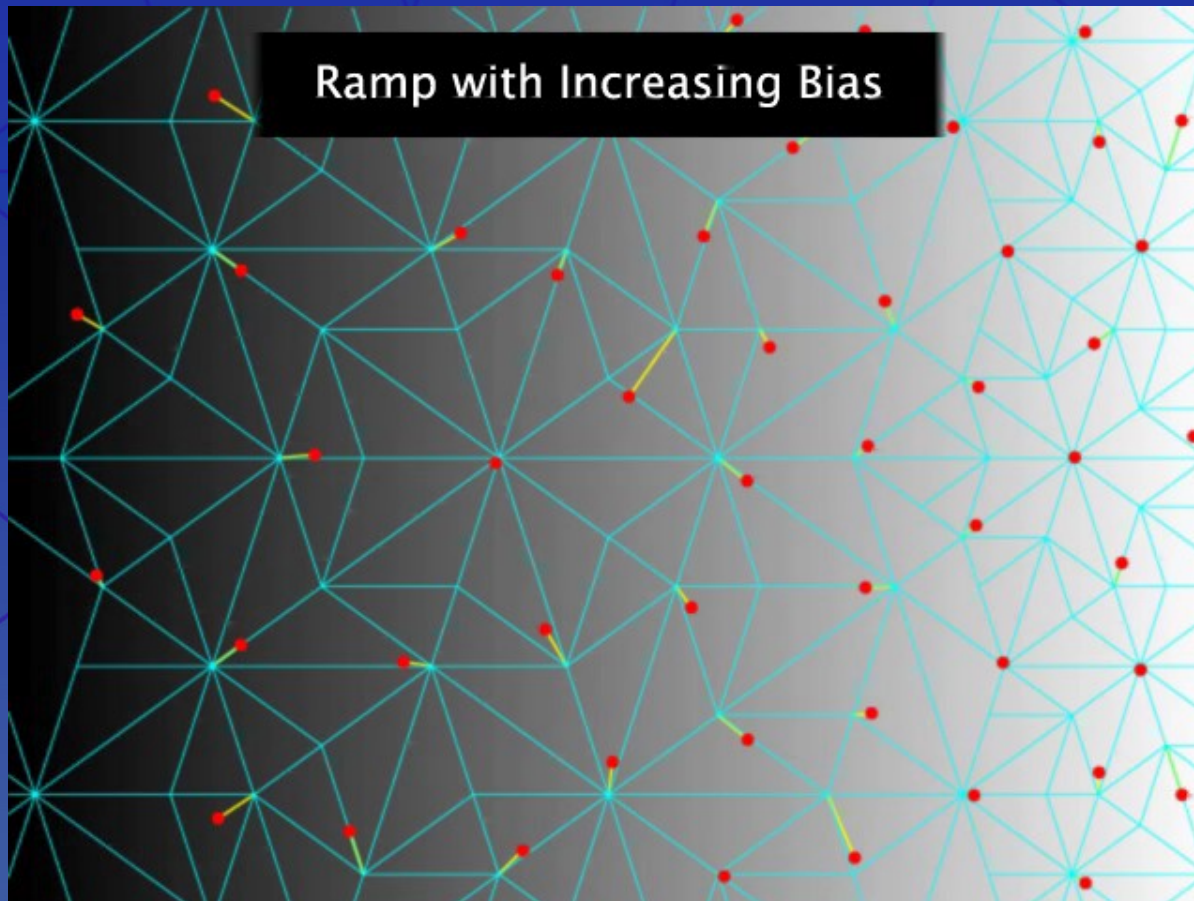
System Summary

Final Sampling



Results

Results





SIGGRAPH2004

Case Study: Environment Map Sampling

Results

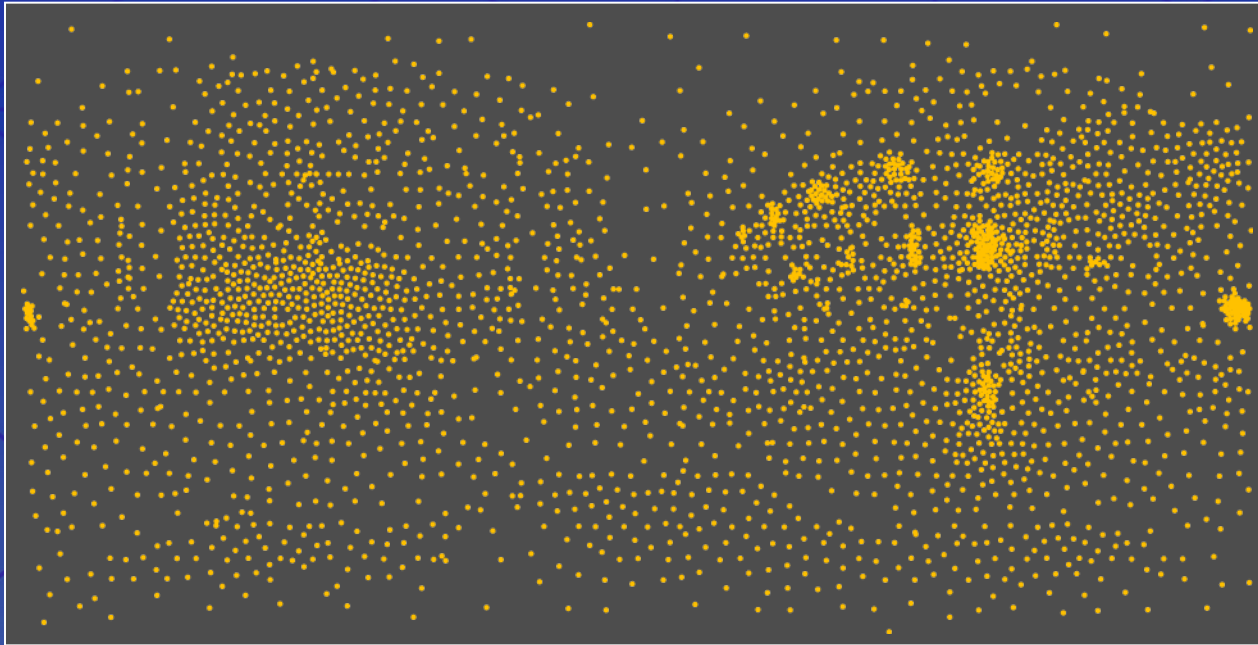


Image credits: Paul Debevec

Results

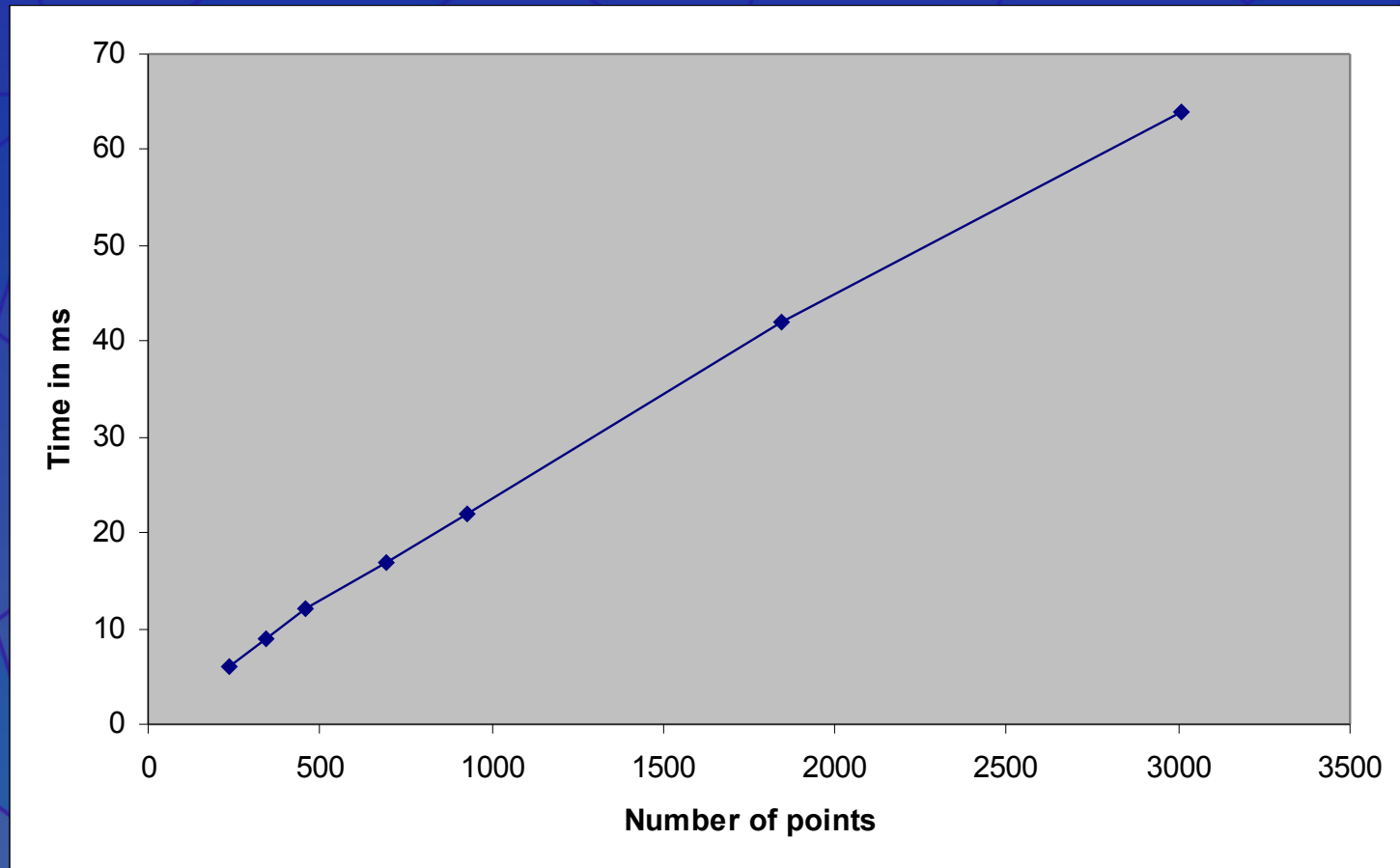


Results



Results

Timings



Polyominoes-based Sampling System [SIGGRAPH'2007]

Penrose tiling vs. Polyomino-based sampling

Penrose tiling

- *Deterministic production rules*
- *F-code attributes: integers in Fibonacci Number System*
- *Structural indices: local geometrical neighborhoods*
- *Simple lookup table-based run-time algorithm*

Idem

Polyominoes

- *Deterministic production rules*
- *Digital code attributes: integers in base ^A*
- *Structural indices: local geometrical neighborhoods*
- *Simple lookup table-based run-time algorithm*

Advantages

- *Less residual picks in Fourier amplitude spectrum*
- *Structure aligned with square grid*

*Improve-
ment*

D

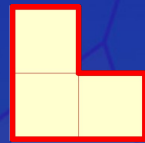
Polyominoes



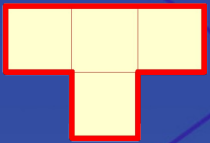
Monomino



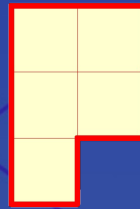
Domino



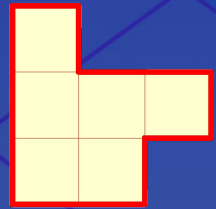
L-tromino



T-tetromino

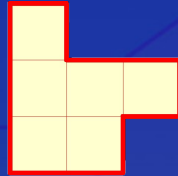


P-pentomino

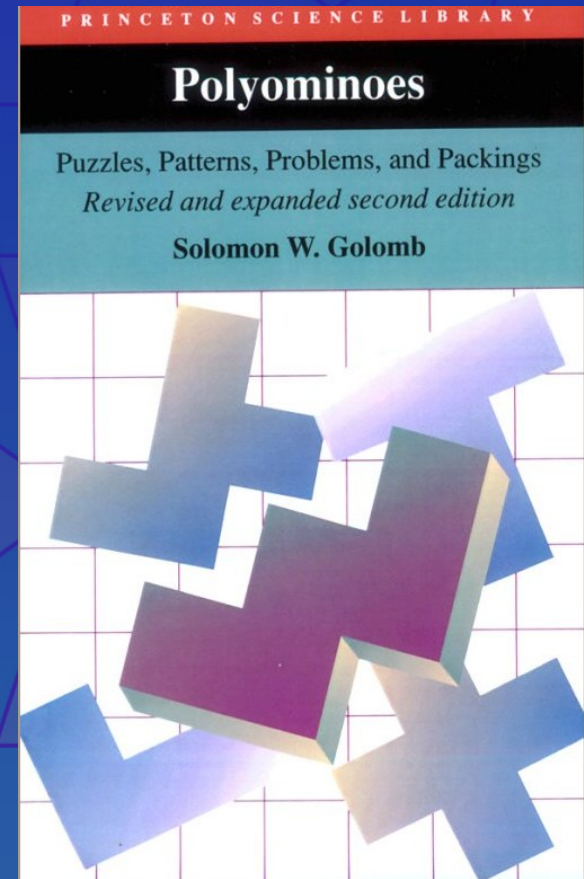
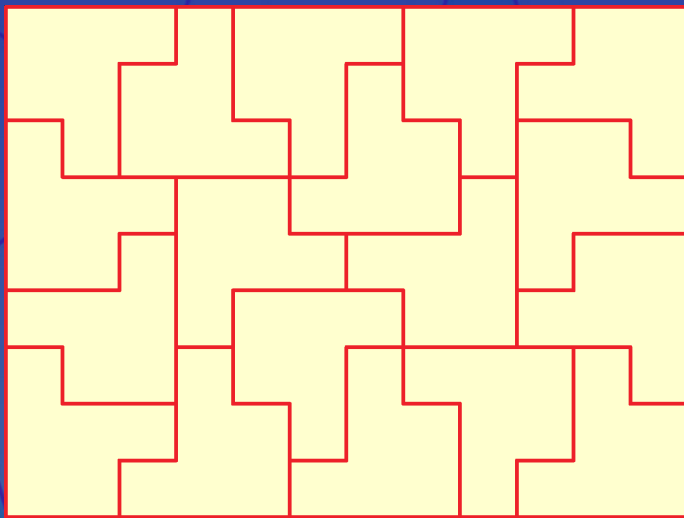


G-hexomino

Rectifiable polyominoes

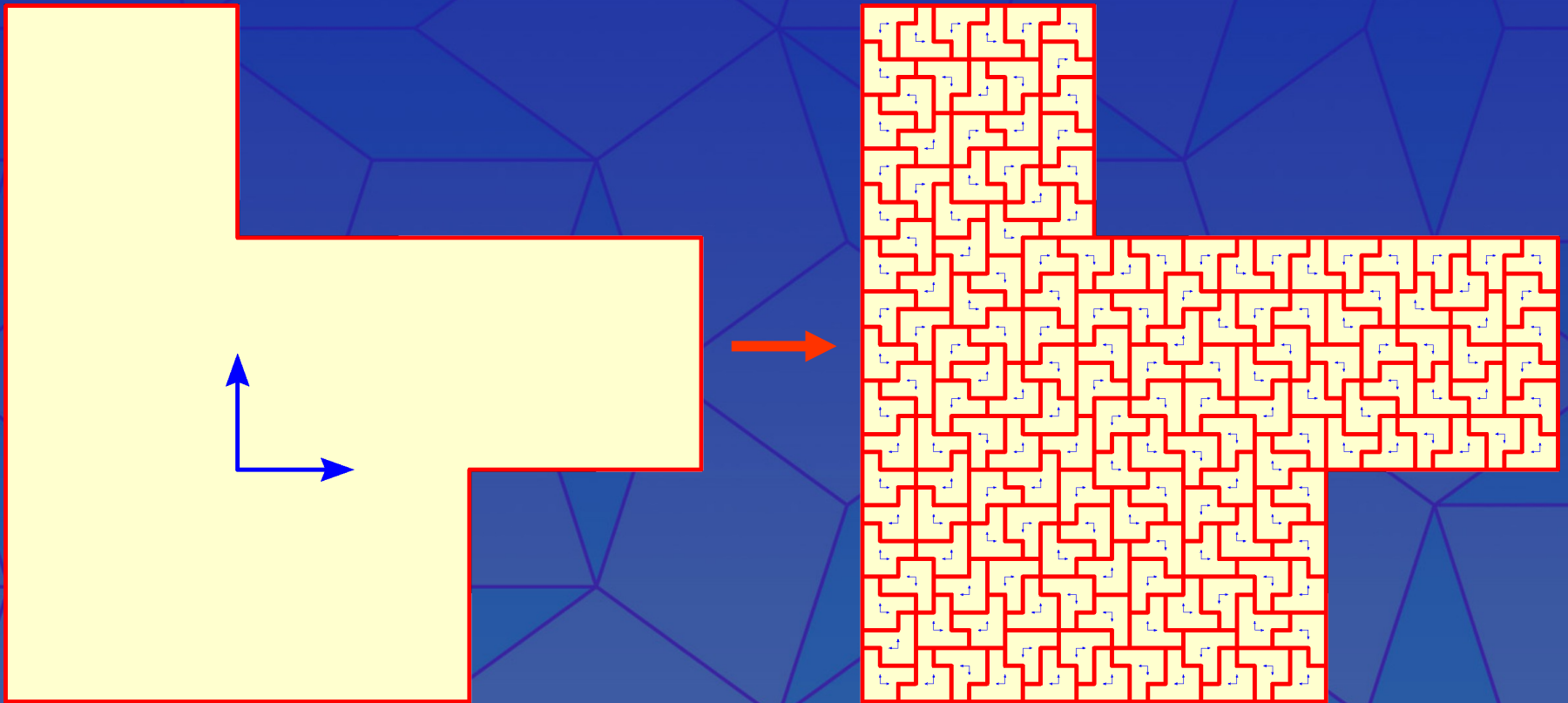


G-hexomino

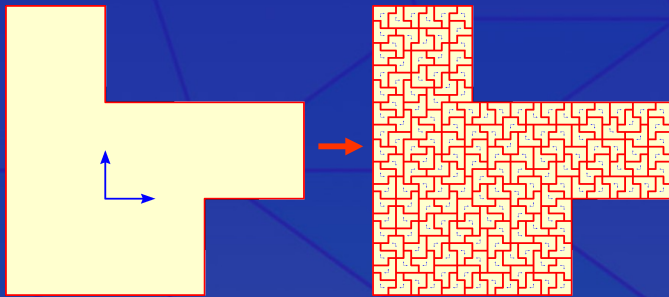


Self-similar ($\mathcal{A}$ -rep) rectifiable polyominoes, or reptiles

9^2 -rep or 81-rep G-hexominoes
Area scaling factor $\mathcal{A} = 9^2 = 81$

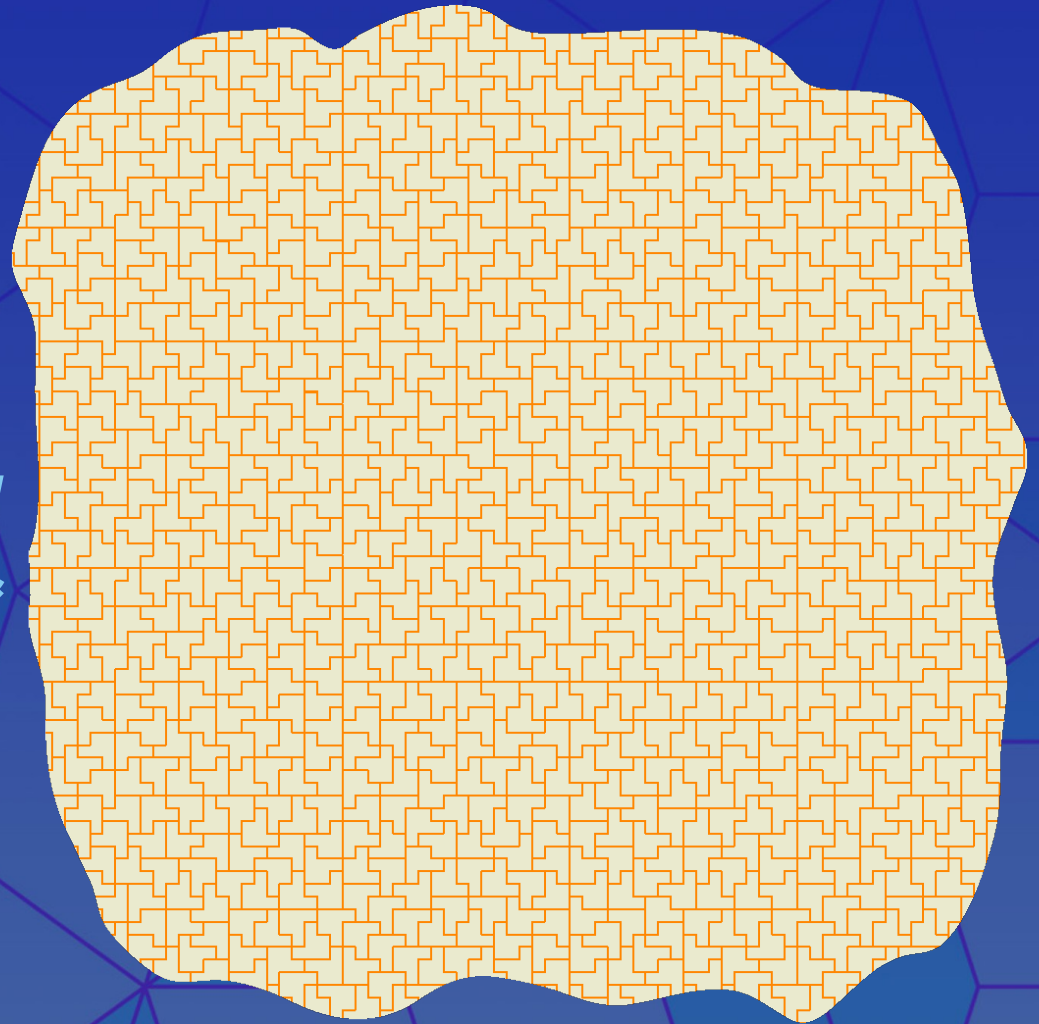


A-rep rectifiable polyominoes, or reptiles



→ *Unique production rule*
According to “*Tilings and Patterns*” [Grunbaum & Shephard ‘86]

