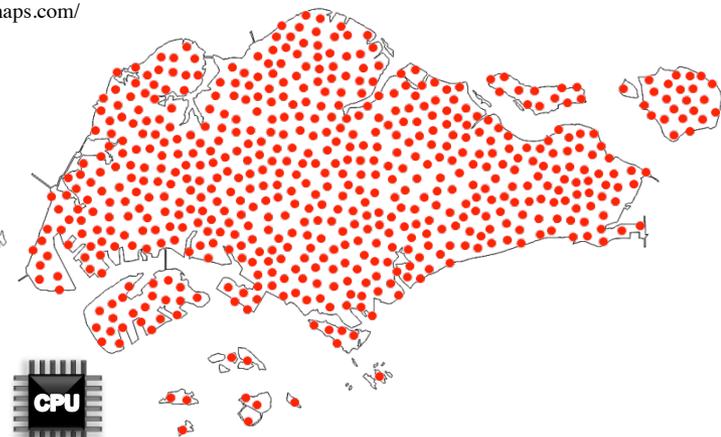
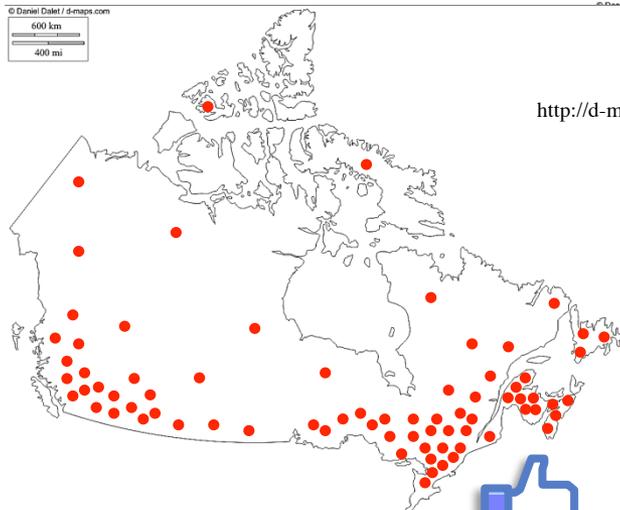


© Daniel Dalel / d-maps.com  
600 km  
400 mi

<http://d-maps.com/>



# Efficient Maximal Poisson-Disk Sampling

Mohamed S. Ebeida, Anjul Patney, Scott A. Mitchell, Andrew A. Davidson,  
Patrick M. Knupp, John D. Owens

Sandia National Laboratories, University of California, Davis

Scott - presenter  
SIGGRAPH2011

# Maximal Poisson-Disk Sampling

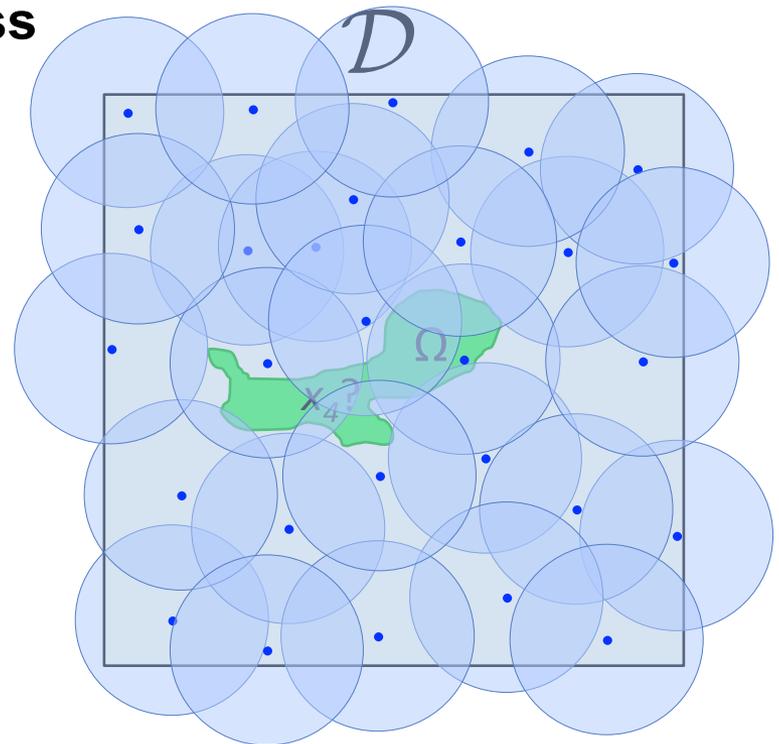
- What is MPS?
  - Dart-throwing
  - Insert random points into a domain, build set  $X$ 
    - With the “Poisson” process

Empty disk:  $\forall x_i, x_j \in X, x_i \neq x_j : \|x_i - x_j\| \geq r$

Bias-free:  $\forall x_i \in X, \forall \Omega \subset \mathcal{D}_{i-1} :$

$$P(x_i \in \Omega) = \frac{\text{Area}(\Omega)}{\text{Area}(\mathcal{D}_{i-1})}$$