- *Fills the entire plane non-periodically*
- *Self-similarity*

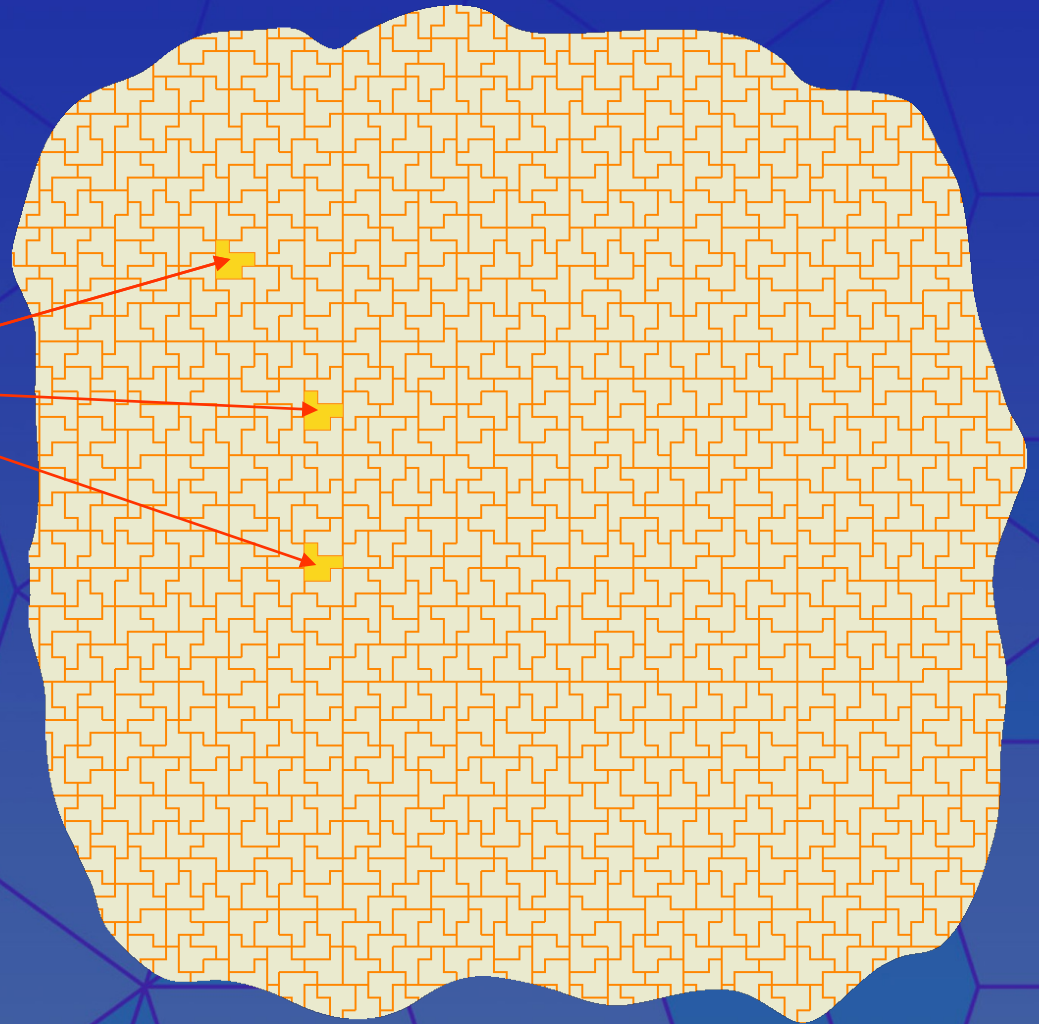


Infinite plane

81-rep G-hexominoes

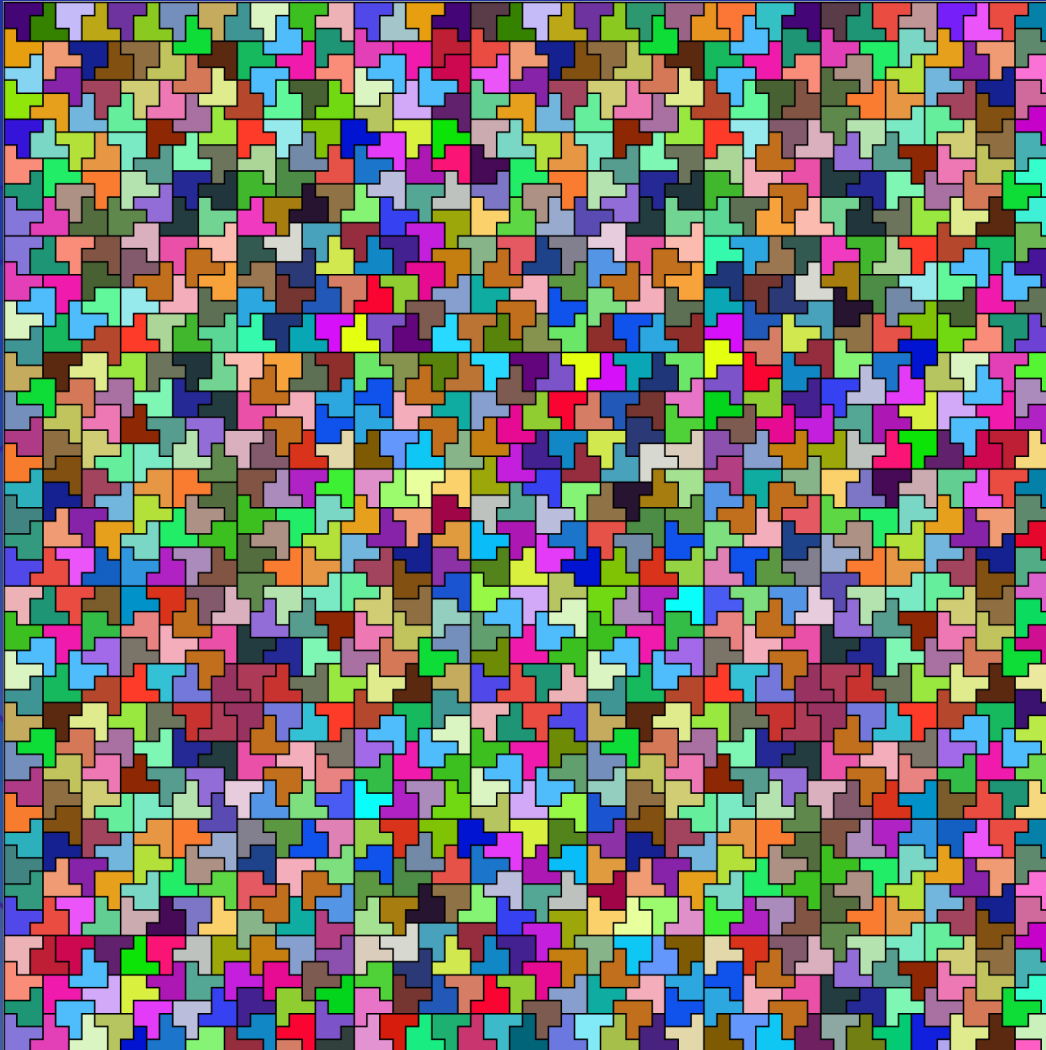
Different neighborhoods

→ *Problem of finding
geometrical neighborhoods
(structural indices)*



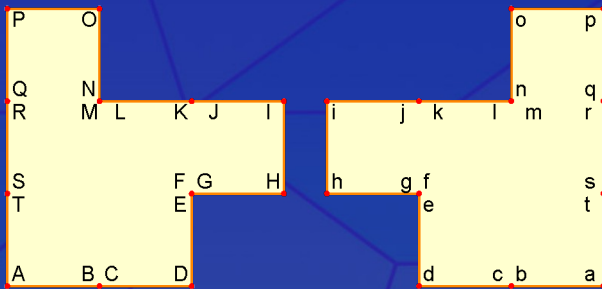
Infinite plane

Apparent Chaos



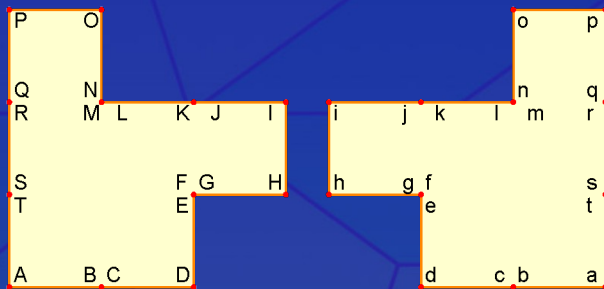
*Different colors =
Different neighborhoods =
Different structural indices*

Finding structural indices



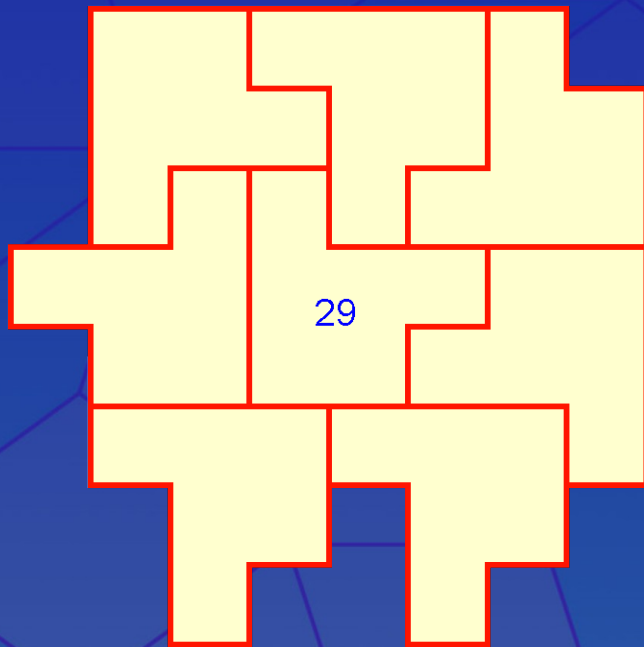
CB	Aa	bc	dp	qr	st	al	Hp	qr	s
.	Tt	.	eo	nm	.	bJ	Go	nm	
	Ss		fg	hl		cK	FE	DI	
LM	Rr	ml	kj	ik	fe	dL	.	Ck	f
AN	Qq	no	pP	Oj	gO	NM		Bj	g
TO	Pp	on	qQ	Ni	hP	QR	ST	AI	h
Si	j k	lm	rR	ML	KJ	ID	CB	AP	C
Rh	g f	.	sS	.	FG	HE	.	TQ	N
Qh	ie	.	tT	.	EH	GF	.	SR	M
Pg	j d	c b	aA	BC	DI	JK	LM	RS	
ef	kp	qr	st	ap	qr	st	aN	QT	
.	l o	nm	.	b o	nm	.	bO	PA	E
	mn	ol	.	ch	il	.	cA	TS	R
ts	r q	p k	f e	d g	j k	f e	d B	.	M
pa	b c	d j	g d	e f	k j	g l	H C	.	L
qt	.	e i	h c	.	l i	h J	GD	E F	K
rs	.	f g	h b	.	m n	o K	FE	D G	J

Finding structural indices



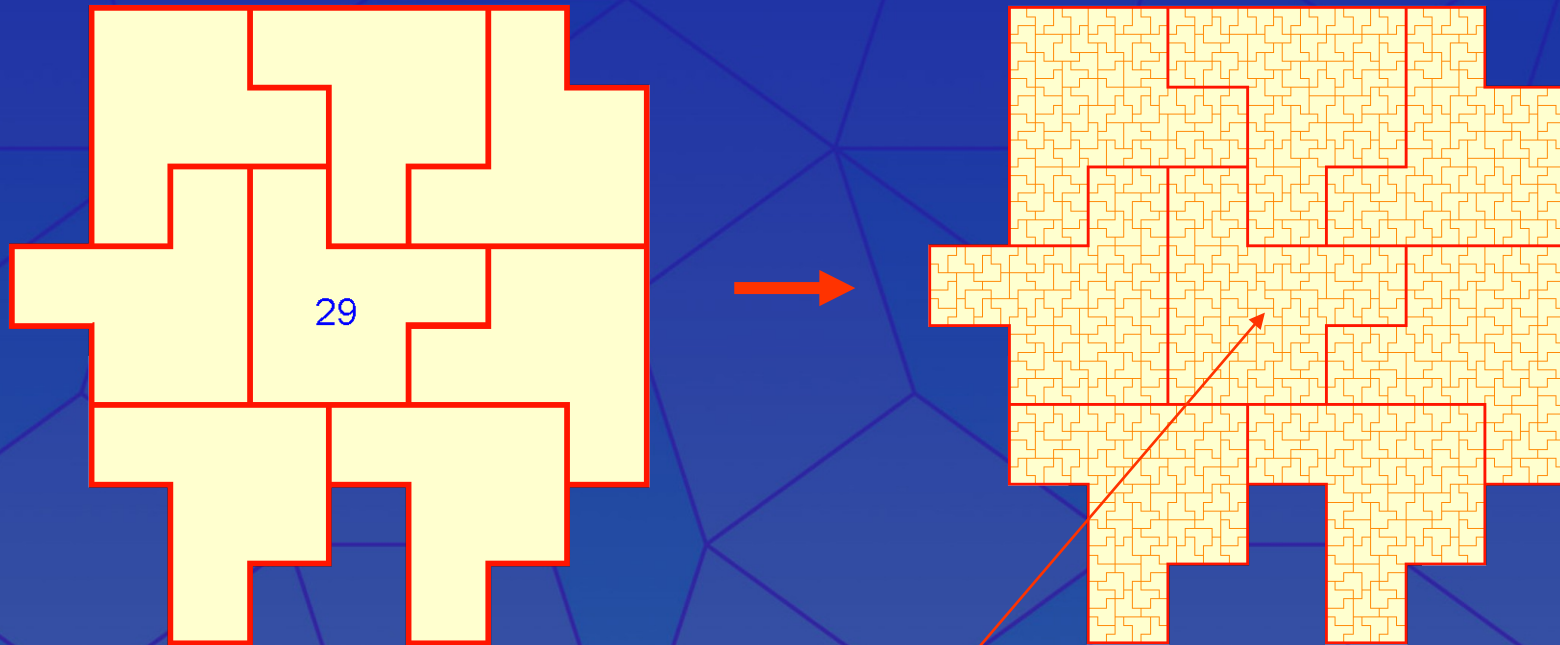
CB	Aa	bc	dp	qr	st	al	Hp	qr	s
219	Tt Ss	220	eo fg	nm hl	221	bJ cK	Go FE	nm DI	222
LM	Rr	ml	kj	ik	fe	dL	222	Ck	f
AN	Qq	no	pP	Oj	gO	NM		Bj	g
TO	Pp	on	qQ	Ni	hP	QR	ST	Ai	h
Si	j k	lm	rR	ML	KJ	ID	CB	AP	C
Rh	g f	28	sS	29	FG	HE		TQ	N
Qh	ie		st	T	EH	GF	30	SR	M
Pig	j d	c b	aA	BC	DI	JK	LM	RS	
ef	kp	qr	st	ap	qr	st	aN	QT	
21	lo	nm	22	bo	nm	23	bo	PA	E
	mn	ol		ch	il		ca	TS	R
ts	rg	pk	fe	dg	jk	fe	dB		M
pa	bc	dj	gd	ef	kl	gl	HC	212	L
qt		ei	hc	18	li	hJ	GD	EF	K
rs	17	fg	hb		mn	oK	FE	DG	J

Finding structural indices



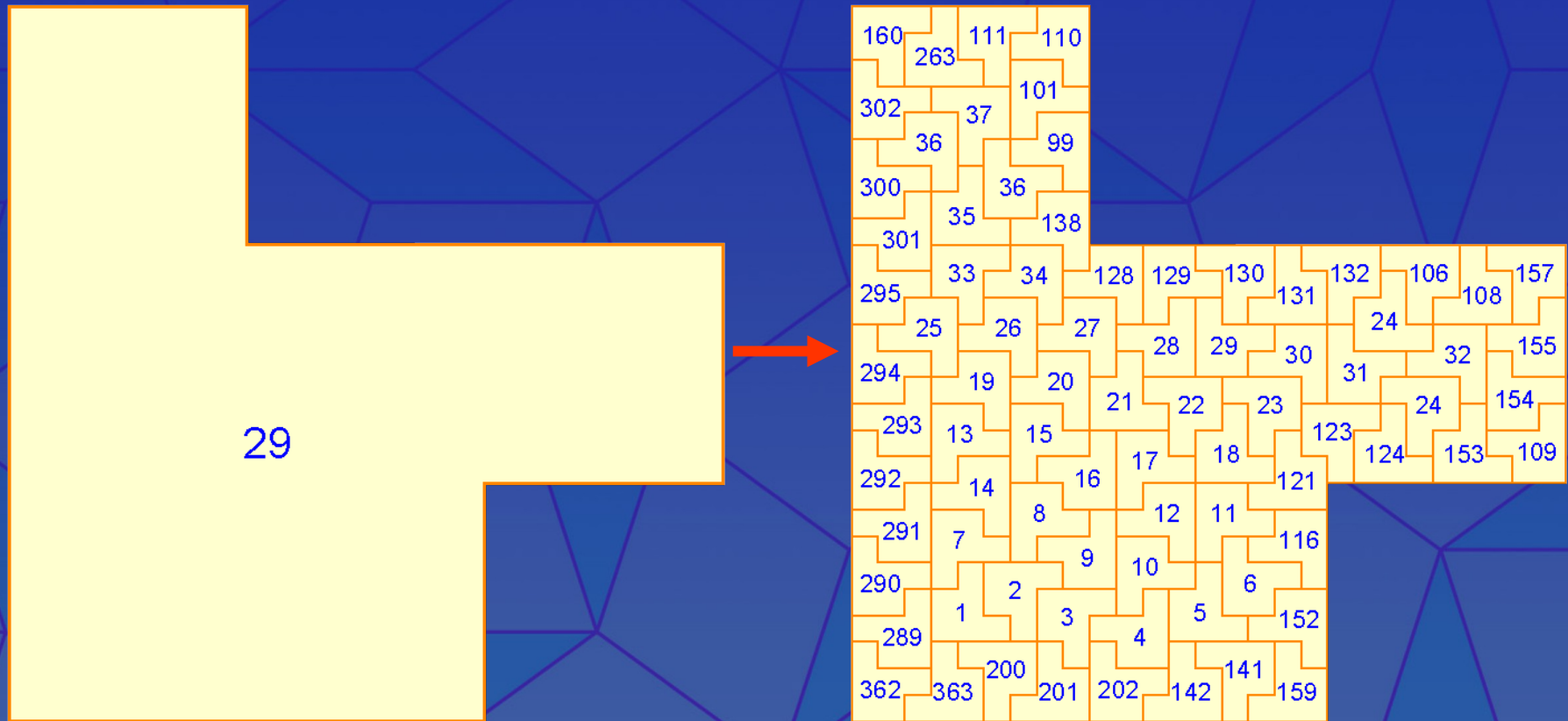
CB	Aa	bc	dp	qr	st	al	Hp	qr	s
219	Tt Ss	220	eo fg	nm hl	221	bJ cK	Go FE	nm DI	222
LM AN	Rr Qq	ml no	kj pP	ik Oj	fe gO	dL NM	222	Ck Bj	f g
TO Si	Pp jk	on lm	qQ rR	Ni ML	hP KJ	QR ID	ST CB	Ai AP	h C
Rh Qh	gf ie	28	sS tT	29	FG EH	HE GF	30	TQ SR	N M
Pg ef	jd kp	cb qr	aA st	BC ap	DI qr	JK st	LM aN	RS QT	
21	lo mn	nm ol	22	bo ch	nm il	23	bo cA	PA TS	E R
ts pa	rg bc	pk dj	fe gd	dg ef	jk kj	fe gl	dB HC	212	M L
qt rs	17	ei fg	hc hb	18	li mn	hJ oK	GD FE	EF DG	K J

Finding structural indices



*Well-determined
neighborhoods*

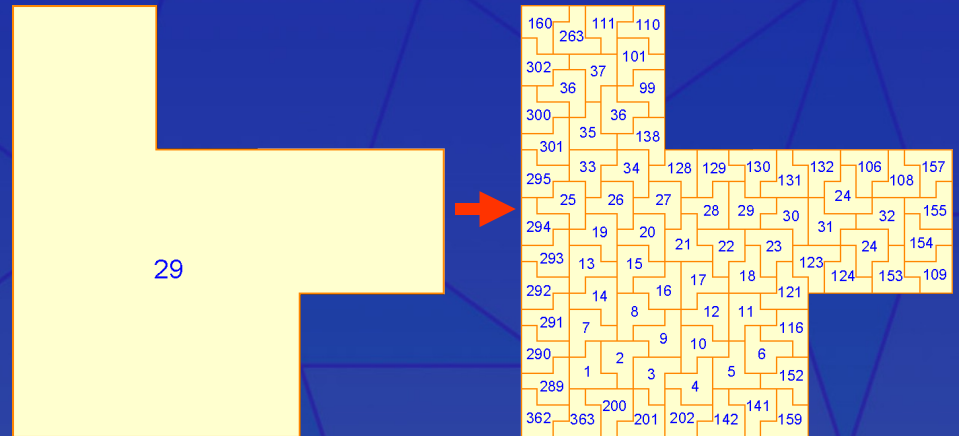
Finding structural indices



Optimization

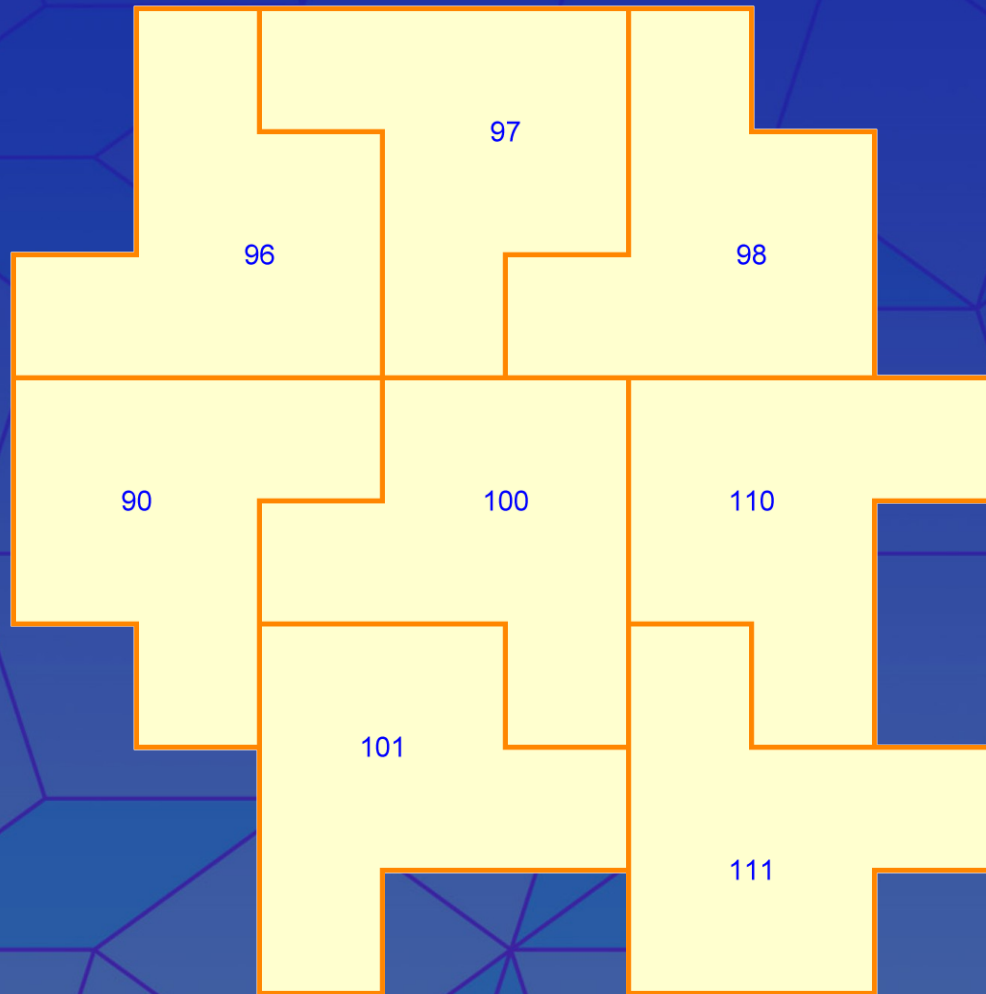
Decomposition process:

- *Production rules*
 - *Structural indices characterize local geometrical neighborhoods*
 - *Mapping of structural indices (finite number)*
- *Tiles with similar geometrical neighborhoods (structural indices) behave similarly*



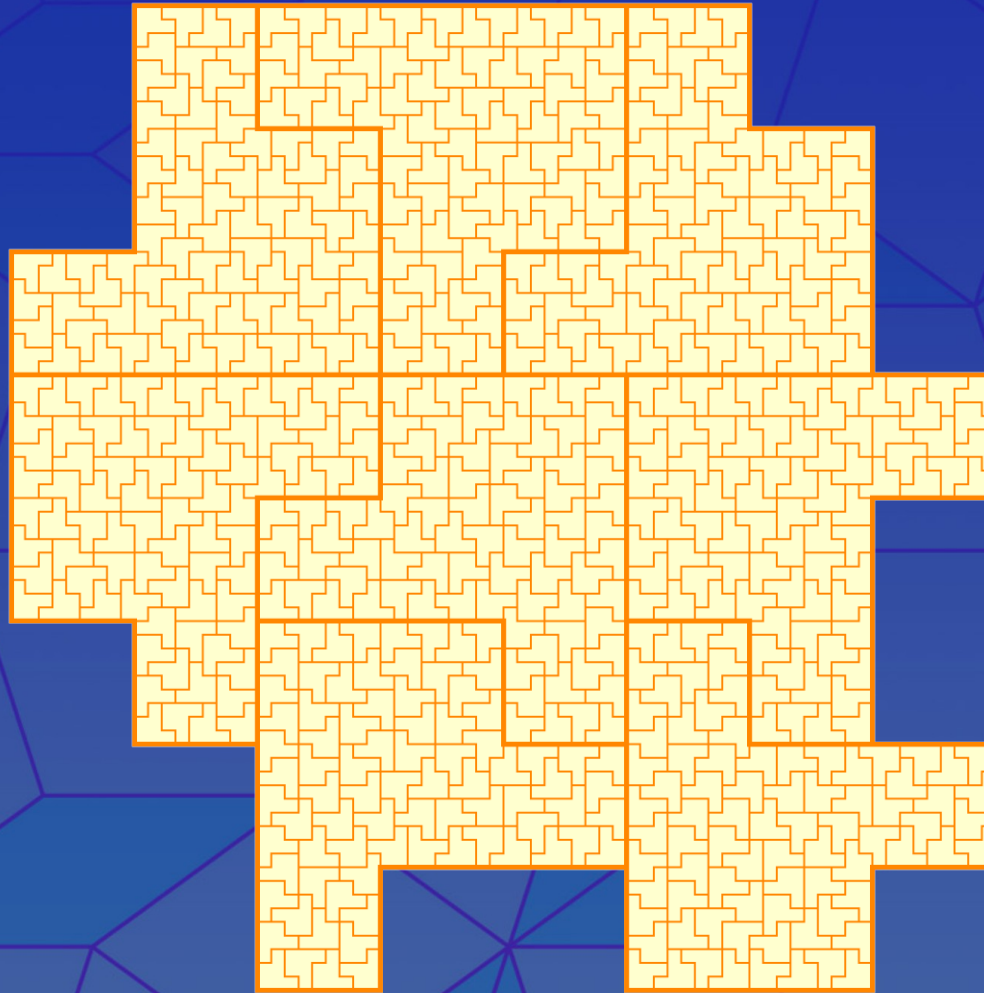
Optimization I

Patch



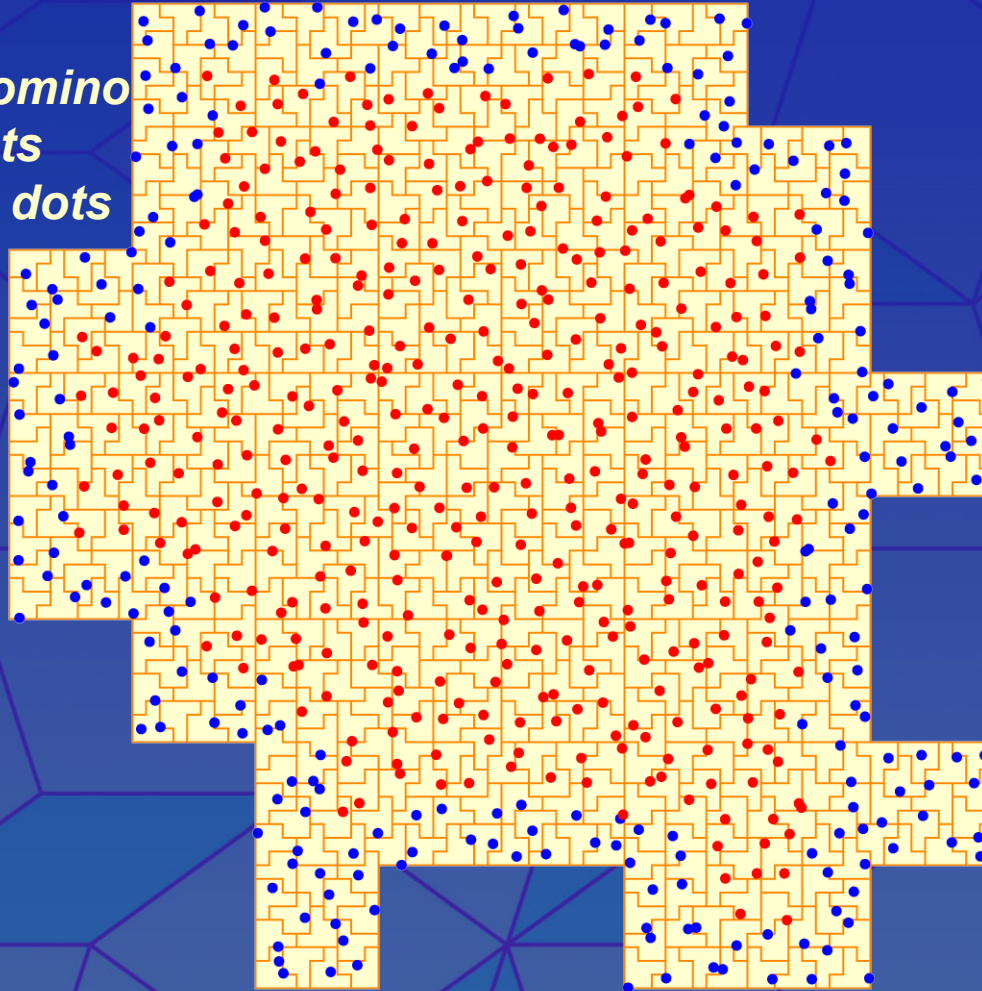
Optimization I

Subdivision



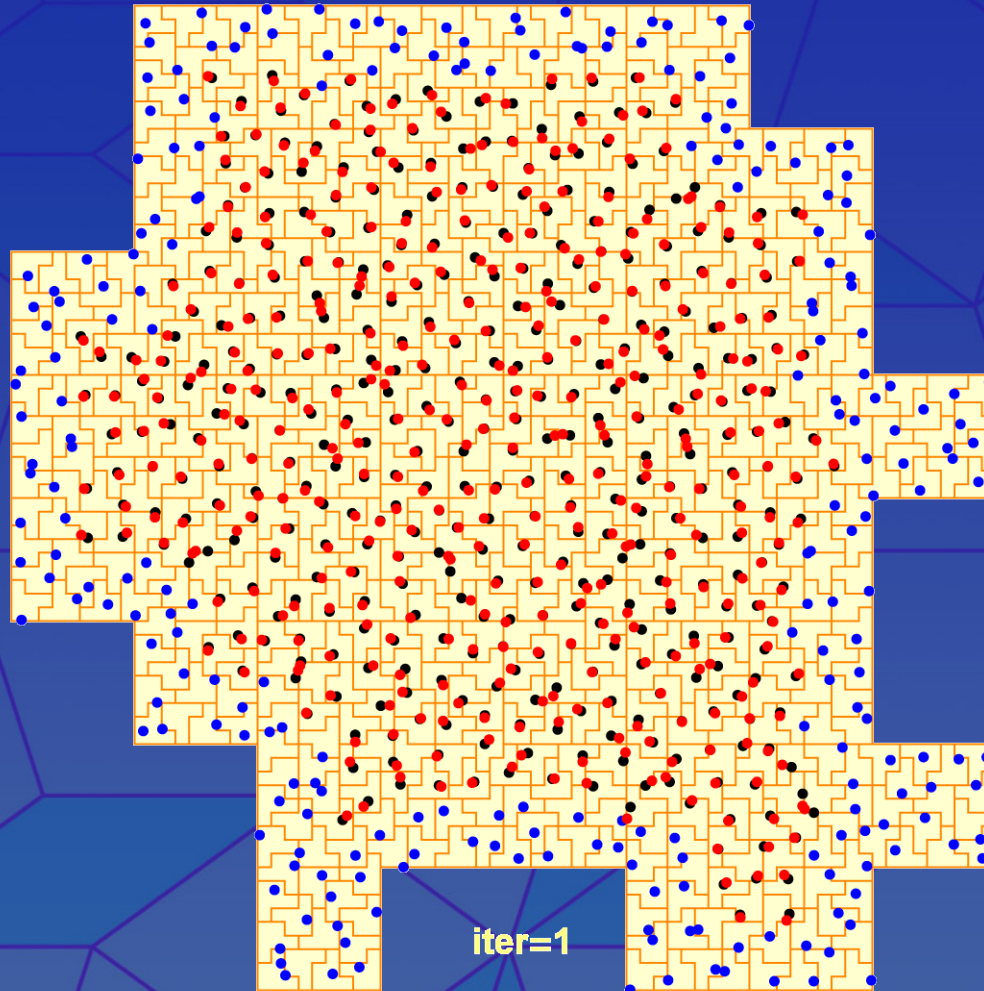
Optimization I

*Random dots
within each polyomino
Red: movable dots
Blue: immovable dots*



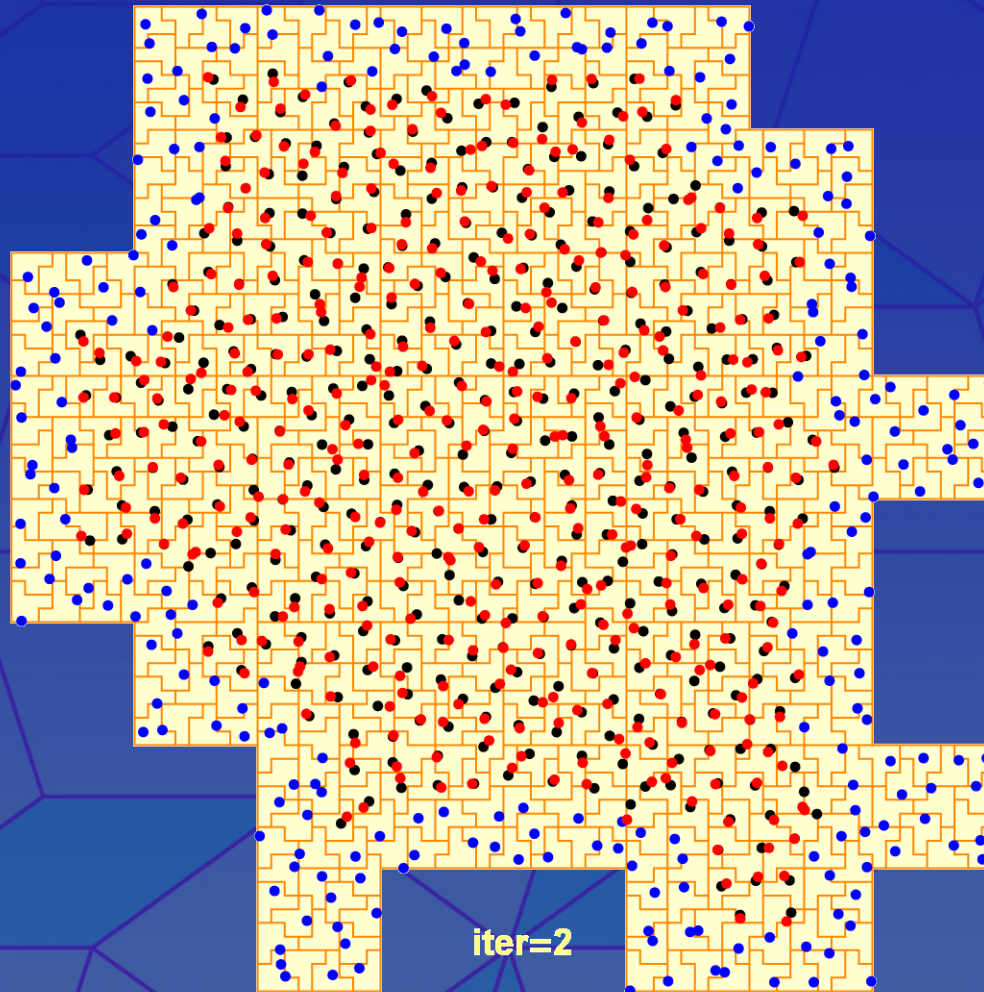
Optimization I

Relaxation



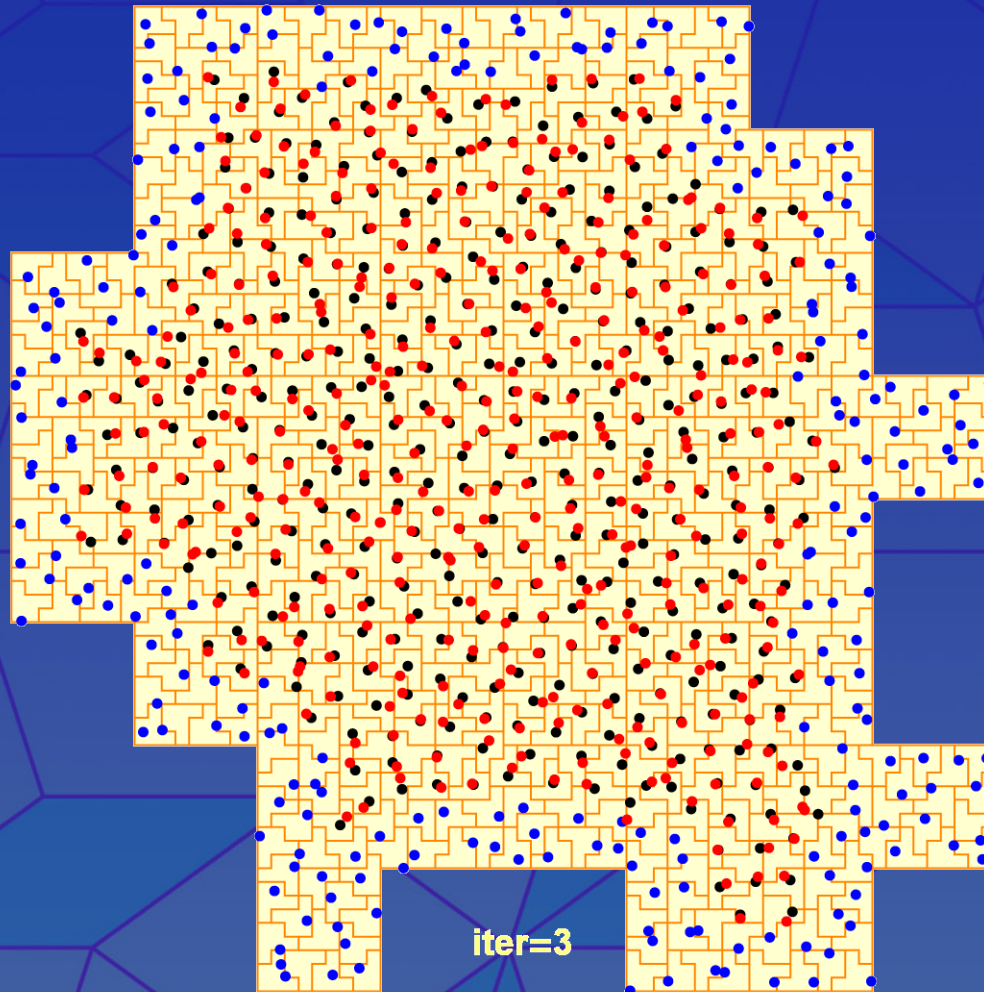
Optimization I

Relaxation



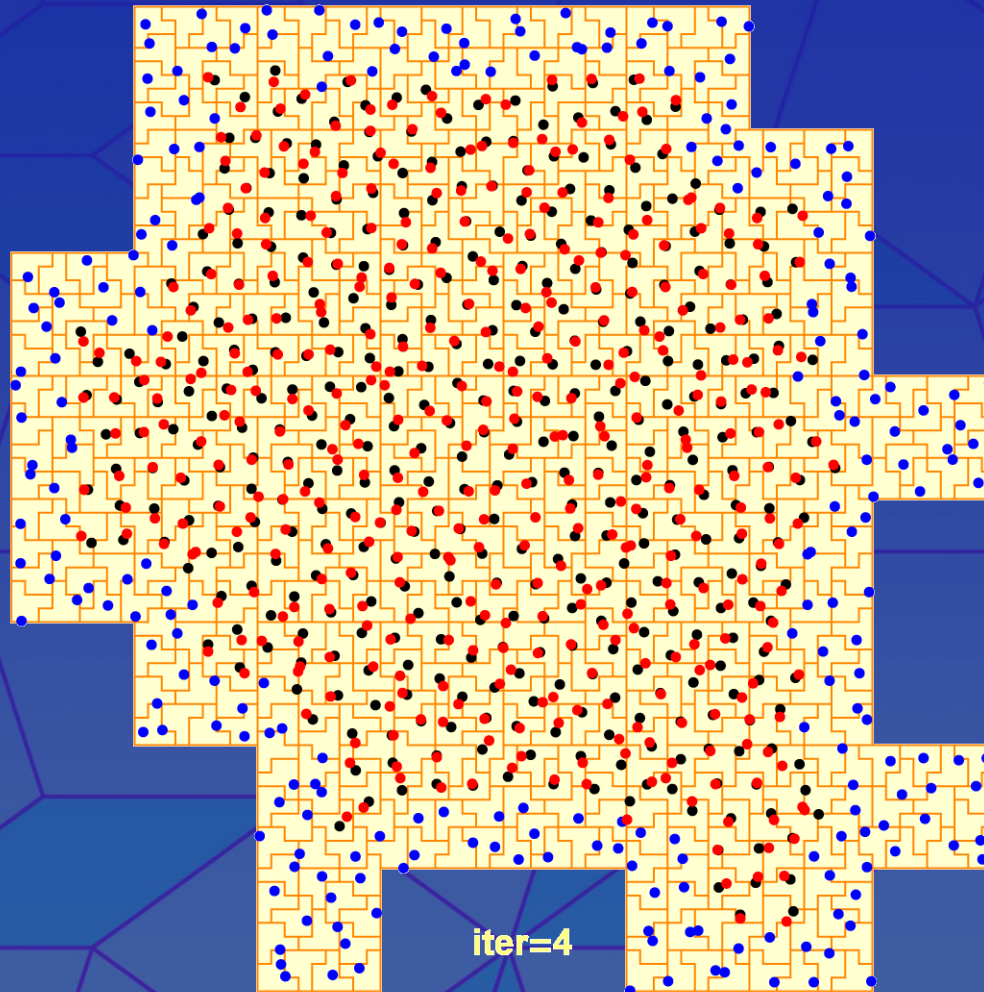
Optimization I

Relaxation



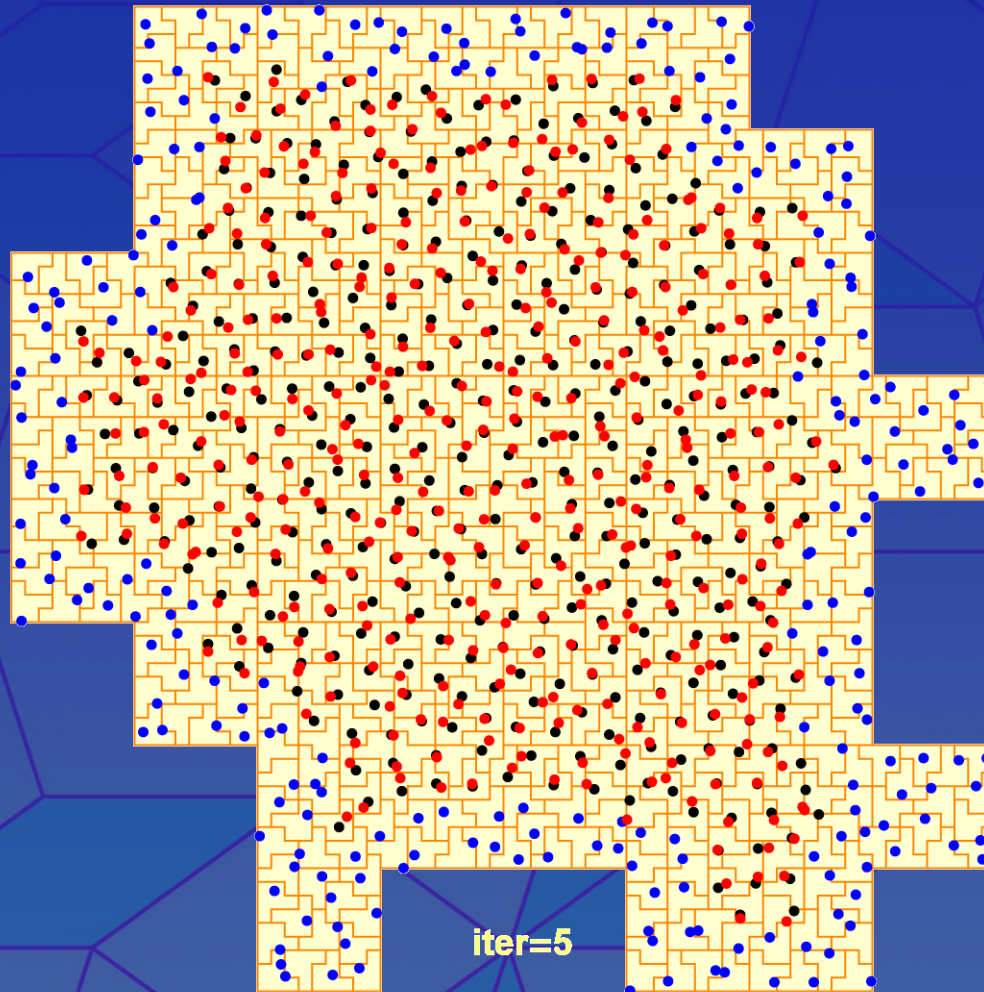
Optimization I

Relaxation



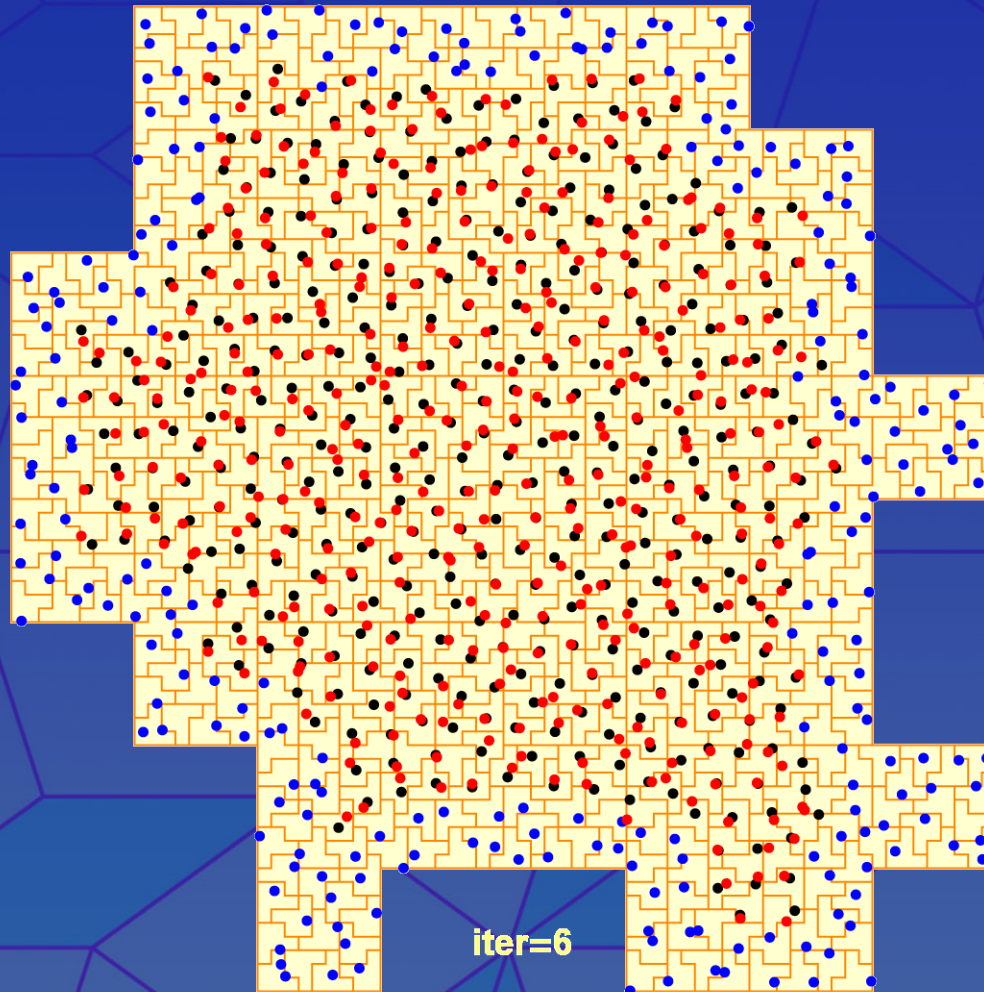
Optimization I

Relaxation



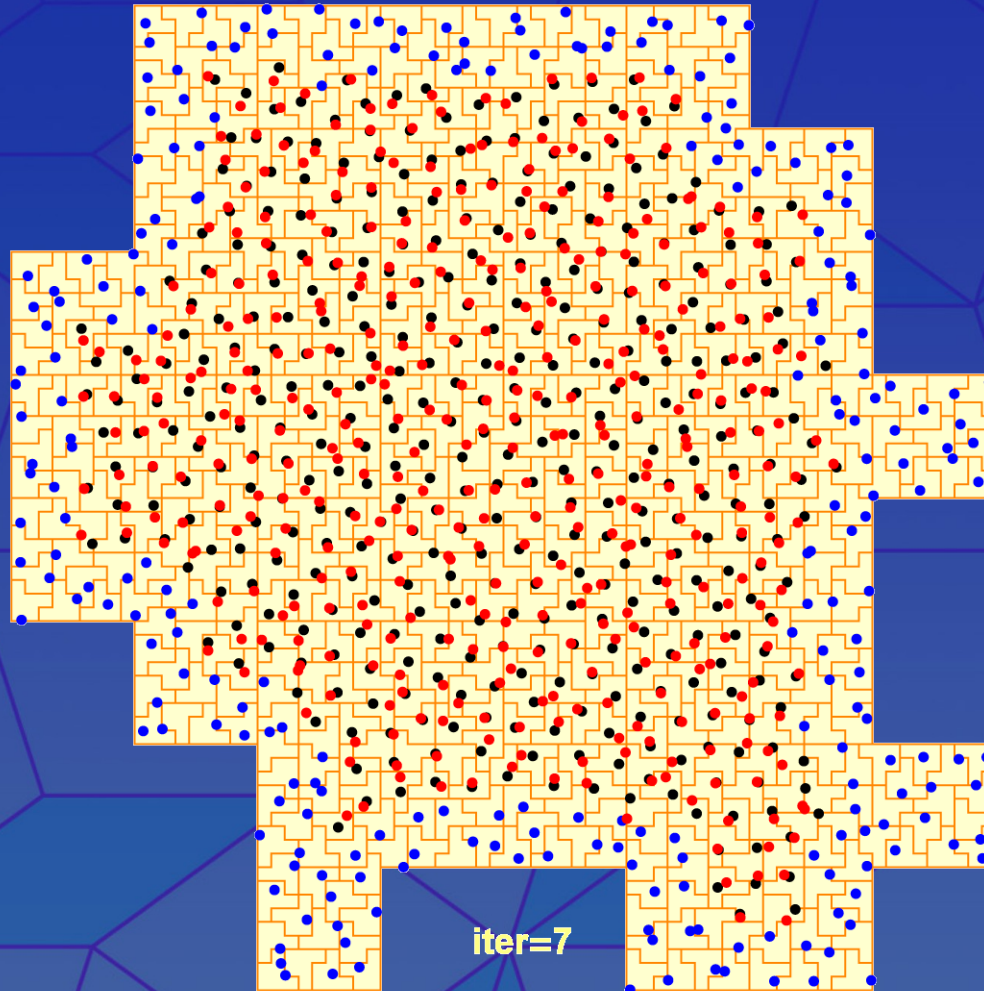
Optimization I

Relaxation



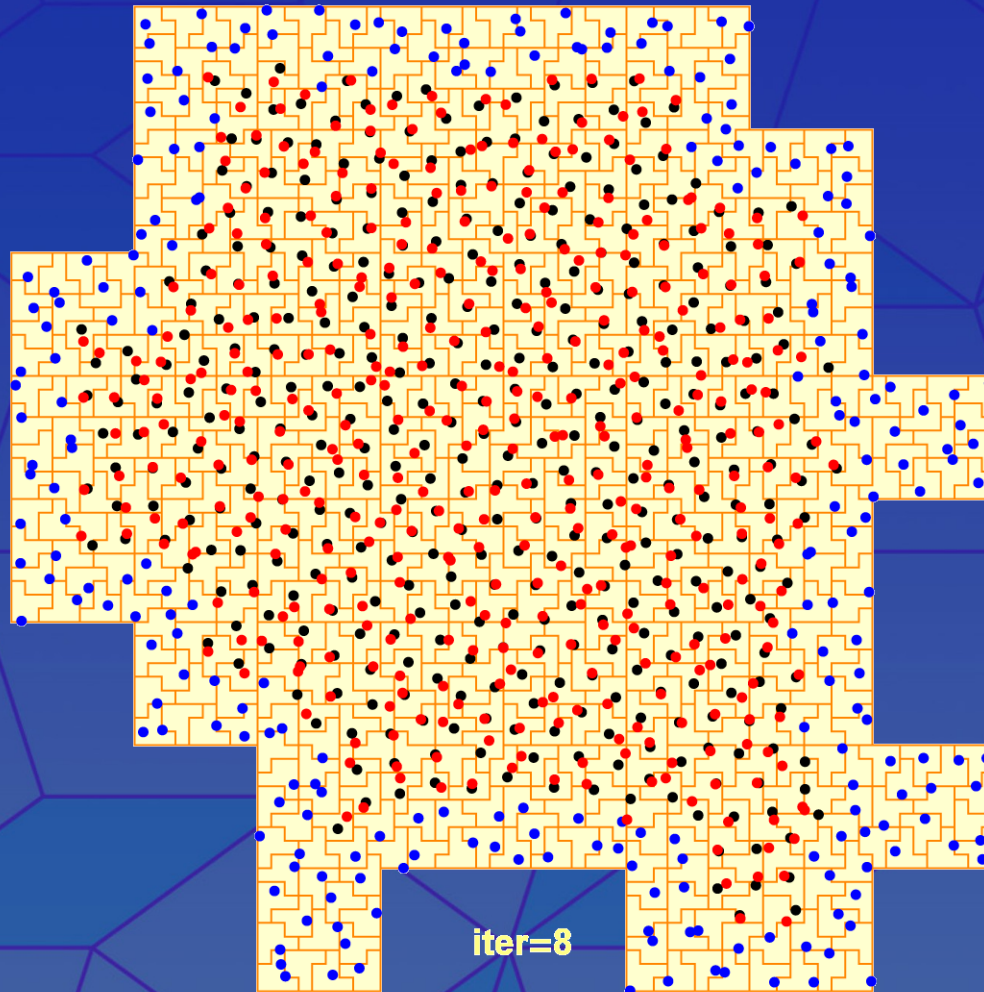