Maximal:  $\forall x \in \mathcal{D}, \exists x_i \in X : \|x - x_i\| < r$



# MPS a.k.a.

- **Statistical processes**

- **Hard-core Strauss disc processes**

- **Non-overlap: inhibition distance  $r_1$**
- **cover domain: disc radius  $r_2$**

- **Nature**

- **Trees in a forest**

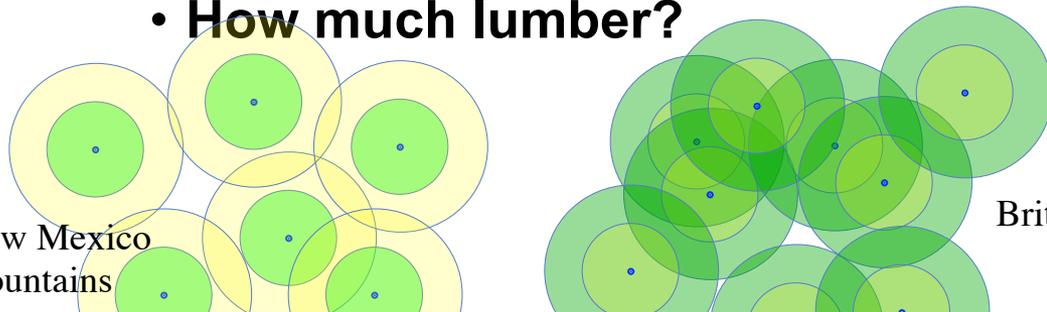
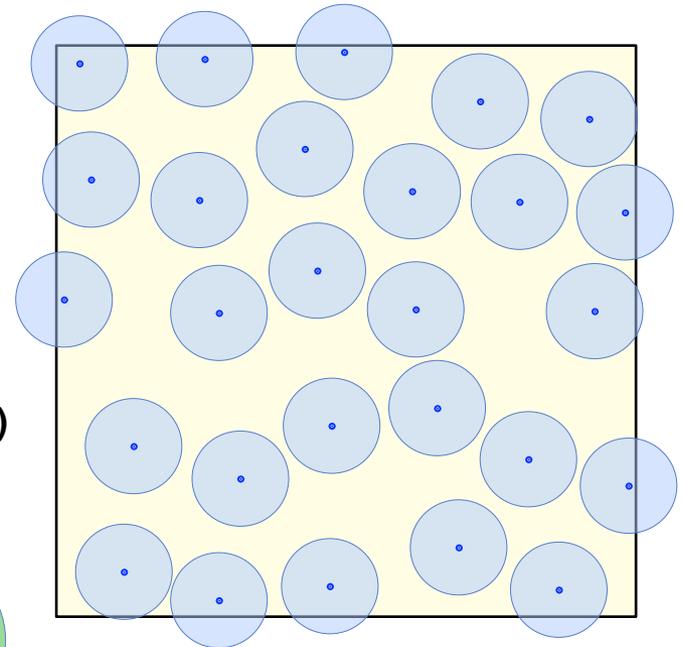
- **Variable disk diameter = tree size**
- **Points are tree trunks**
- **Disks are tree leaves or roots**

- **Given satellite pictures (non-maximal)**

- **How many trees are there?**
- **How much lumber?**

- **Random sphere packing**

- **Non-overlapping  $r/2$  disks**
- **Atoms in a liquid, crystal**



New Mexico  
mountains

British Columbia



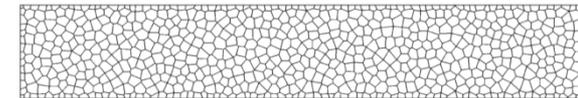
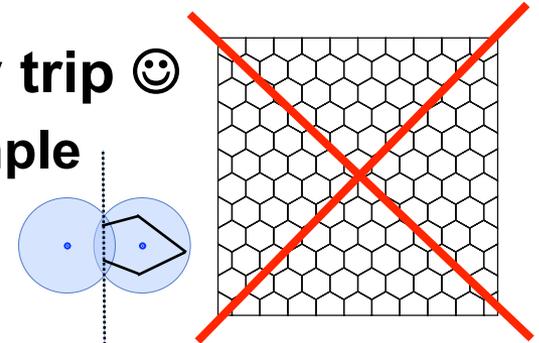
# What is MPS good for?

---

- **Graphics – sample points for texture synthesis**
  - **Generate blue noise distributions for anti-aliasing**
  - **Without Moire and other visible patterns**
- **Unbiased process leads to points with**
  - **No visible patterns between distant points.**
    - pairwise distance spectrum close to truncated blue noise powerlaw
  - **Our eyes sensitive to patterns**
  - **Randomness hides imperfections**
    - **stare at dry-wall in your house sometime, try to find the seams**

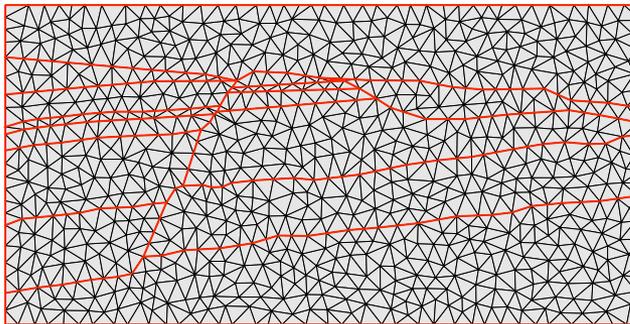
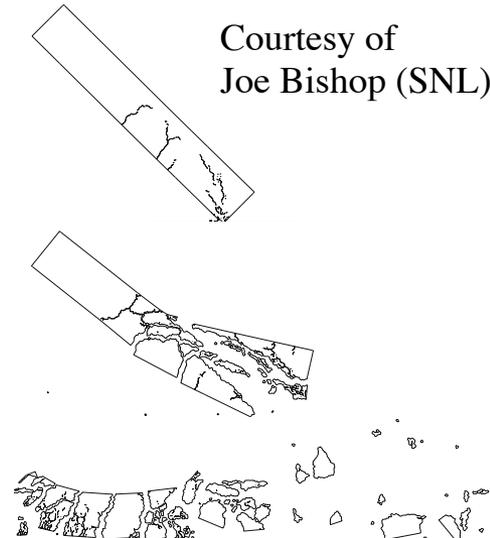
# What is MPS good for?

- **Physics simulations – why SNL paid for my trip ☺**
  - **Voronoi mesh, cell = points closest to a sample**
  - **Fractures occur on Voronoi cell boundaries**
    - Mesh variation  $\subset$  material strength variation
    - CVT, regular lattices give unrealistic cracks
      - **Unbiased sampling gives realistic cracks**
  - **Ensembles of simulations**
  - **Domains: non-convex, internal boundaries**

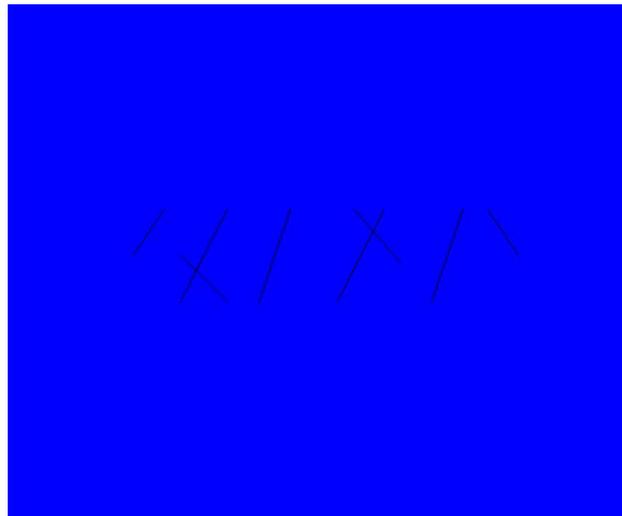


## Fracture Simulations

Courtesy of  
Joe Bishop (SNL)



Seismic Simulations  
maximal helps  $\Delta$  quality

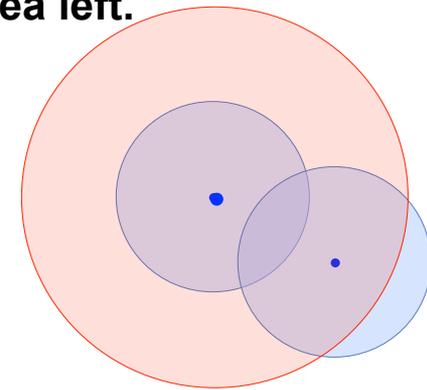
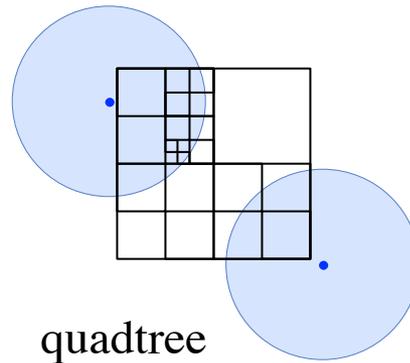


# Algorithm for MPS

- **Classic algorithm**

- Throw a point, check if disk overlaps, keep/reject
- Fast at first, but slows due to small uncovered area left.

Can't get maximal.