Optimization I

Relaxation



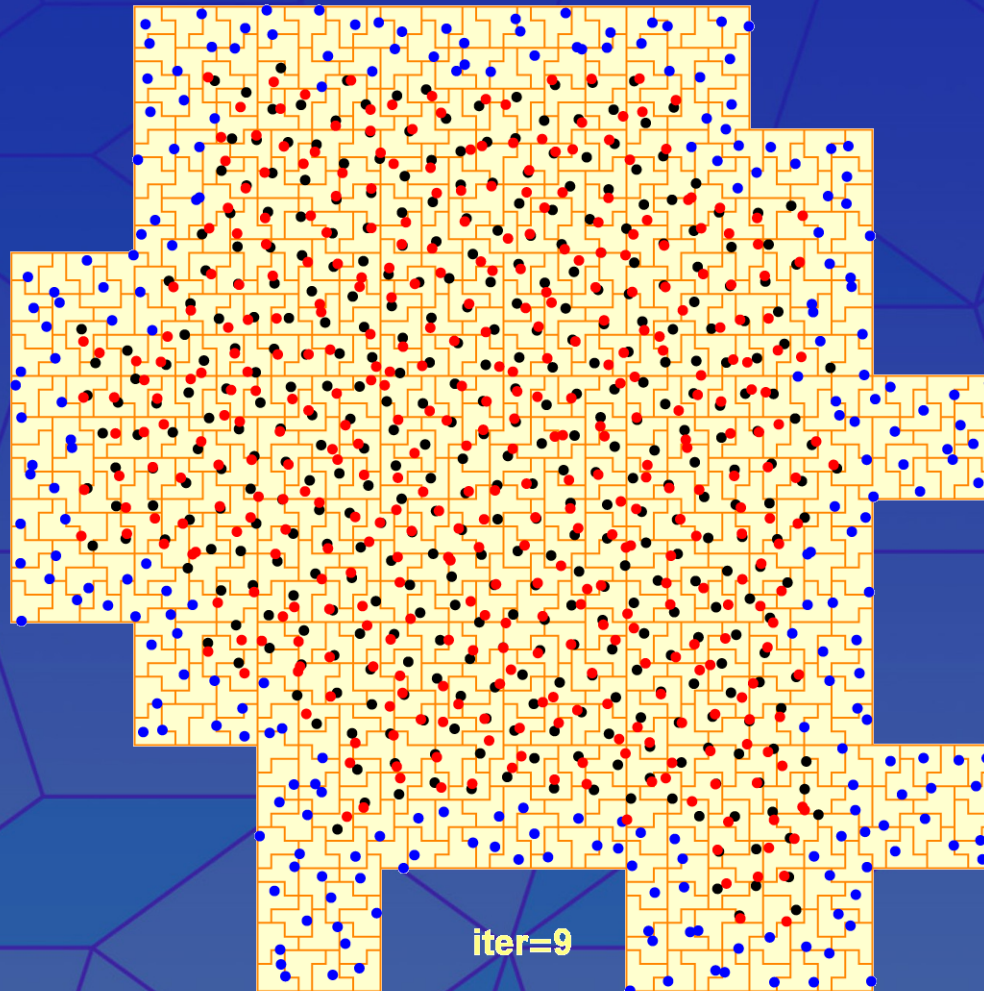
Optimization I

Relaxation



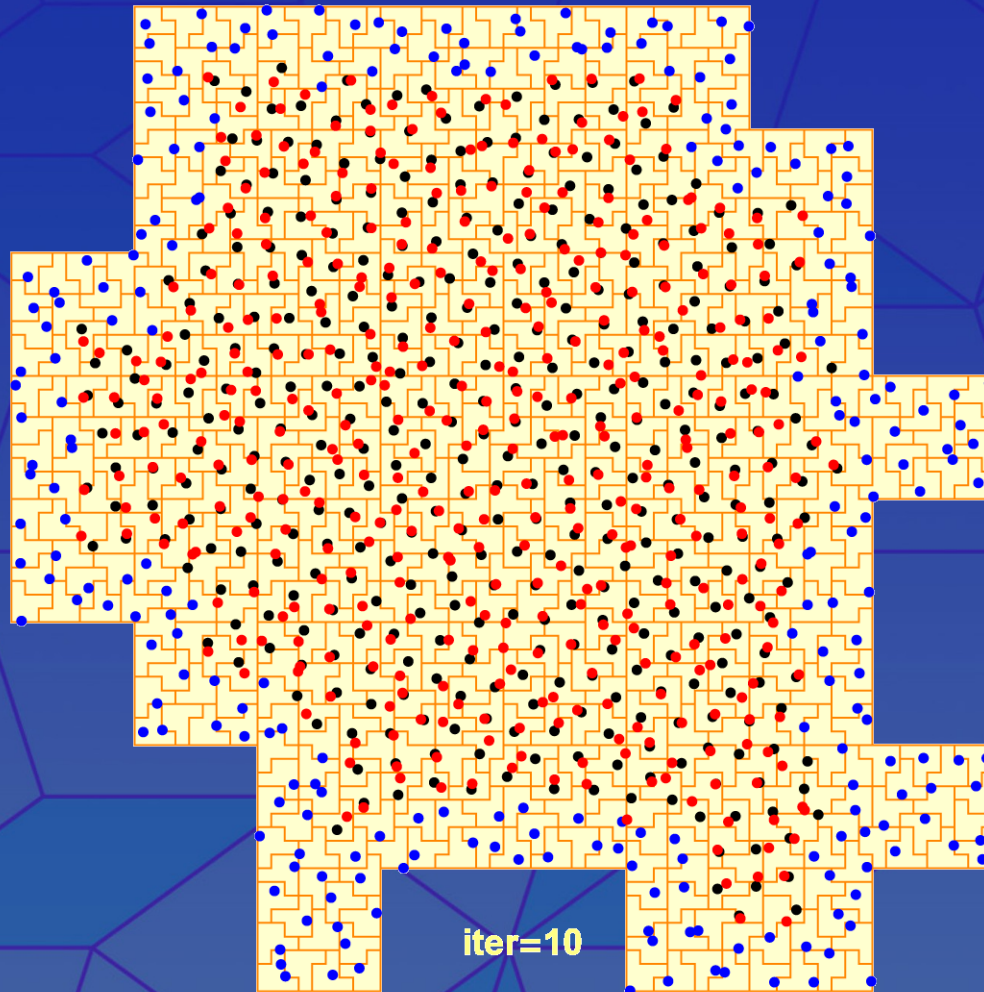
Optimization I

Relaxation



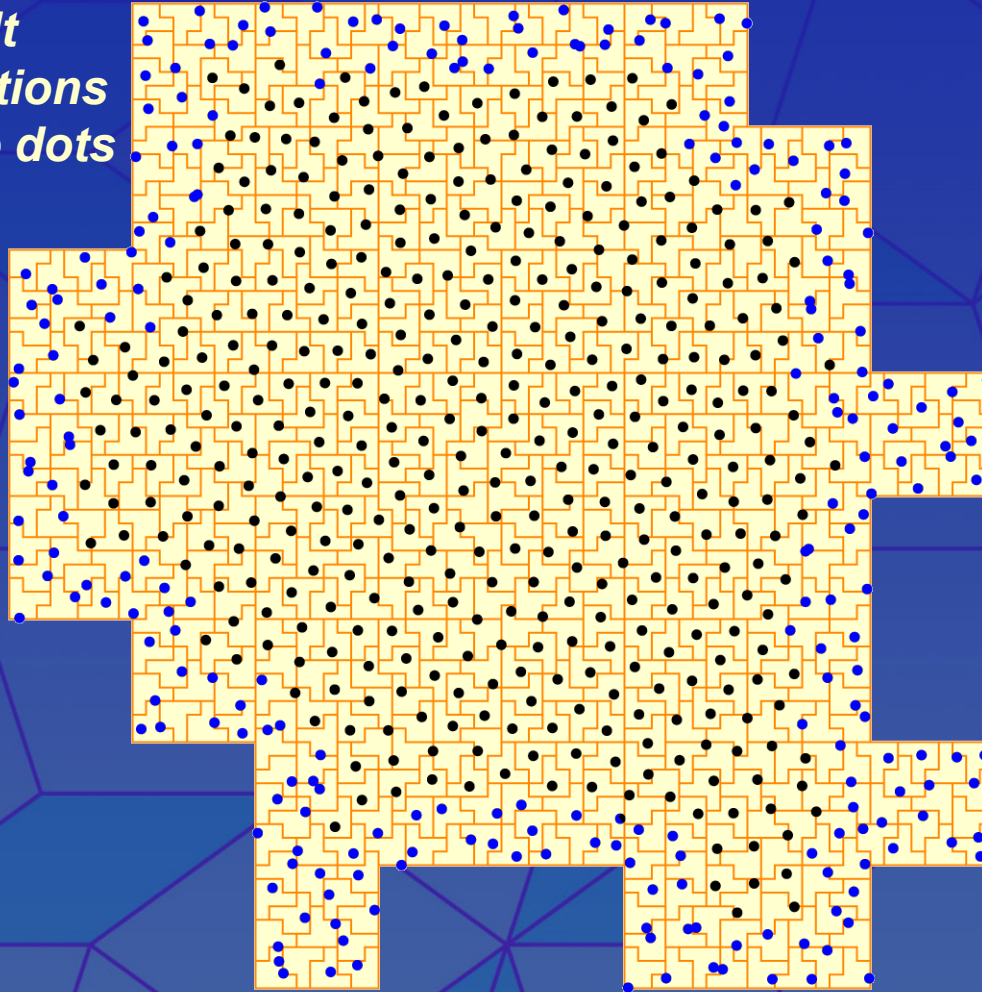
Optimization I

Relaxation



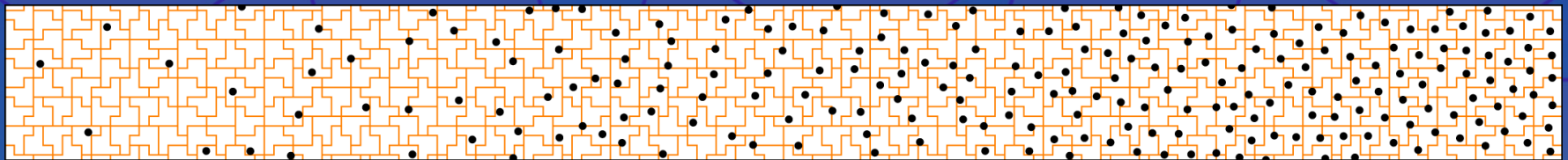
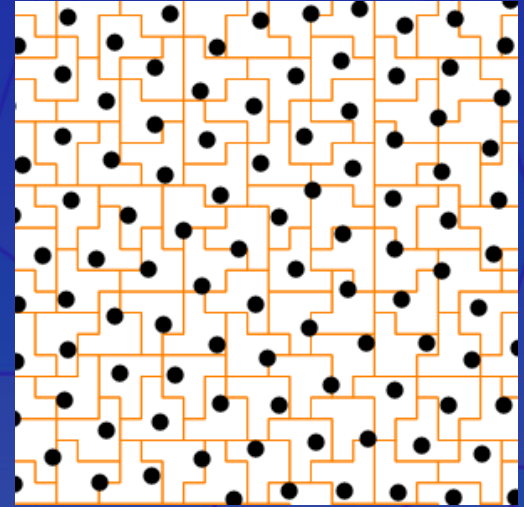
Optimization I

*Black: final result
after 10 relaxations
Blue: immovable dots*



Optimization in two phases

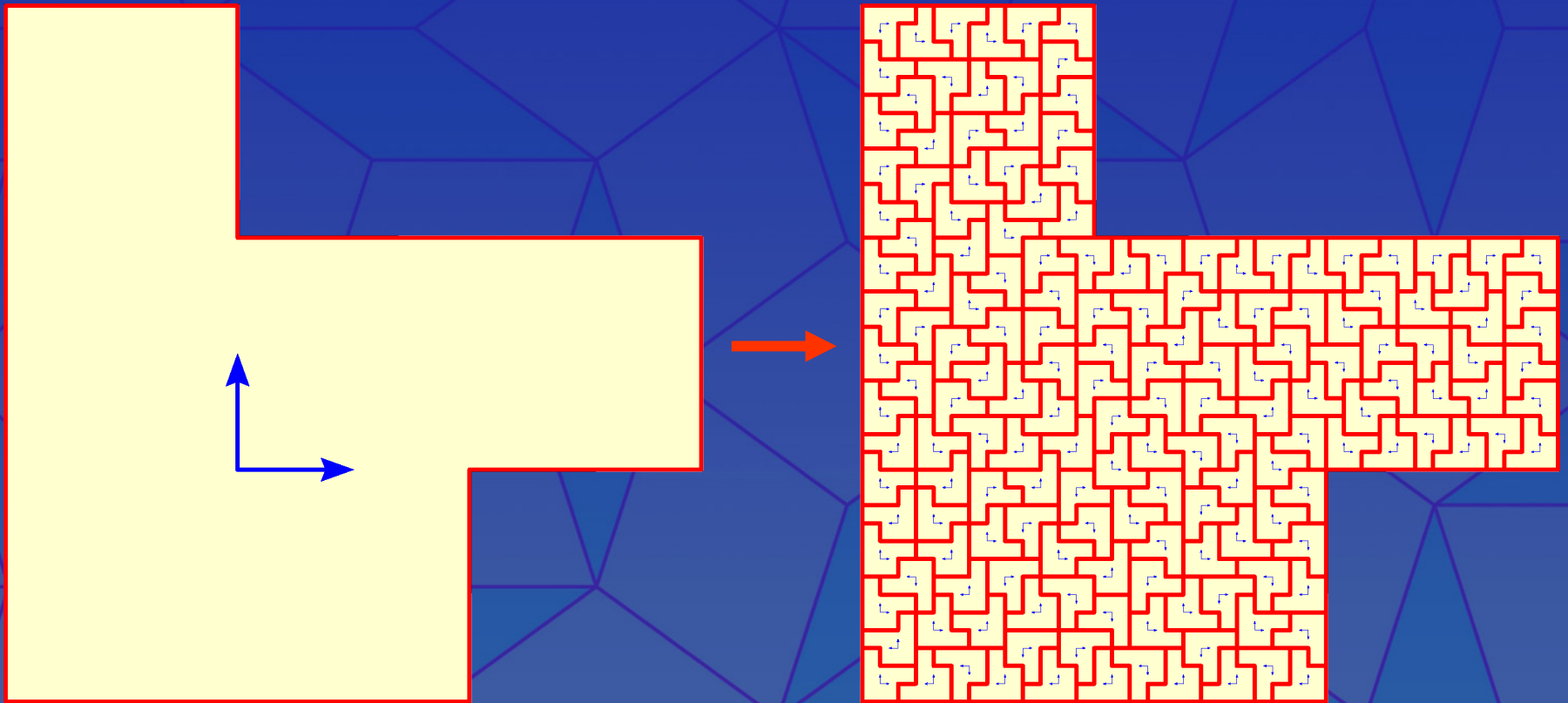
- *Optimization I: finding optimal position of the sampling point within each polyomino, for constant density*
- *Optimization II: ranking and relaxation*



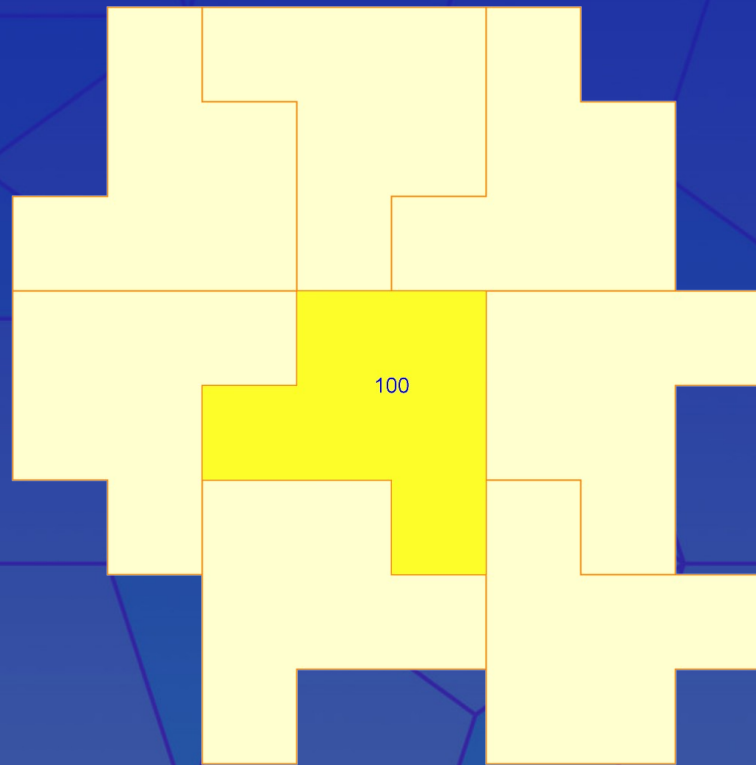
Ref: “The Void-and-cluster Method for Dither Array Generation” [Ulichney’93]