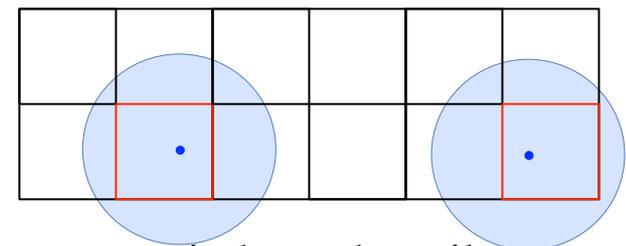
advancing front

- **Speedup by targeting just the uncovered area**

- Others use quadtrees to approximate the uncovered area
- Others use advancing front to sample locally
- Others use tiles to aid parallelism

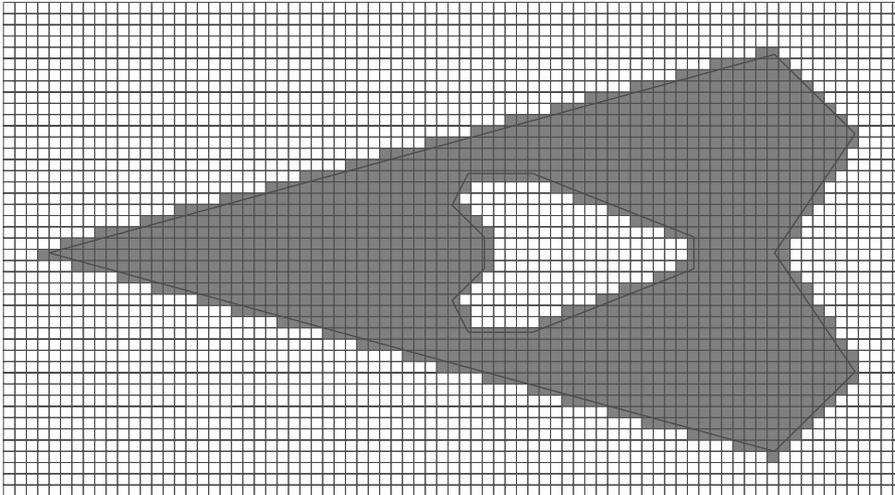
- **Common issues**

- Not strictly “unbiased” **process**
  - **Outcome** may be indistinguishable from an unbiased process’s outcome
- Not maximal: dependent on finite precision
- Memory or run-time complexity
- Ours is first provably bias-free, maximal,  $E(n \log n)$  time  $O(n)$  space

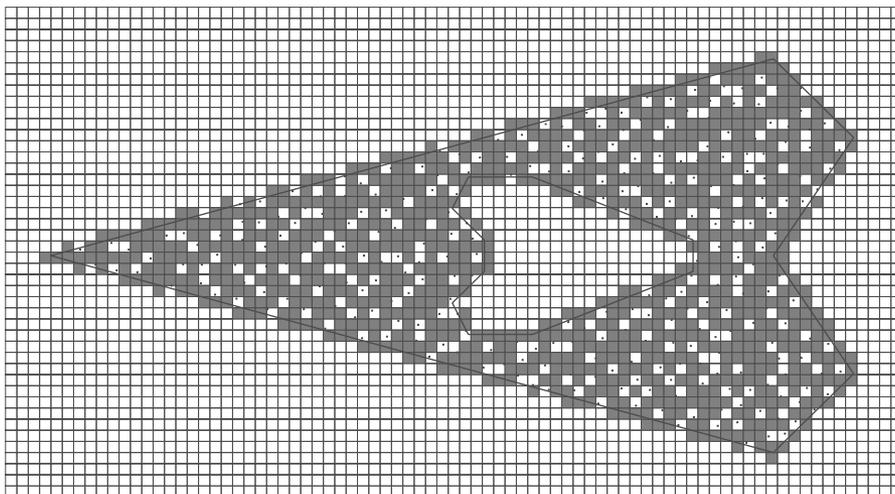


independent tiles

# Algorithm

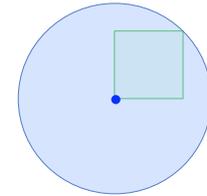


Initial Pool  $C$



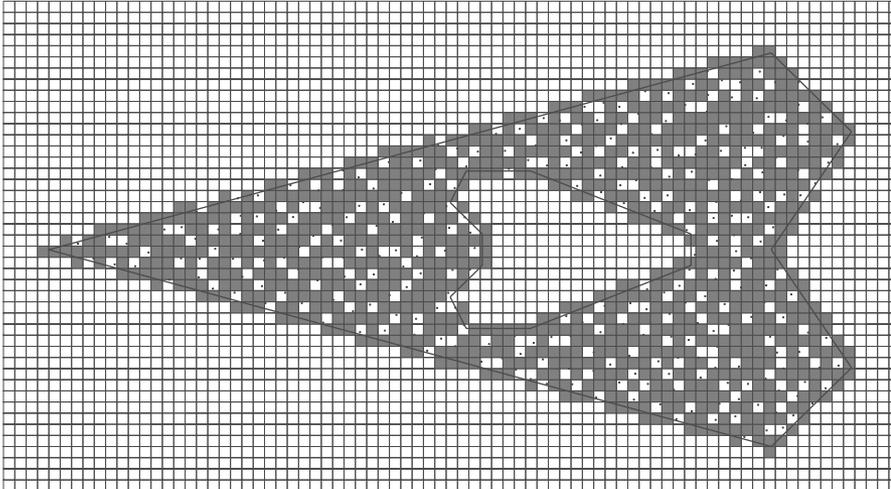
End of Phase I: white cells with a point