Optimization II-ranking

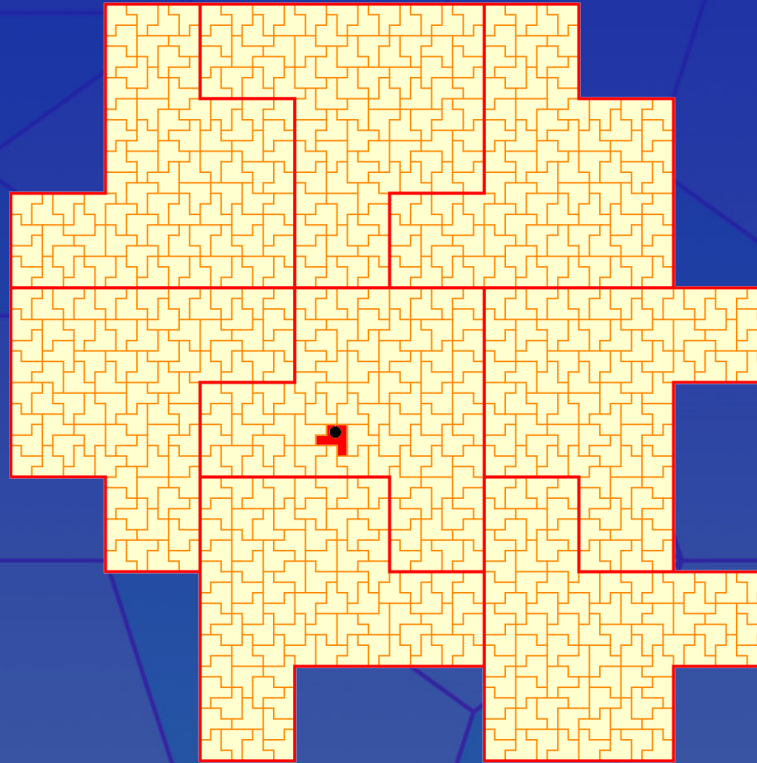
81-rep G-hexominoes



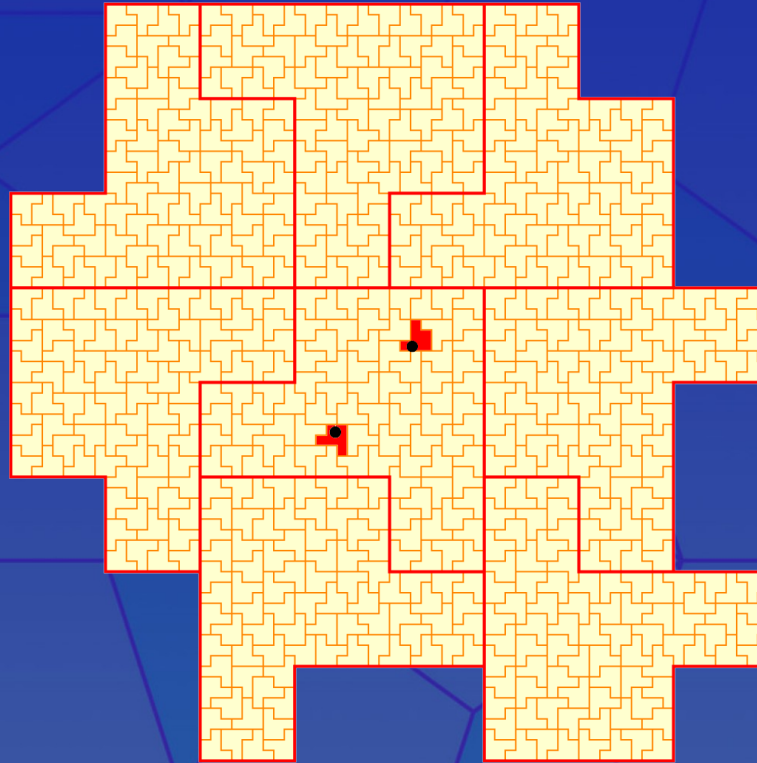
Optimization II-ranking



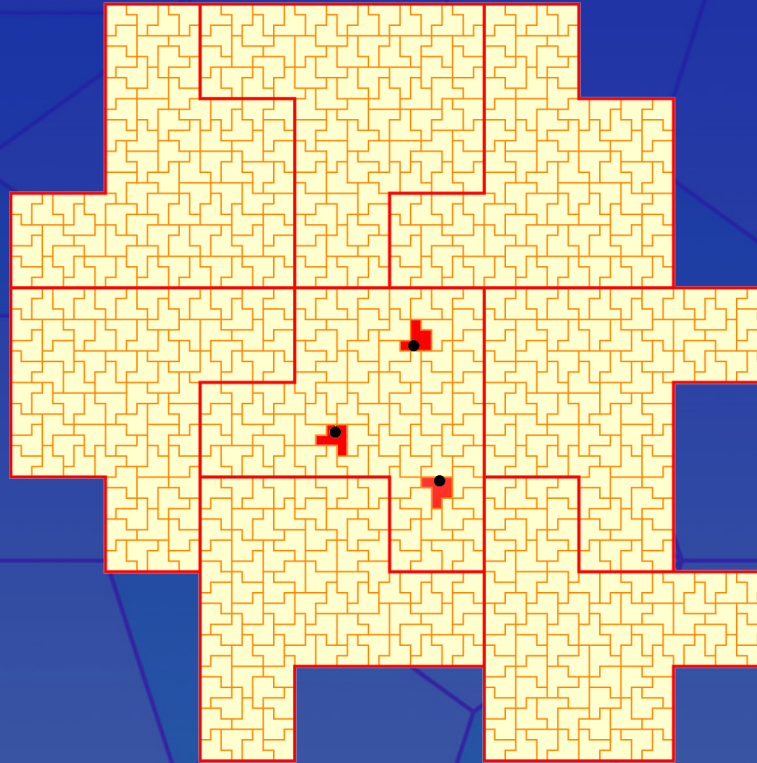
Optimization II-ranking



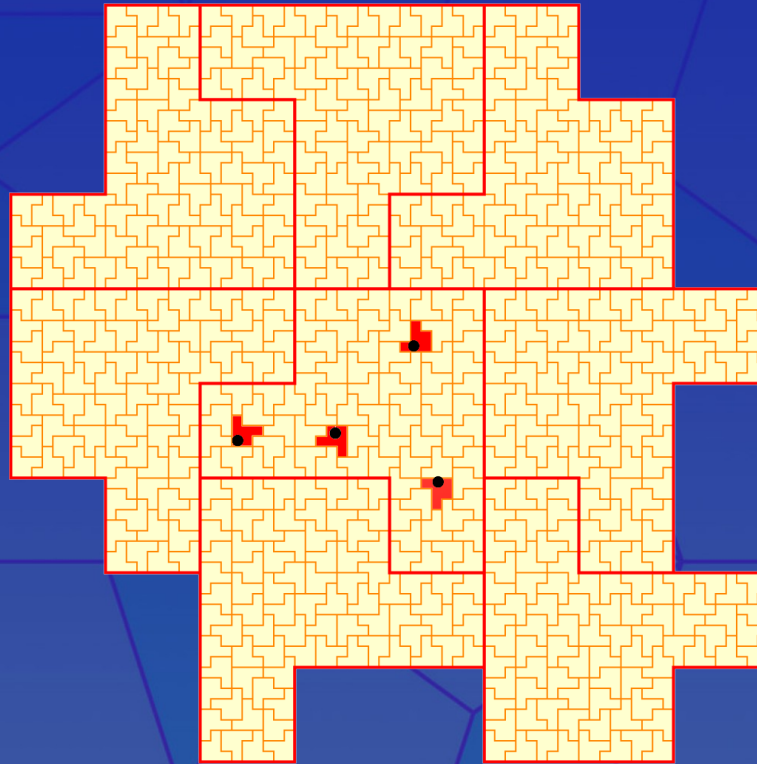
Optimization II-ranking



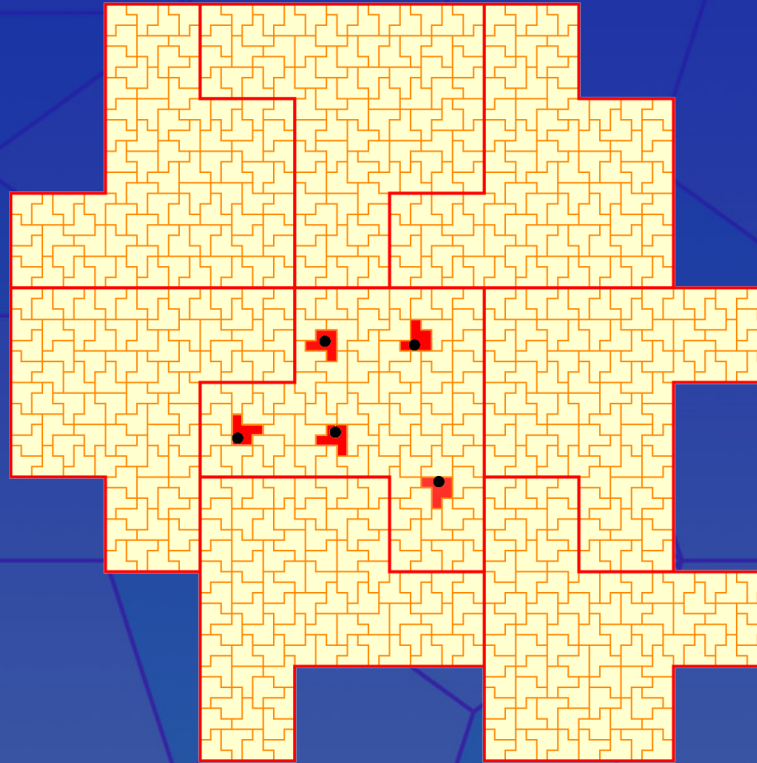
Optimization II-ranking



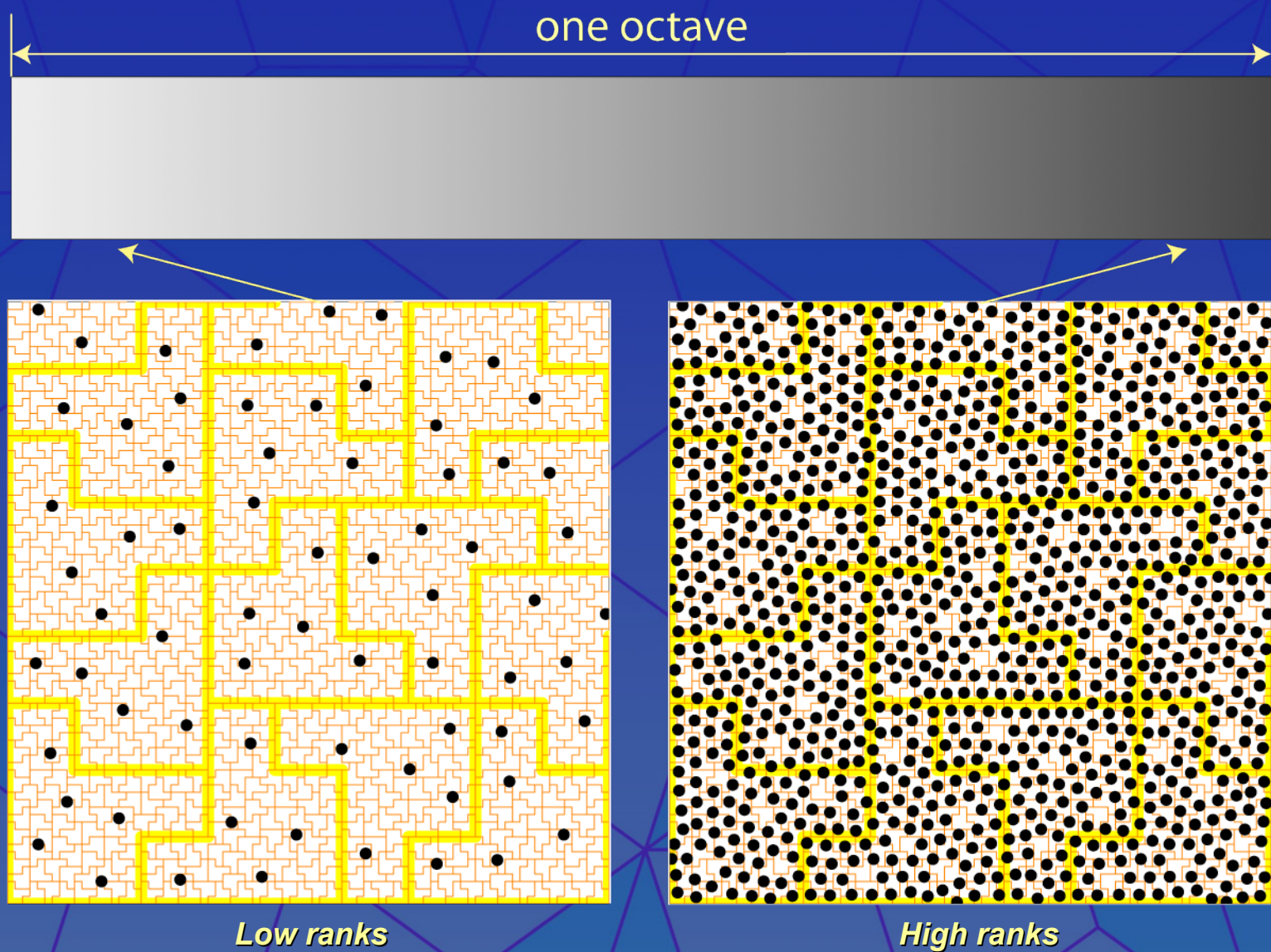
Optimization II-ranking



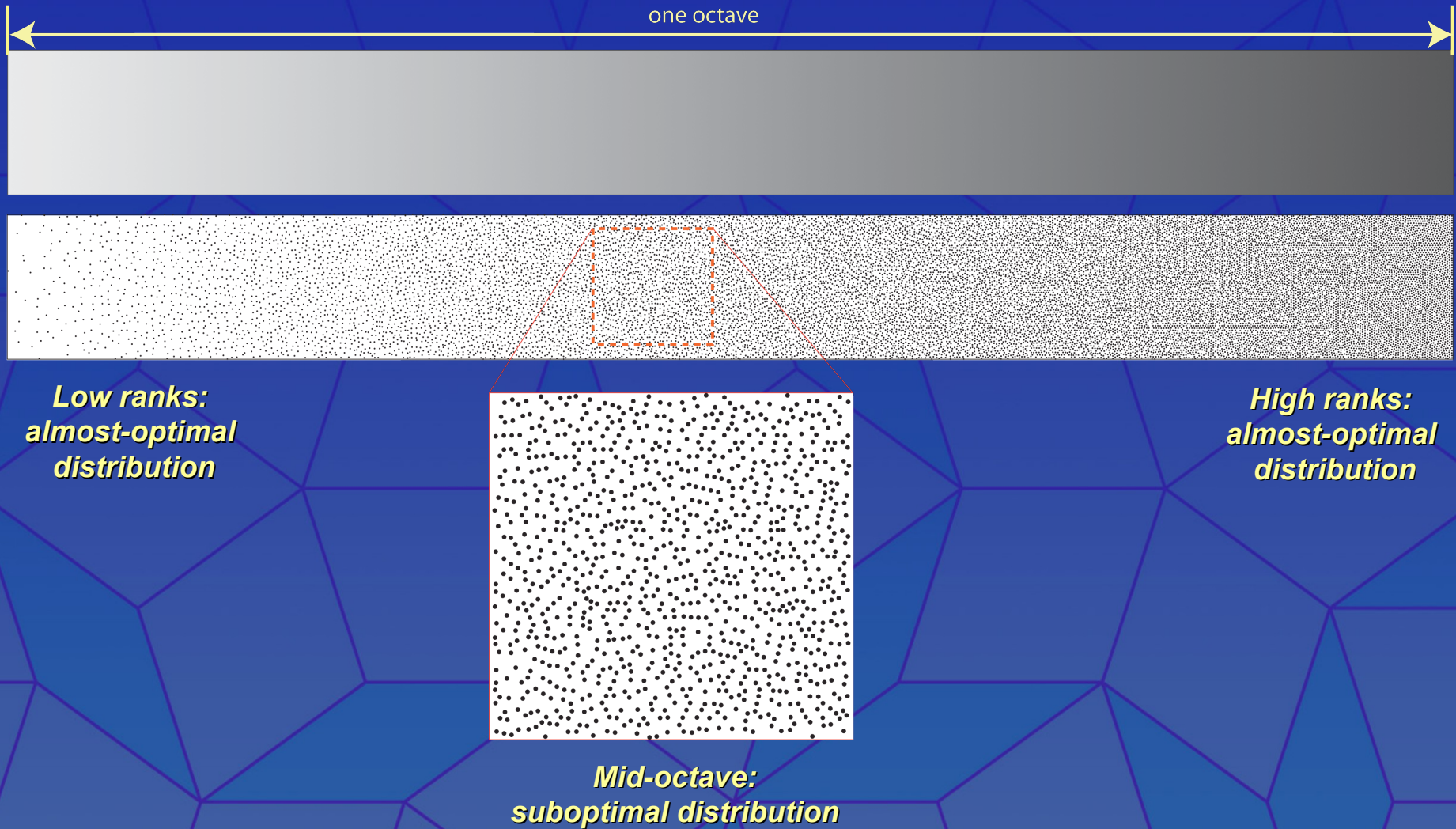
Optimization II-ranking



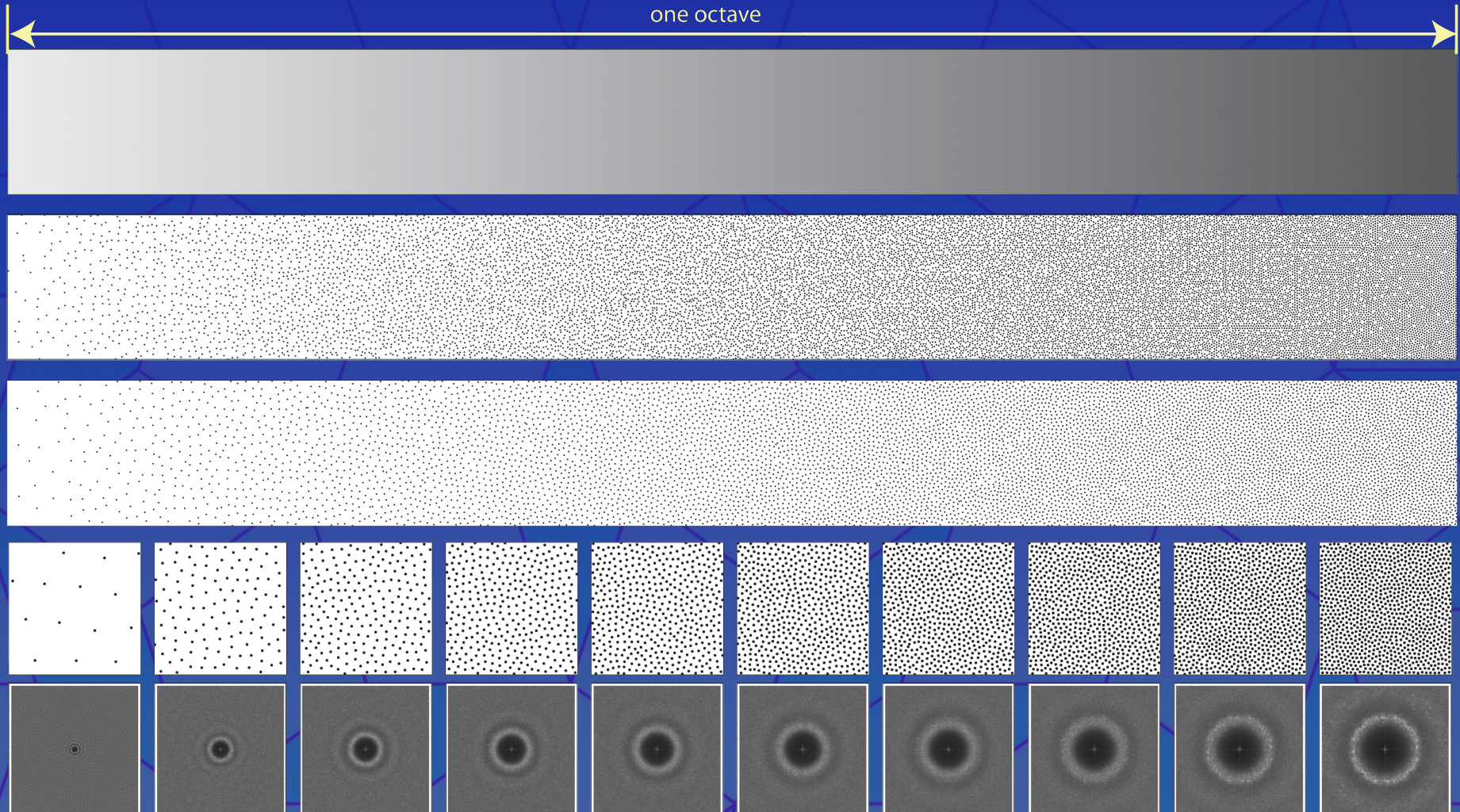
Optimization II-ranking



Optimization II-ranking



Optimization II-relaxation



Run-time pseudo-code

ADAPTIVESAMP(*p* of type *polyomino*)

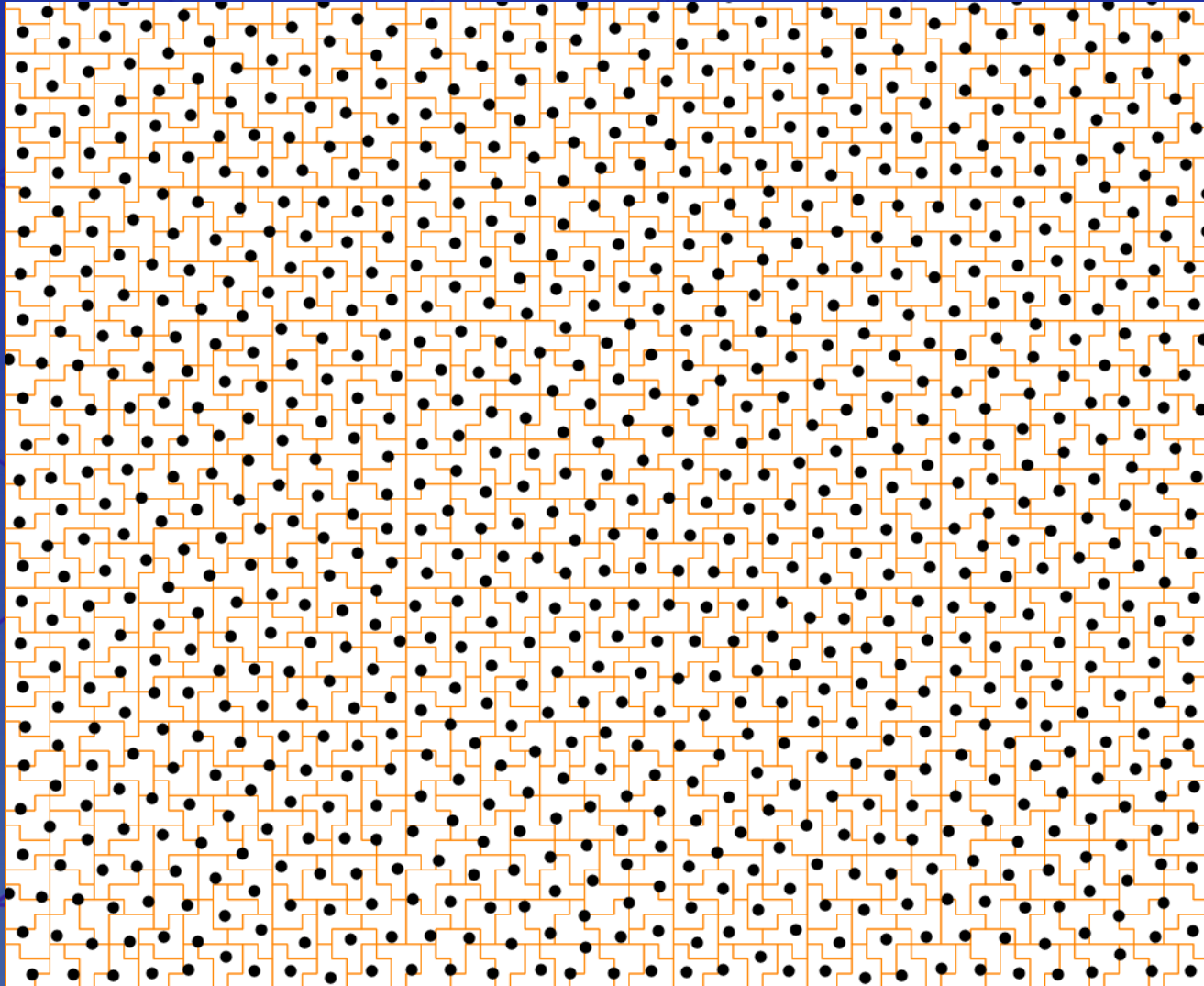
```
1      ▷ Structure polyomino contains the fields:
2      ▷   structuralIndex
3      ▷   LOS: level of subdivision
4      ▷   refPoint
5      ▷    $v_1, v_2$ : polyomino's basis vectors
6      ▷   code: used for computing threshold
7      local_LOS  $\leftarrow$  GETMAXLOSWITHINPOLYOMINO(p)
8      if p.LOS  $\geq$  local_LOS
9          then      ▷ Terminal: no need for more subdivisions
10         local_importance  $\leftarrow$  GETLOCALIMPORTANCE(p)
11         threshold  $\leftarrow$  INTEGERINBASE $\mathcal{A}$ (p.code)
12         if local_importance  $\geq$  threshold
13             then      ▷ Output selected sample
14                 iimp  $\leftarrow$  GETIMPORTANCEINDEX(local_importance)
15                  $\{k_1, k_2\} \leftarrow$  lut[p.structuralIndex, iimp]
16                 position  $\leftarrow$  p.refPoint +  $k_1 * p.v_1 + k_2 * p.v_2$ 
17                 OUTPUTSAMPLE(position)
18         return
19     else      ▷ Need more subdivisions
20          $\{p_1, \dots, p_{\mathcal{A}}\} \leftarrow$  SUBDIVIDEUSINGPRODUCTIONRULES(p)
21         return {ADAPTIVESAMP(p1), ..., ADAPTIVESAMP(p $\mathcal{A}$ )}
```


Run-time pseudo-code

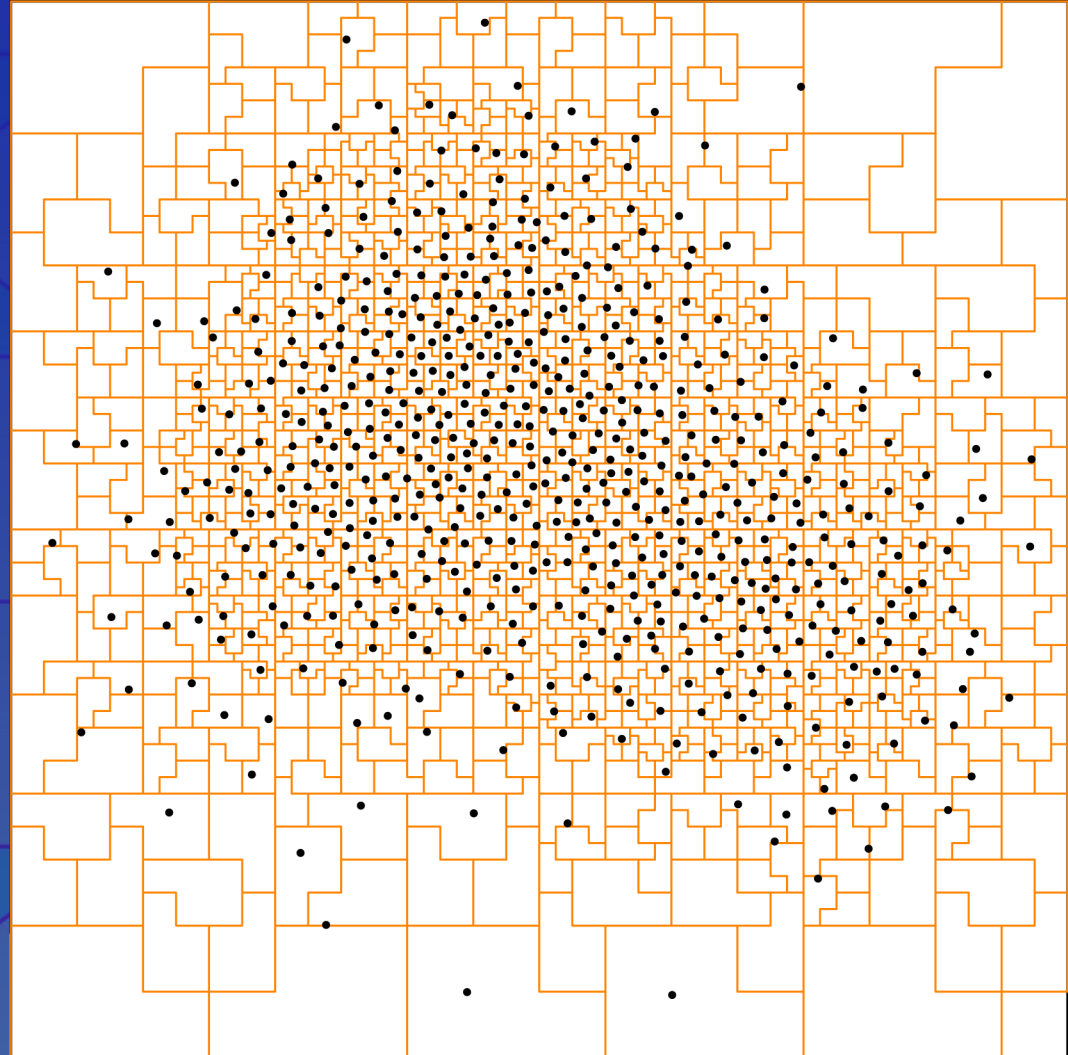
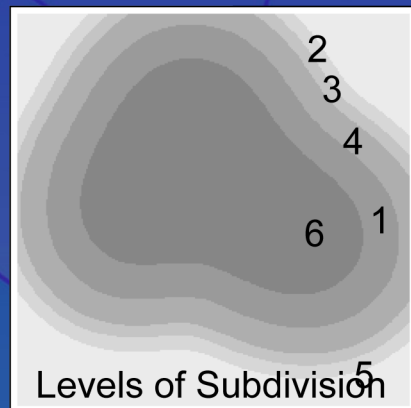
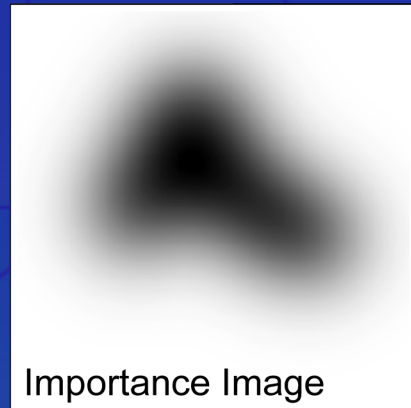
ADAPTIVESAMP(*p* of type *polyomino*)

```
1      ▷ Structure polyomino contains the fields:
2      ▷   structuralIndex
3      ▷   LOS: level of subdivision
4      ▷   refPoint
5      ▷    $v_1, v_2$ : polyomino's basis vectors
6      ▷   code: used for computing threshold
7      local_LOS  $\leftarrow$  GETMAXLOSWITHINPOLYOMINO(p)
8      if p.LOS  $\geq$  local_LOS
9          then      ▷ Terminal: no need for more subdivisions
10         local_importance  $\leftarrow$  GETLOCALIMPORTANCE(p)
11         threshold  $\leftarrow$  INTEGERINBASE $\mathcal{A}$ (p.code)
12         if local_importance  $\geq$  threshold
13             then      ▷ Output selected sample
14                 iimp  $\leftarrow$  GETIMPORTANCEINDEX(local_importance)
15                  $\{k_1, k_2\} \leftarrow \text{lut}[p.\text{structuralIndex}, i_{\text{imp}}]$ 
16                 position  $\leftarrow p.\text{refPoint} + k_1 * p.v_1 + k_2 * p.v_2$ 
17                 OUTPUTSAMPLE(position)
18         return
19     else      ▷ Need more subdivisions
20          $\{p_1, \dots, p_{\mathcal{A}}\} \leftarrow$  SUBDIVIDEUSINGPRODUCTIONRULES(p)
21         return {ADAPTIVESAMP(p1), ..., ADAPTIVESAMP(p $\mathcal{A}$ )}
```

Results – constant density

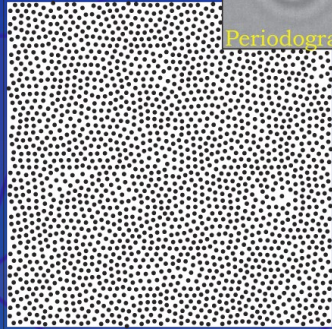
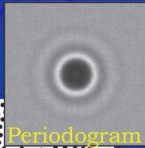


Results – variable density

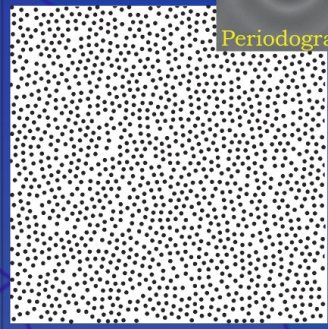
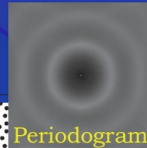


Results – comparison with other methods

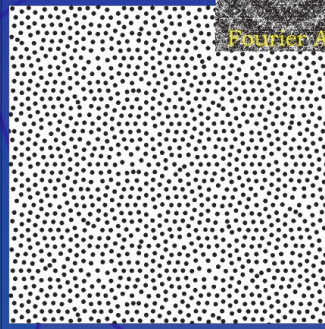
Corner
Wang Tiling



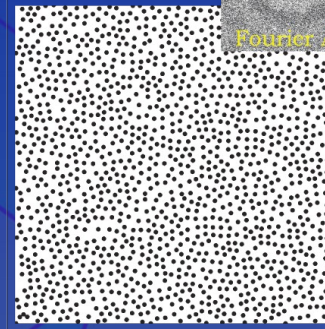
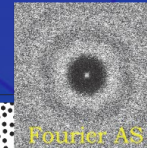
Boundary
Sampling



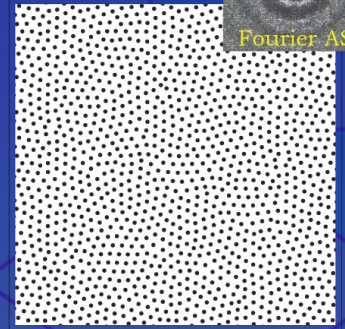
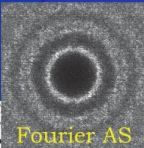
Penrose
Tiling



Recursive
Wang Tiling



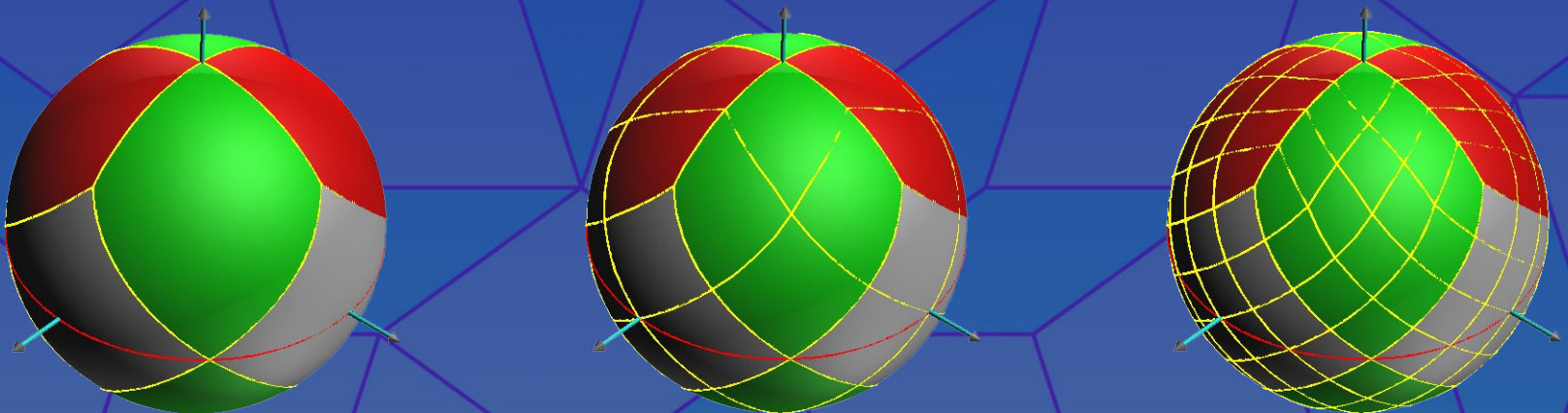
Polyominoes
(G-hexominoes)



Sampling on a sphere

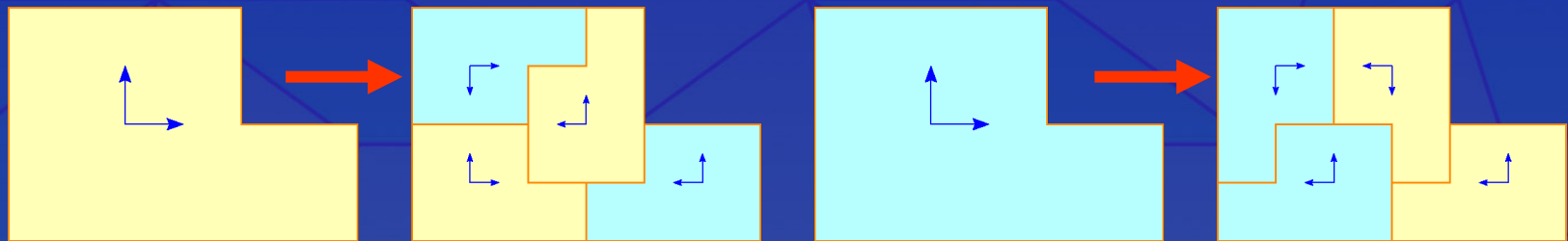
HEALPix spherical mapping [Gorski et al. 2005]

- 12 Equiareal high-level quads
- Jacobian-preserving mapping
 - Subdivision into equiareal quads
- Low distortion

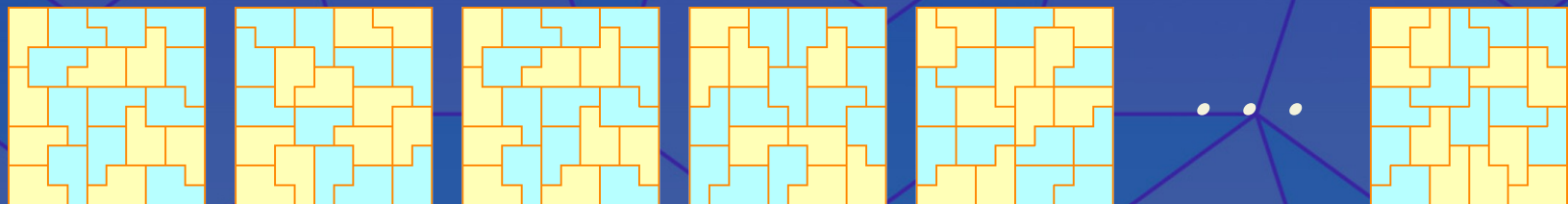


Sampling on a sphere

Pentominoes with two subdivision rules

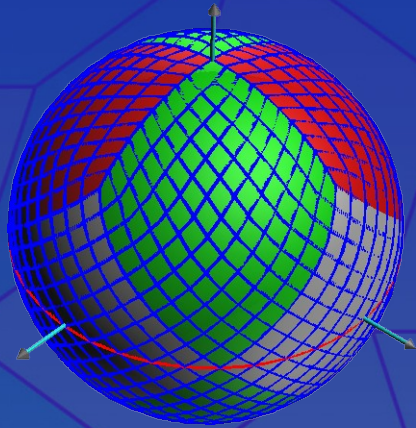


Square patches of 20 pentominoes on 10x10 squares

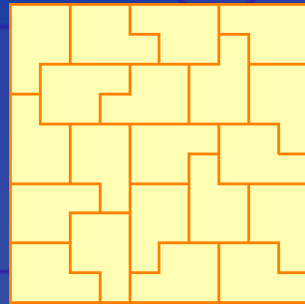


Sampling on a sphere

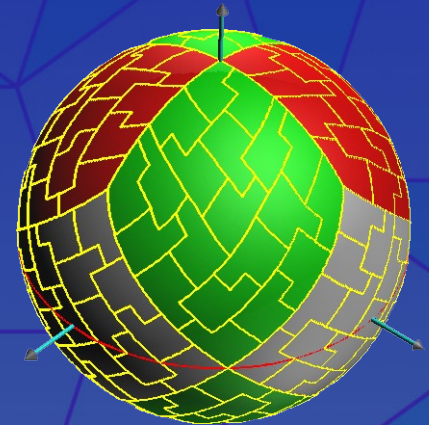
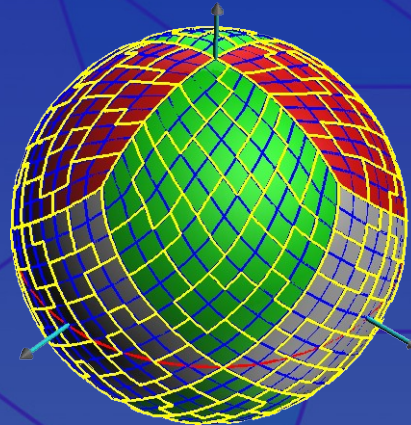
Pentominoes on 12 HEALPix quads



*10x10 grid
on each quad*



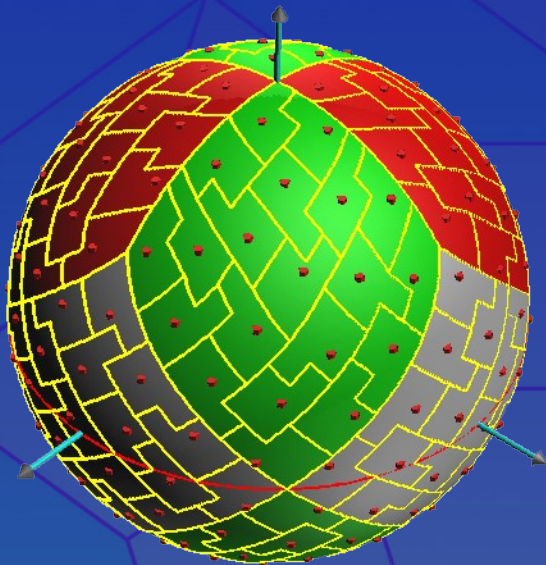
*20 pentominoes
on a 10x10 square*



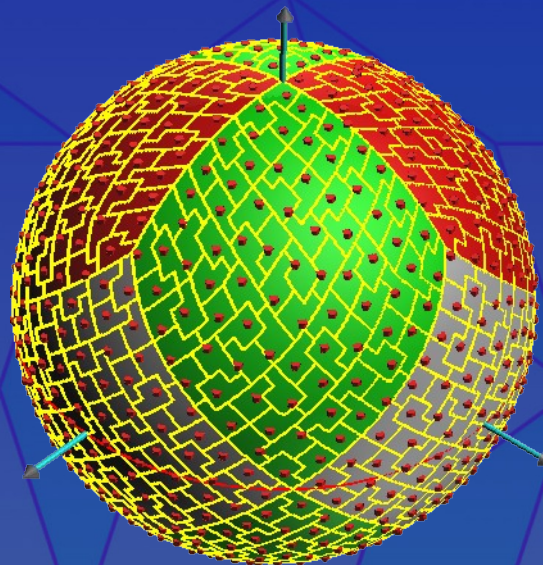
*20 pentominoes
on each quad*

Sampling on a sphere

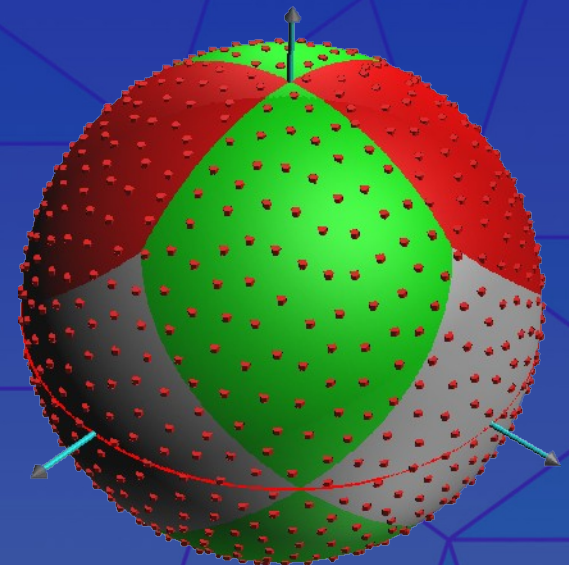
Subdivision and optimization



Level 1



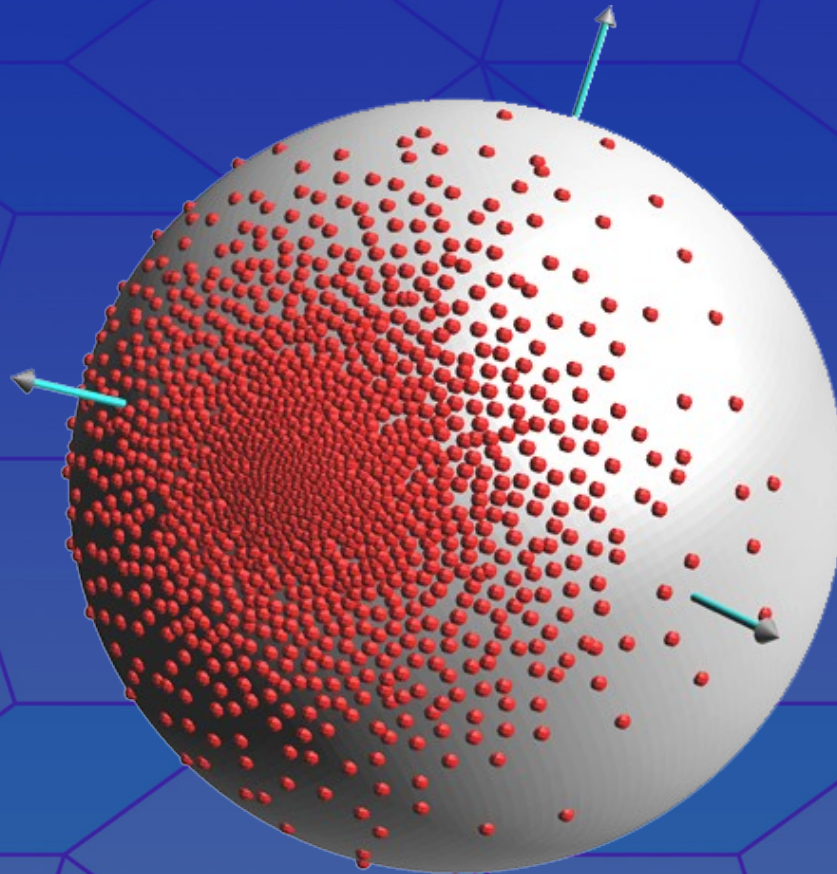
*Level 2
(after one subdivision)*



*Level 2
(dots only)*

Sampling on a sphere

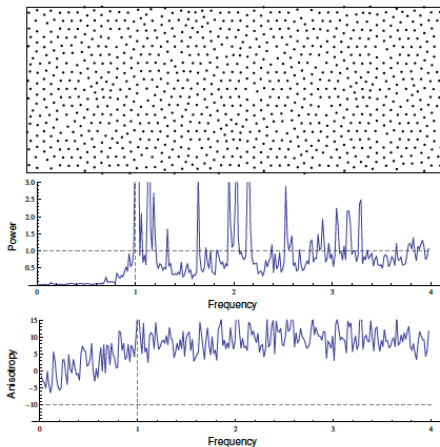
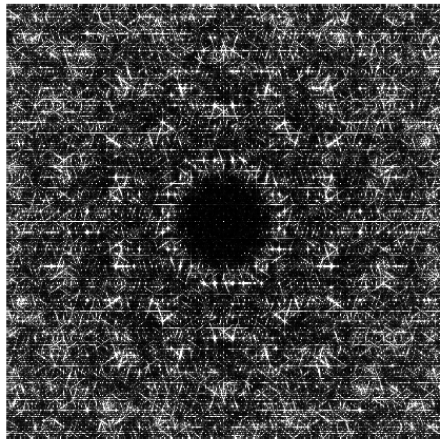
An arbitrary importance (spot) function