- Background square grid
  - Square diagonal =  $r$

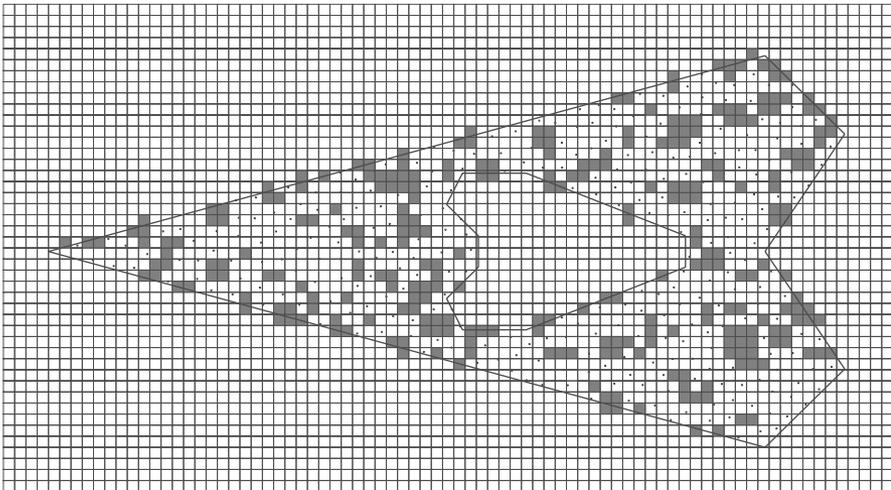


- Flood fill
  - Build pool of cells  $C$  : not-exterior to domain
- Phase I: quickly cover most of the domain
  - Pick a square from pool
  - Pick point in square
  - If point uncovered (likely)
    - Keep point
    - Remove square from pool
  - Repeat  $a|C|$  times

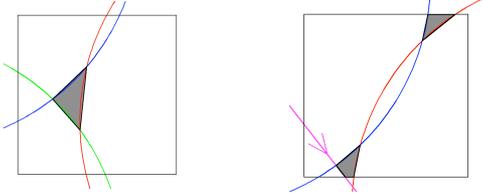
# Algorithm



End of Phase I: white cells with a point



Start of Phase II: dark cells not-covered

- Target remaining uncovered area
  - Construct square \ disks
    - Polygon easy surrogate for arc-gon
- 
- Replace pool of squares by polygons
  - Phase II: repeat
    - Pick polygon from pool
      - Weighted by its area (only log n step)
    - Pick point in polygon
    - If uncovered
      - Keep point
      - Remove polygon from pool
      - Update nearby polygons
  - Works well because
    - Voids are scattered
    - Small arc-gons are well approximated by polygons

# Algorithm Nuance - Phase II stages

- “Algorithm is simple,... in a good way” - Reviewer
- **Lazy update of polygons’ areas and pool, in “stages”**
  - More simple datastructures
  - No tree needed, flat array for pool, fewer pointers
  - Run-time proof gets more complicated

## Prior slide

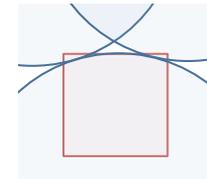
Phase II: repeat  
Pick polygon from pool  
    Weighted by its area (only log n step)  
Pick point in polygon  
If uncovered  
    Keep point  
    Remove polygon from pool  
    Update nearby polygons