Further Improvement of the Spectrum (WIP)

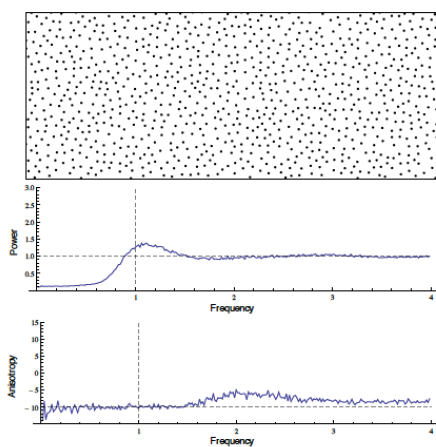
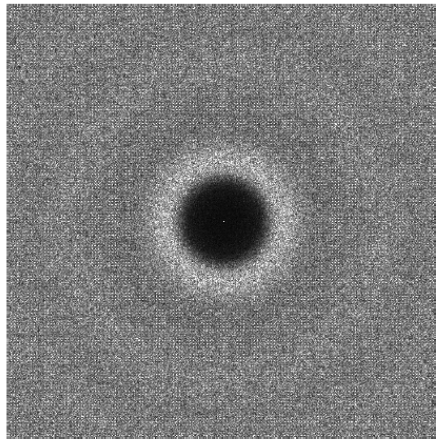
[Ostromoukhov et al. 2004]

1 sample per tile



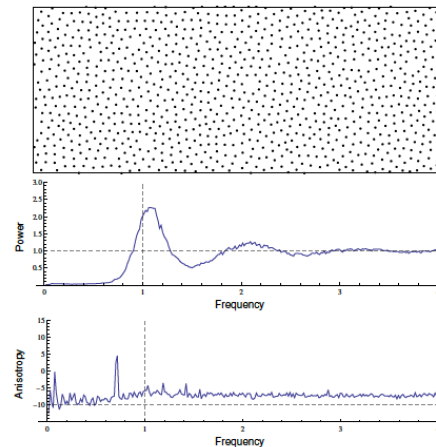
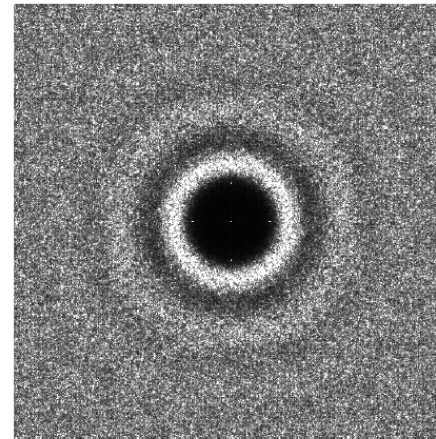
[Kopf et al. 2006]

1024 samples per tile



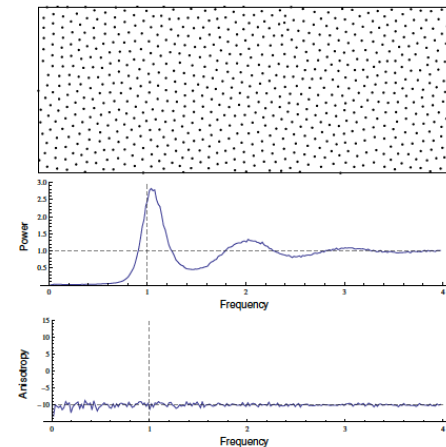
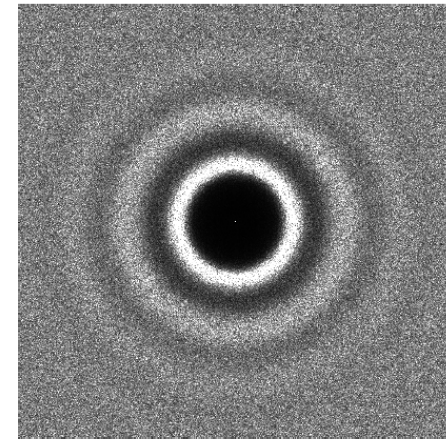
[Ostromoukhov 2007]

1 sample per tile

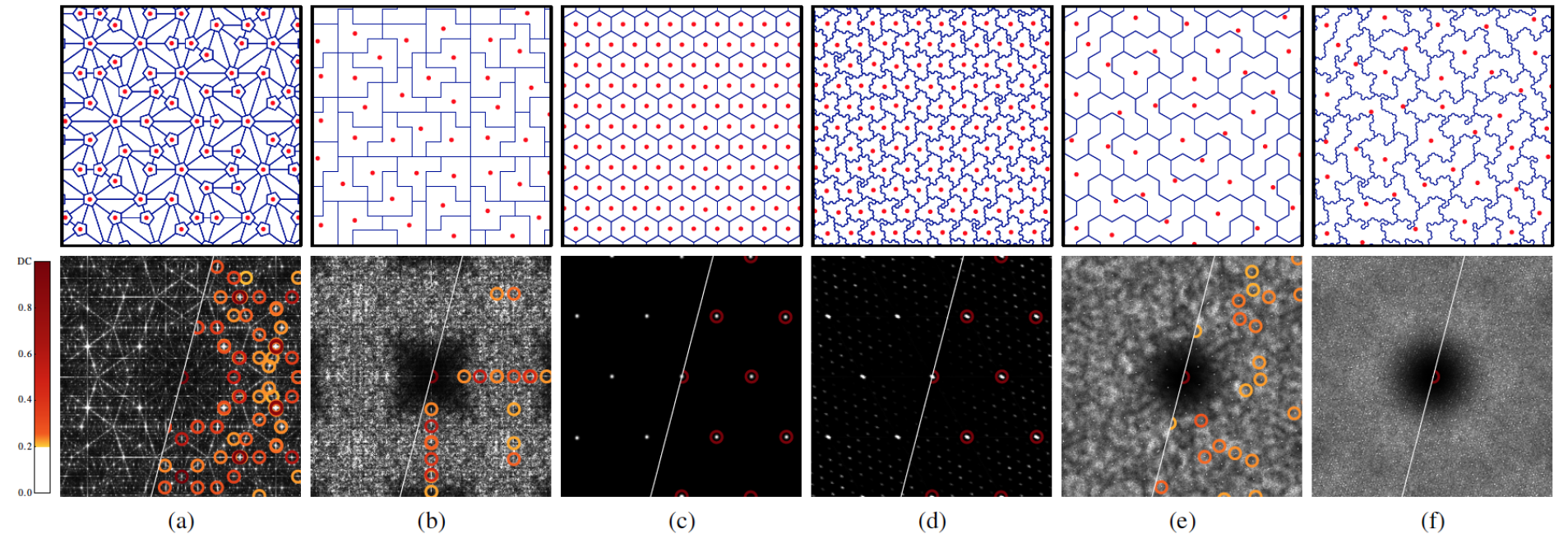


Ours

1 sample per tile



Further Improvement of the Spectrum



Building Blocks

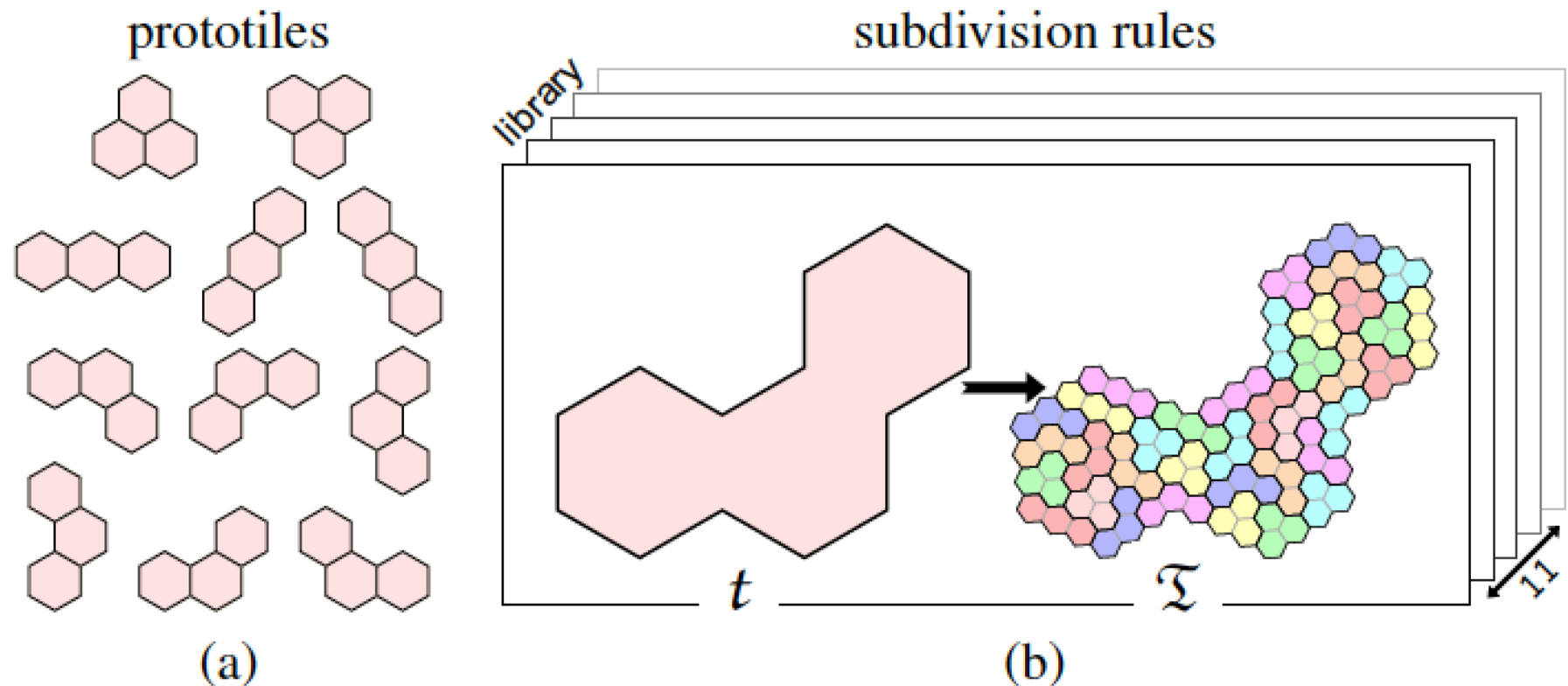
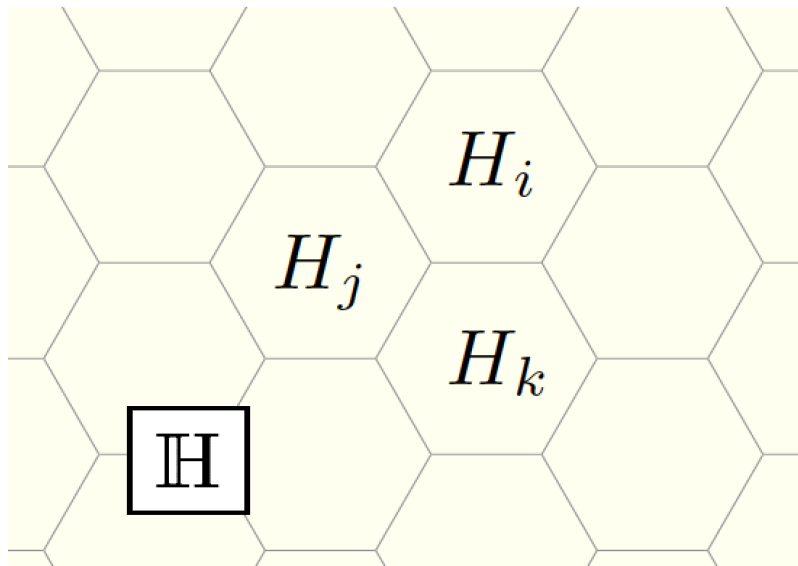
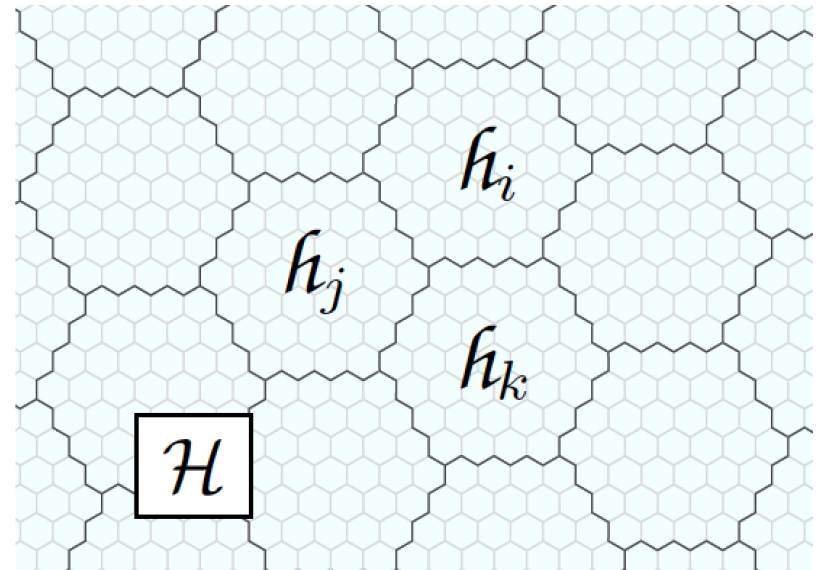


Figure 4: (a) 11 configurations of trihexes on $\mathbb{H}$. (b) The library of subdivision rules $\{t \rightarrow \mathfrak{T}\}$, where the quasi-trihex associated with t after subdivision is partitioned into a set of λ trihexes.

Building Blocks



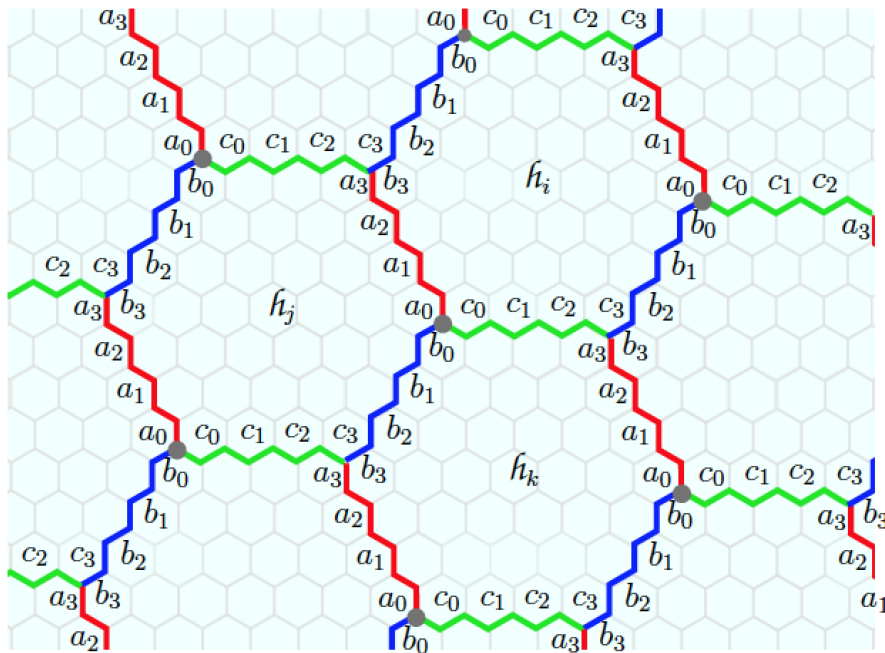
(a)



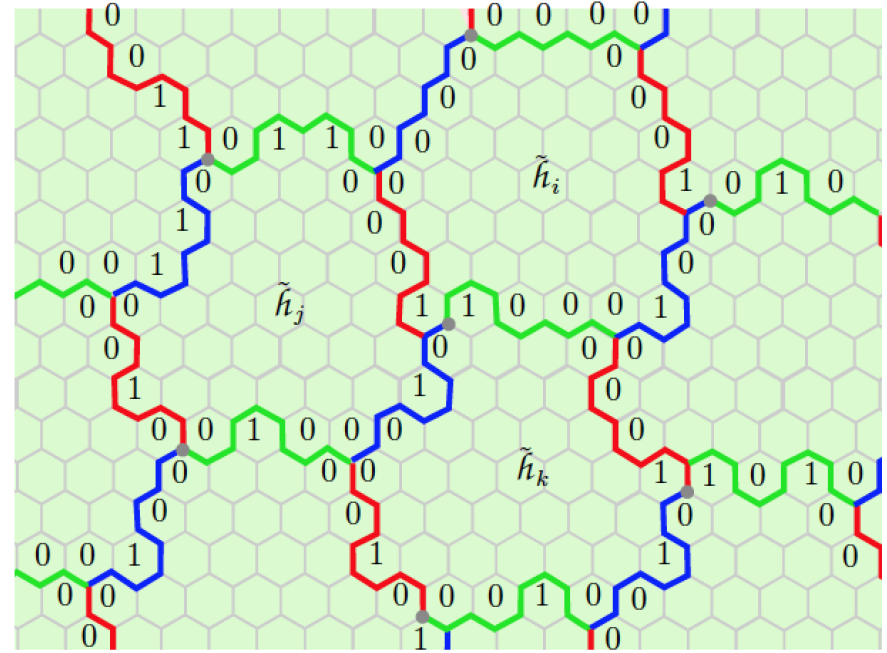
(b)

Figure 2: *Quasi-hexagonal tiles. (a) The unbounded, regular hexagonal lattice $\mathbb{H}$. (b) The unbounded, quasi-hexagonal lattice $\mathcal{H}$. Mention λ and $\mathfrak{h}$.*

Building Blocks



(a)



(b)

Figure 3: *Effect of edge modifiers. From a set of hextiles h_i and random valid edge modifiers (a), a set of irregular tiles $\tilde{h}_i$ in $\tilde{\mathcal{H}}$ is obtained (b) which maintains the topology of the initial tiling as well as each tile's area.*

Building Blocks

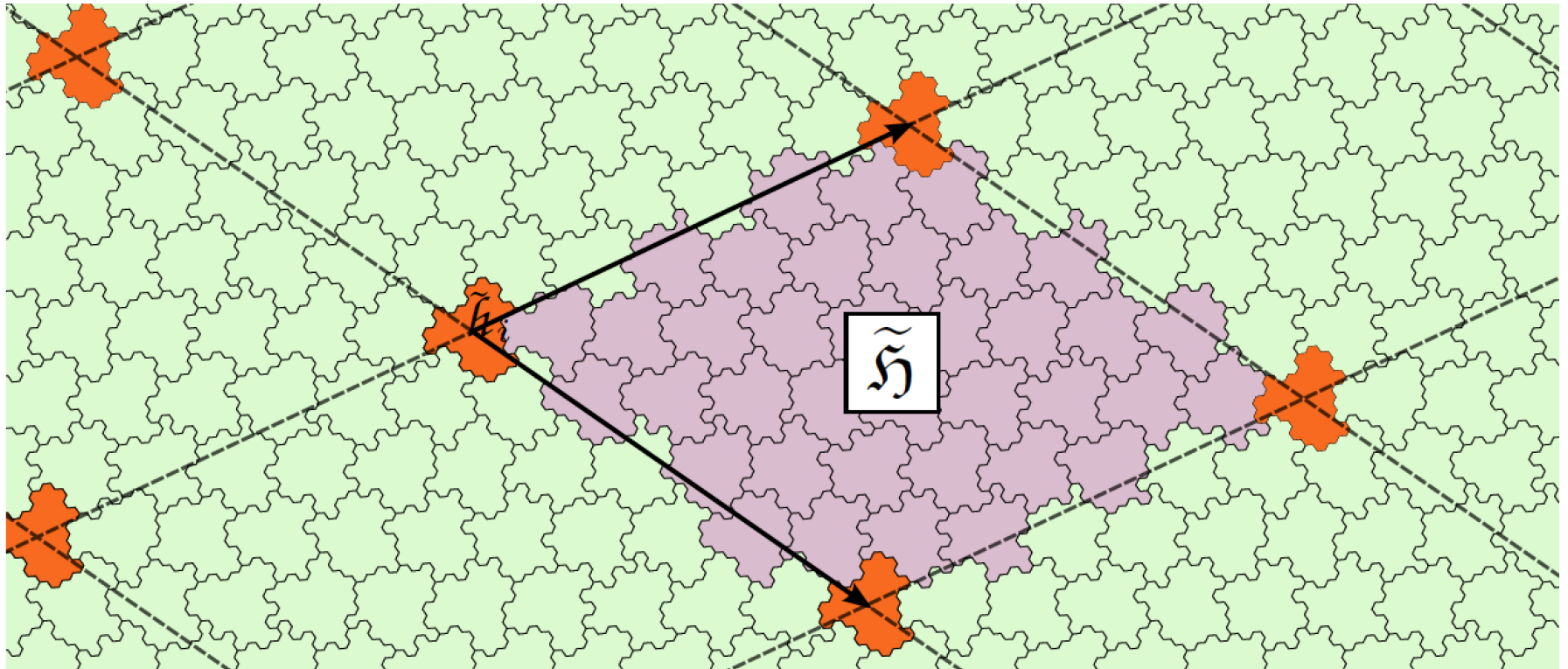
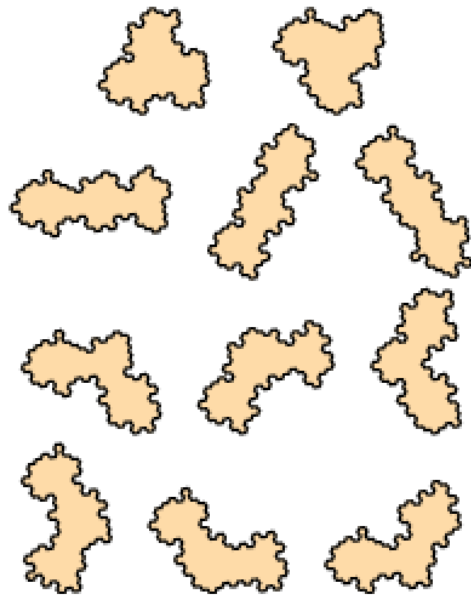


Figure 5: *Periodic tiling construction based on two integer basis vectors, creating a small set $\tilde{\mathcal{H}}$ of λ irregular hextiles $\{\tilde{h}_i\}$ (highlighted).*

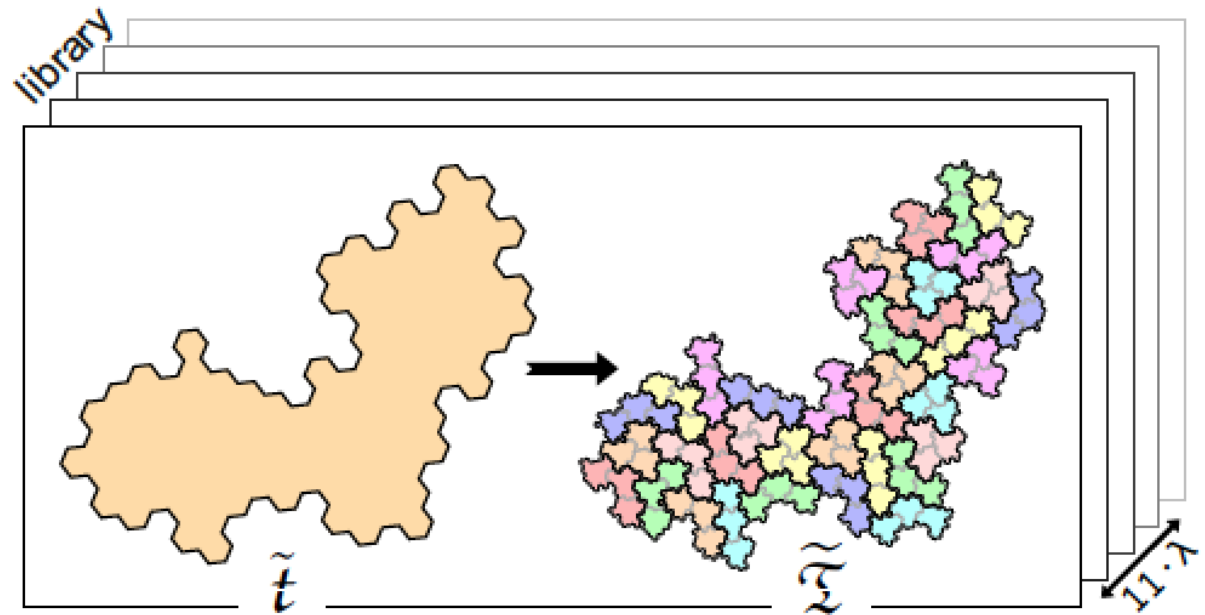
Building Blocks

irregular trihexes (partial)



(a)

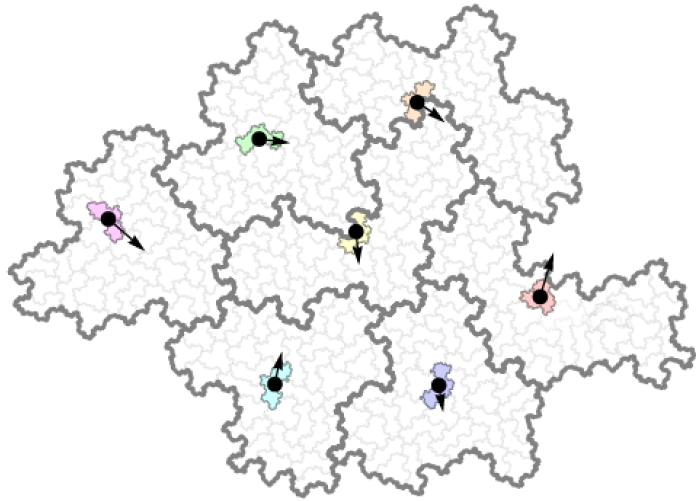
subdivision rules



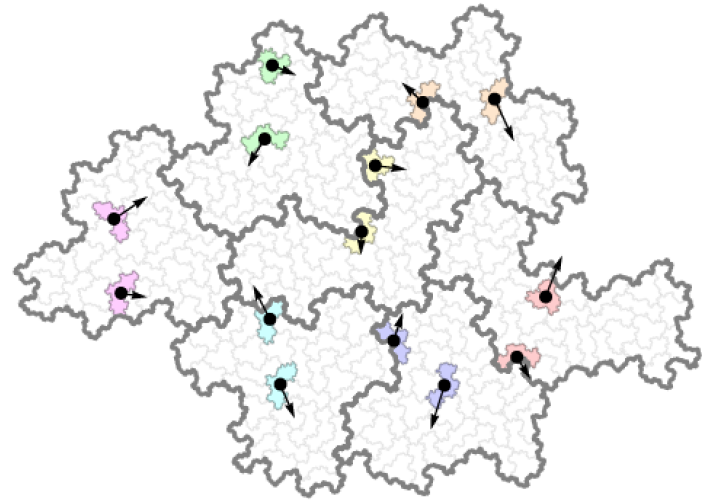
(b)

Figure 7: (a) Examples of irregular trihexes in $\mathcal{T}$ (11 out of $11.\lambda$ are shown).
 (b) The library of subdivision rules $\{\tilde{t} \rightarrow \tilde{\mathcal{T}}\}$ for all irregular trihexes.

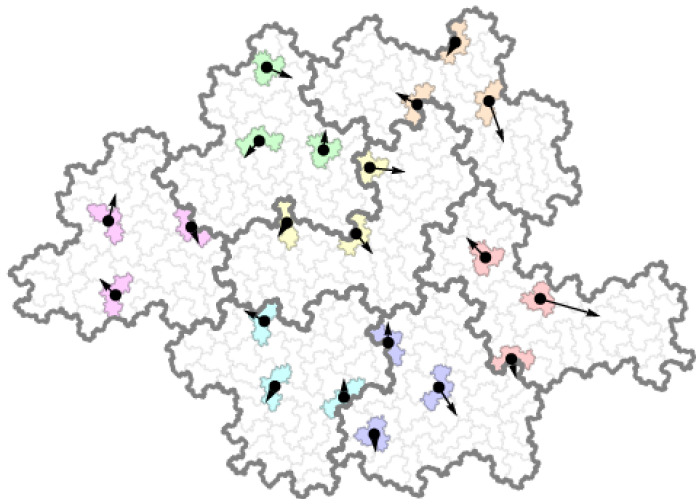
Ranking



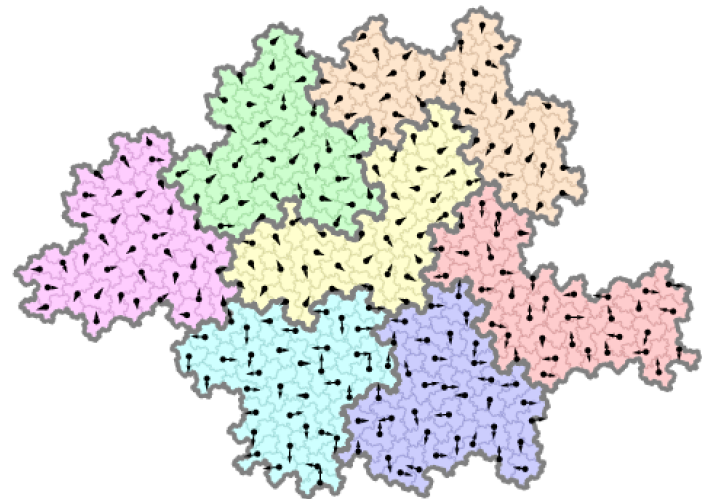
(a) R_1



(b) R_2



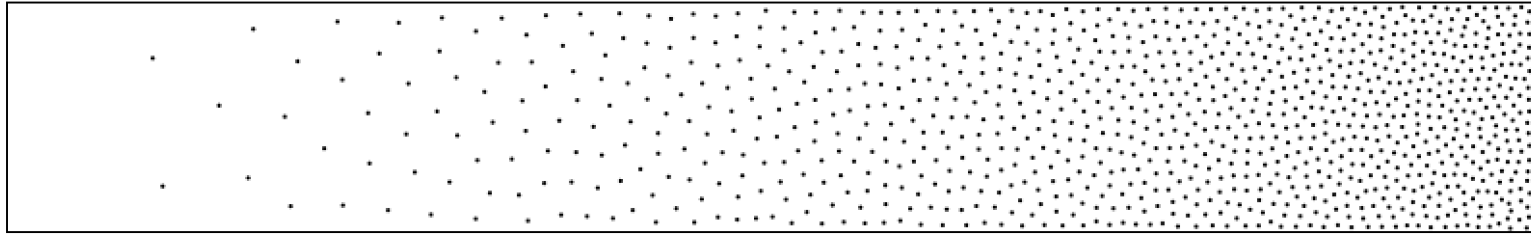
(c) R_3



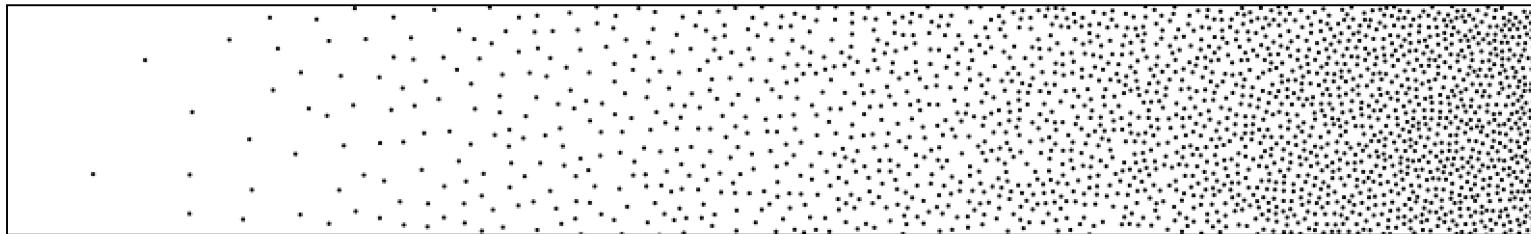
(d) R_{37}

Results

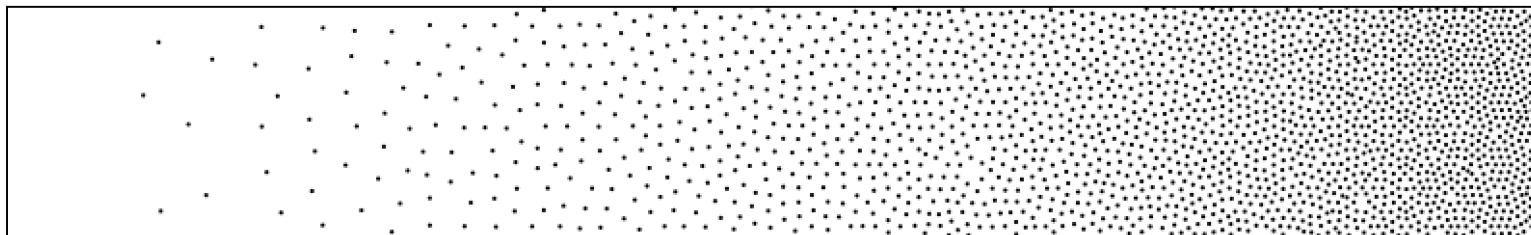
[de Goes et al.
2012]



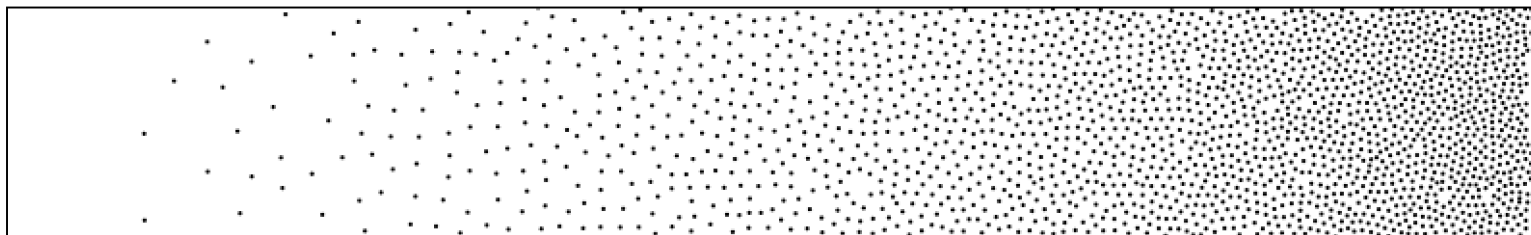
[Kopf et al.
2006]



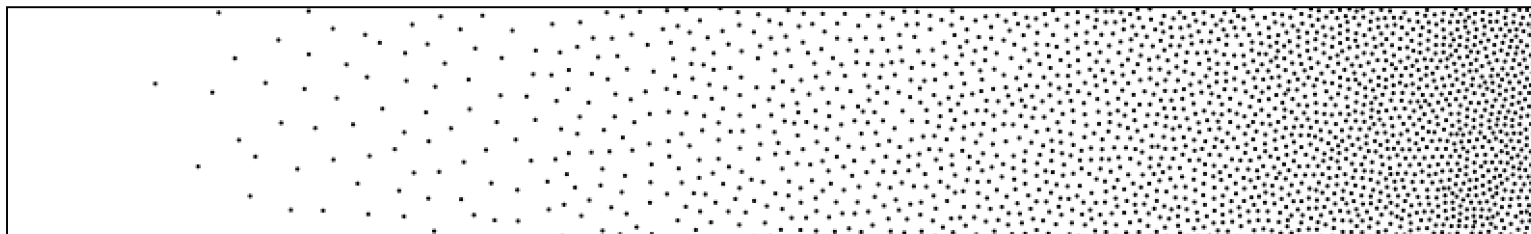
[Ostromoukhov
et al. 2004]



[Ostromoukhov
2007]

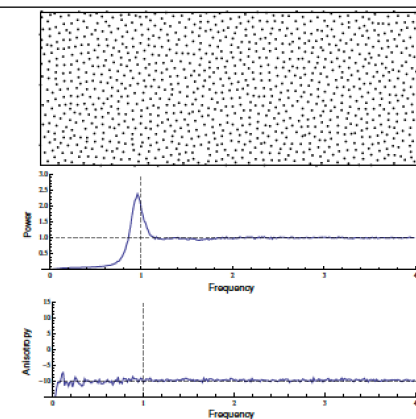
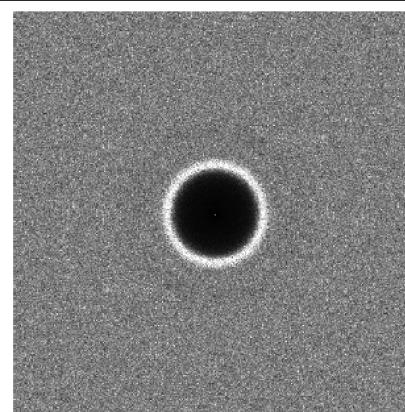
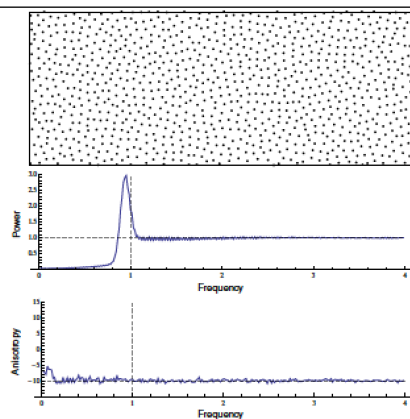
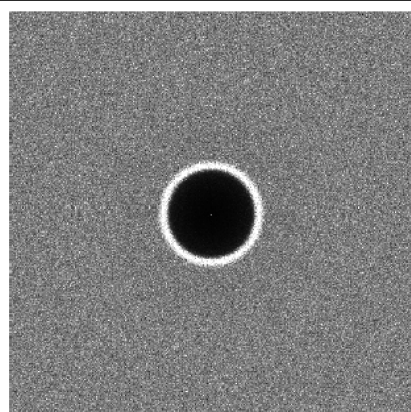
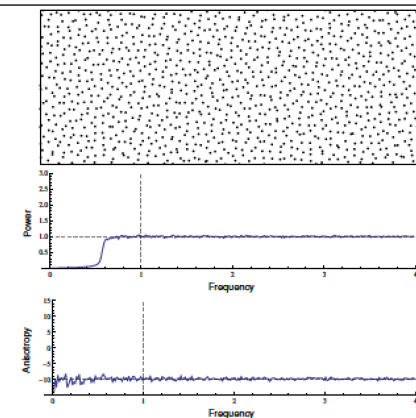
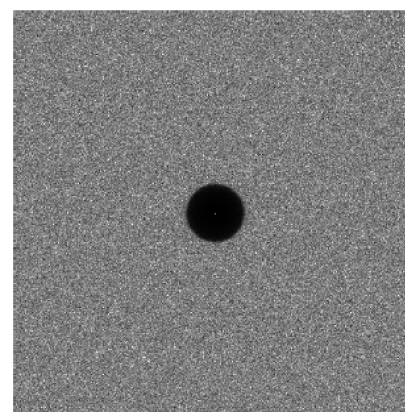
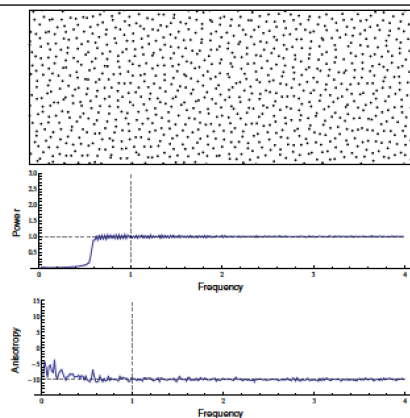
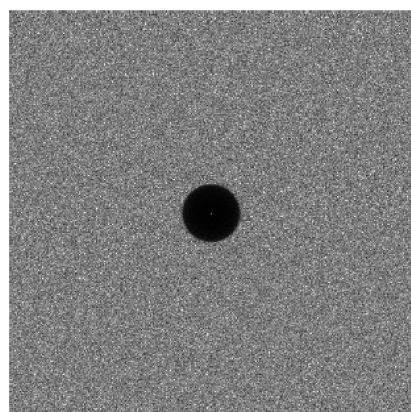
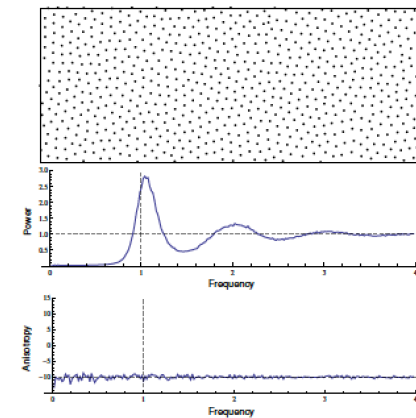
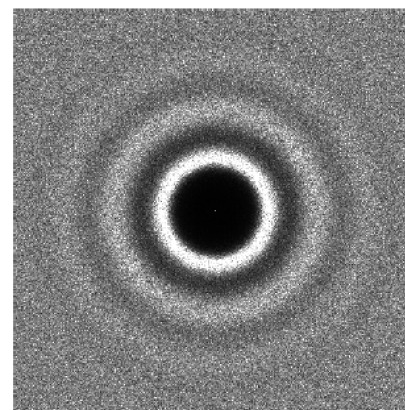
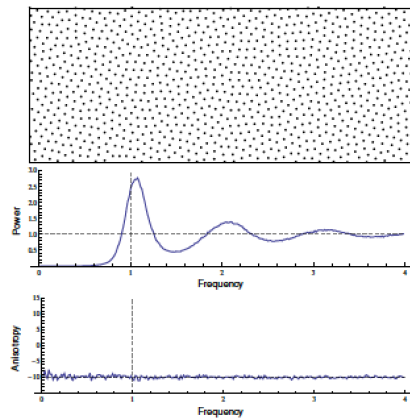
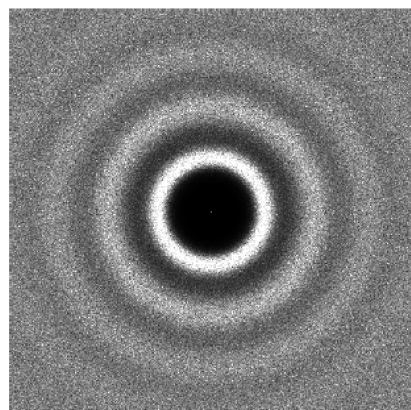


Ours



Original method

Results through our method



Results: Qualitative Analysis

	Qualitative Analysis			
Methods	Speed	Memory footprint	Distribution Quality	Spectral Control
[Schlömer et al. 2011]	poor	good	good	✗
[de Goes et al. 2012]	poor	good	excellent	✗
General Noise (blue noise target)	very poor	good	good	✓
[Ostromoukhov et al. 2004]	excellent	poor	satisfactory	✗
[Kopf et al. 2006]	excellent	poor	good	✗
[Ostromoukhov 2007]	excellent	poor	good	✗
Ours (blue noise target)	excellent	poor	excellent	✓
Ours (general noise target)	excellent	poor	good	✓

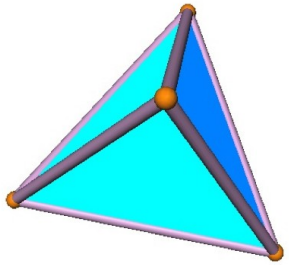
Future Challenges

- Reduce Memory Footprint
- Extend to Higher Dimensions
- Be Able to Build an Irregular Tiling System Approaching Required Spectral Properties (Penrose-like Tiling)
 - Compact
 - High (Or Without) Rotational Symmetry
 - With Appropriate Number System

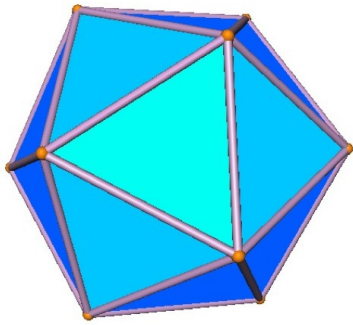
Thank

You!

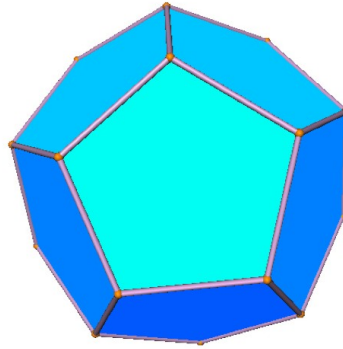
Space-Filling Polyhedra (Plesiohedra)



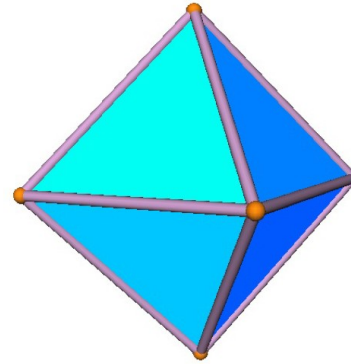
Tetrahedron



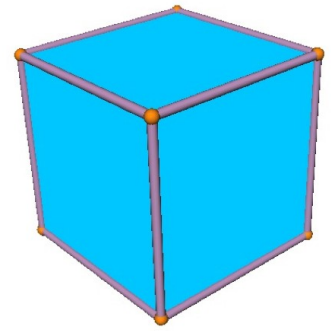
Icosahedron



Dodecahedron



Octahedron



Cube