## Lazy update

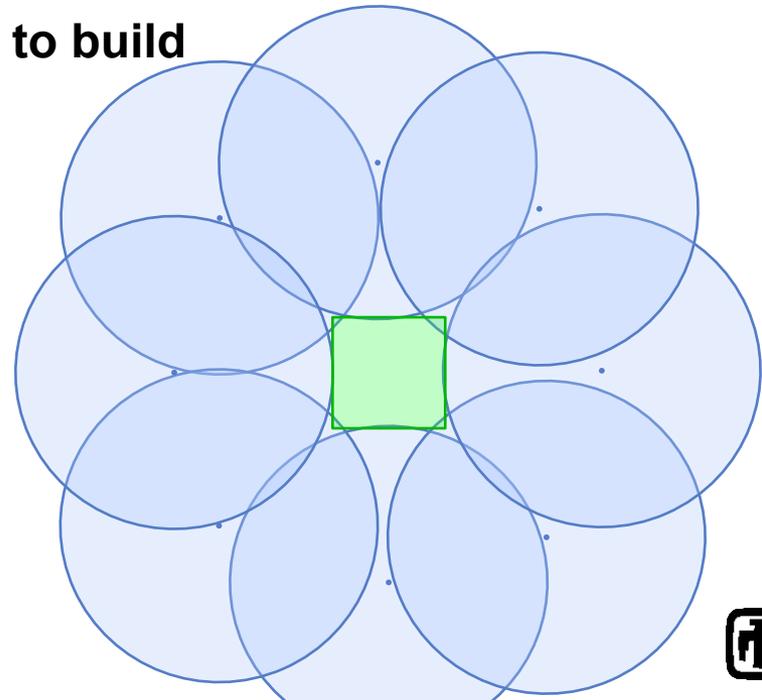
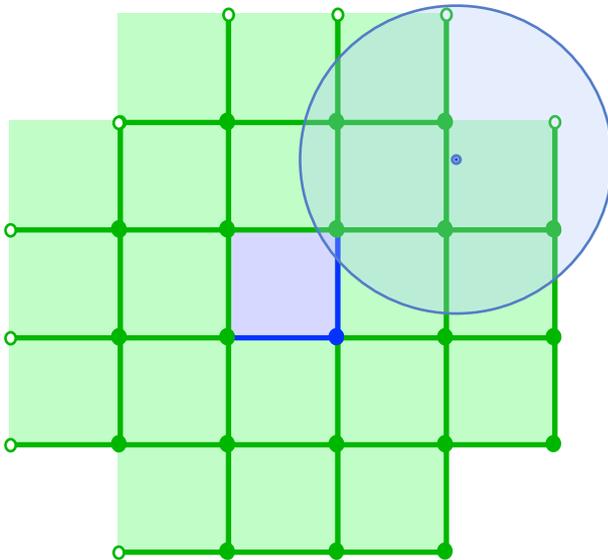
Phase II: repeat  
Repeat cIPool times  
    Pick polygon from pool  
        Weighted by its area (only log n step)  
    Pick point in polygon  
    If uncovered  
        Keep point  
New stage - update all polygons  
Rebuild pool and weights

# Complexity Proofs Sketch

- **WTS constant time & space per point**
  - Everything is local, and constant size
- #squares =  $\theta(\text{\#points\_in\_sample})$
- Sid Meier Civilization template
  - 21 nearby squares, 0 or 1 disks per square
    - By geometry,  $\leq 4$  voids per cell
    - By geometry,  $\leq 9$  (8?) disks bounding a void
- Constant time to check if point is uncovered
- Polygons are constant size, time to build

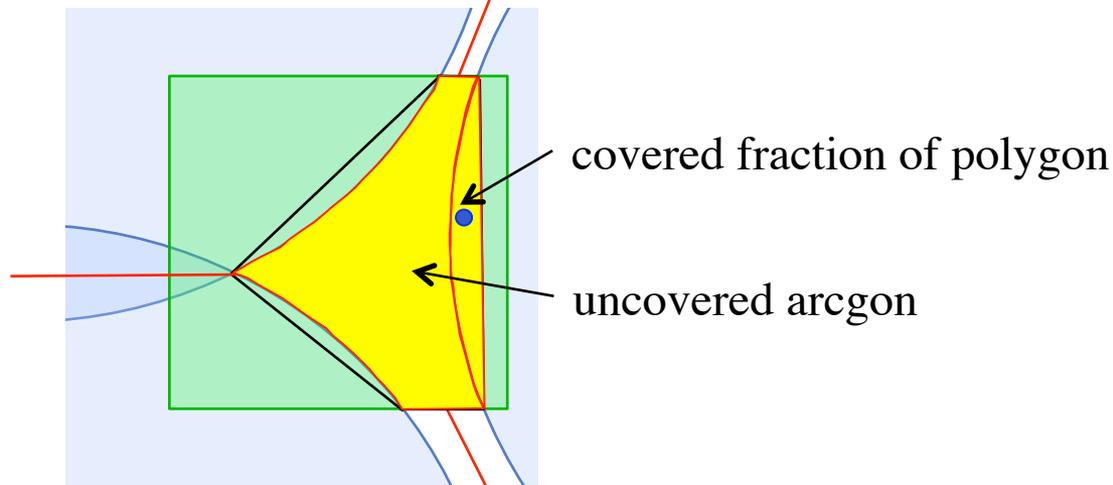


Four voids



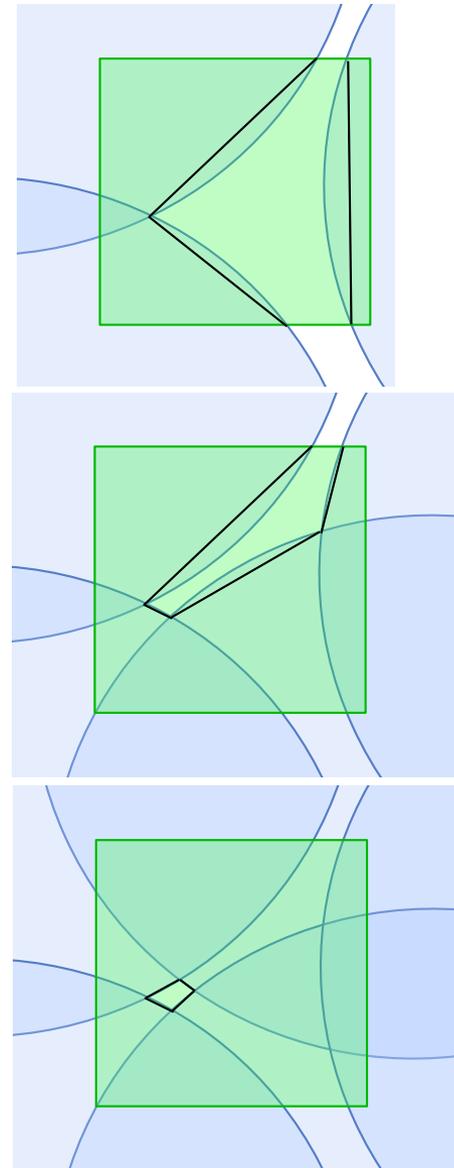
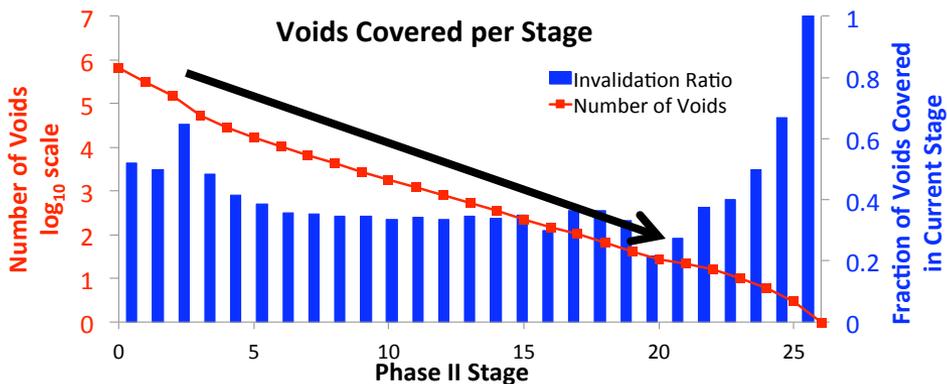
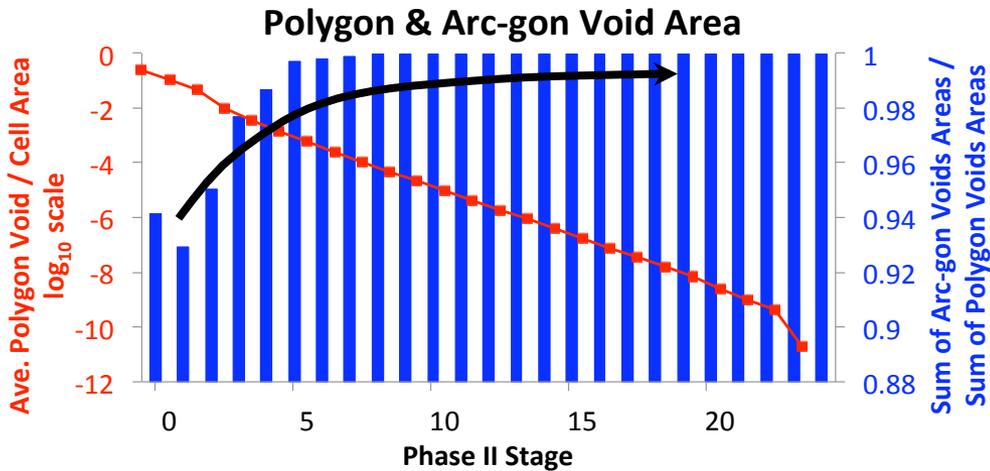
# Complexity Proofs Sketch

- **Constant work per generated point, but what about the rejected (covered) points?**
  - **Phase I,  $O(|C|)$  throws**
  - **Phase II**  
 $\text{Area}(\text{arcgon}) > c \text{Area}(\text{polygon}) \Leftrightarrow P(x_i : \text{uncovered}) > c$   
 $\Leftrightarrow \# \text{ accepted} > c_2 \# \text{ rejected}$
- **Via weighted Voronoi cell of a circle**
  - **Constant curvature and number of edges**



# Fewer Rejected Points Later

- Polygons → arcgon as voids get smaller
  - We get more efficient (contrast)



Algorithm progress



# Complexity

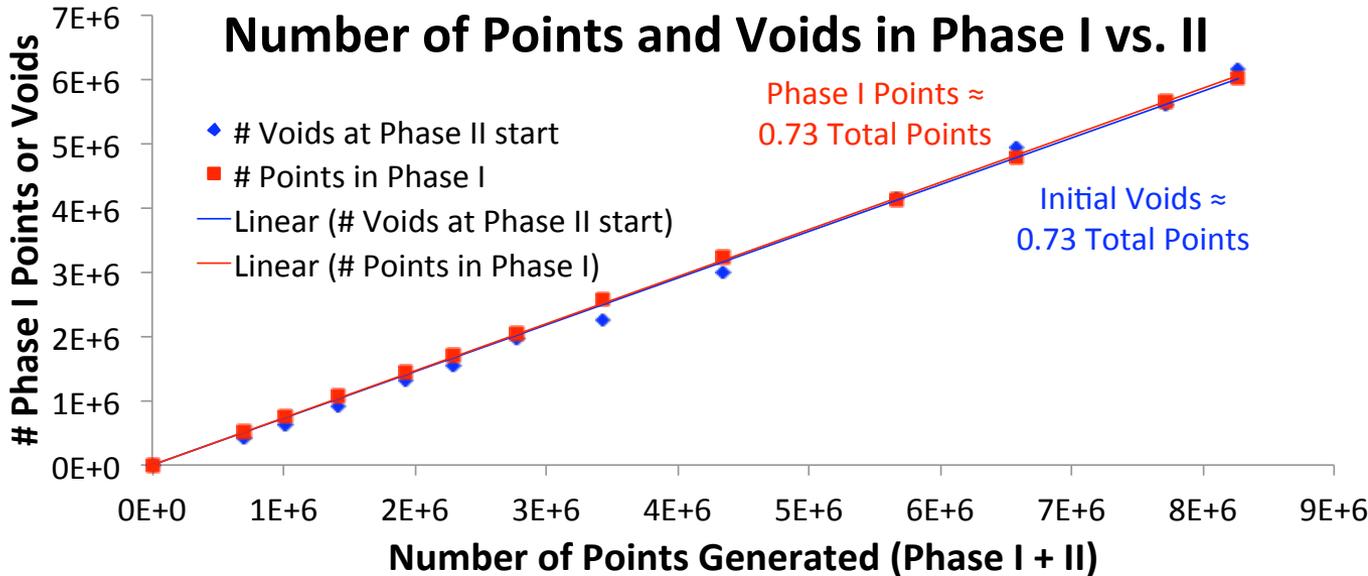
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- **Complexity – everything is local, all steps constant time**
  - except  $\log(n)$  to select a polygon, weighted by area
  - that is a relatively inexpensive step
  - constructing geometric primitives is the expensive part
- **Constant fraction of generated points are output points**

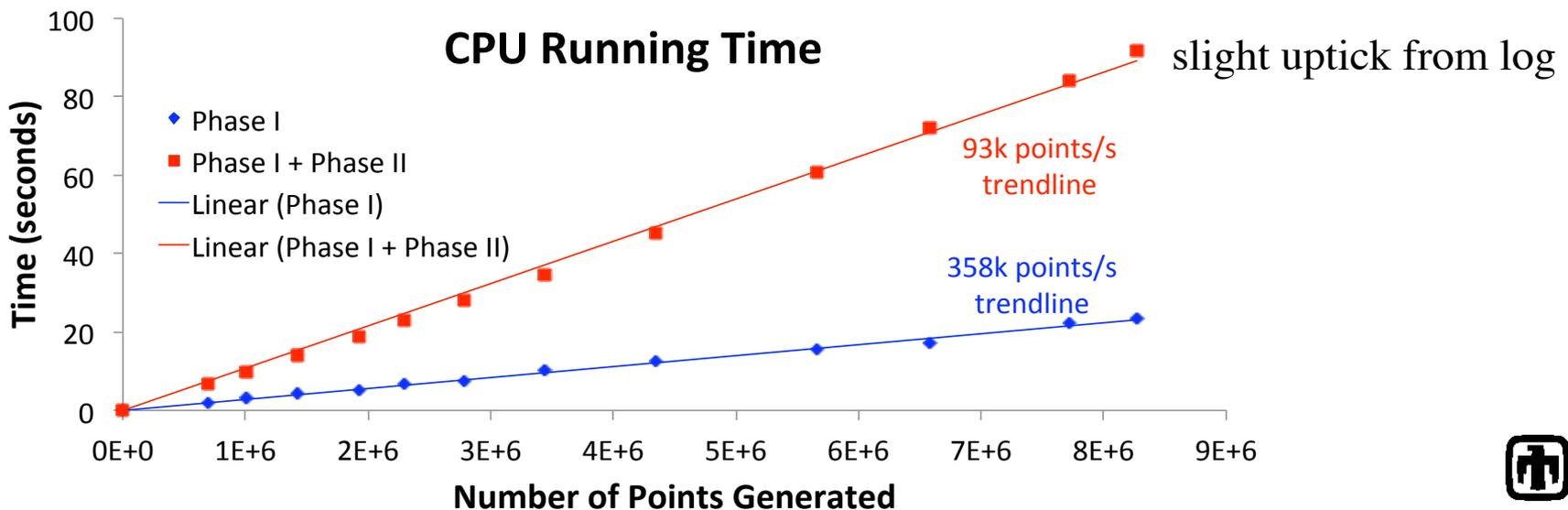
$$\text{Time} = E(Cn + cn \log n)$$

$$\text{Space} = O(n)$$

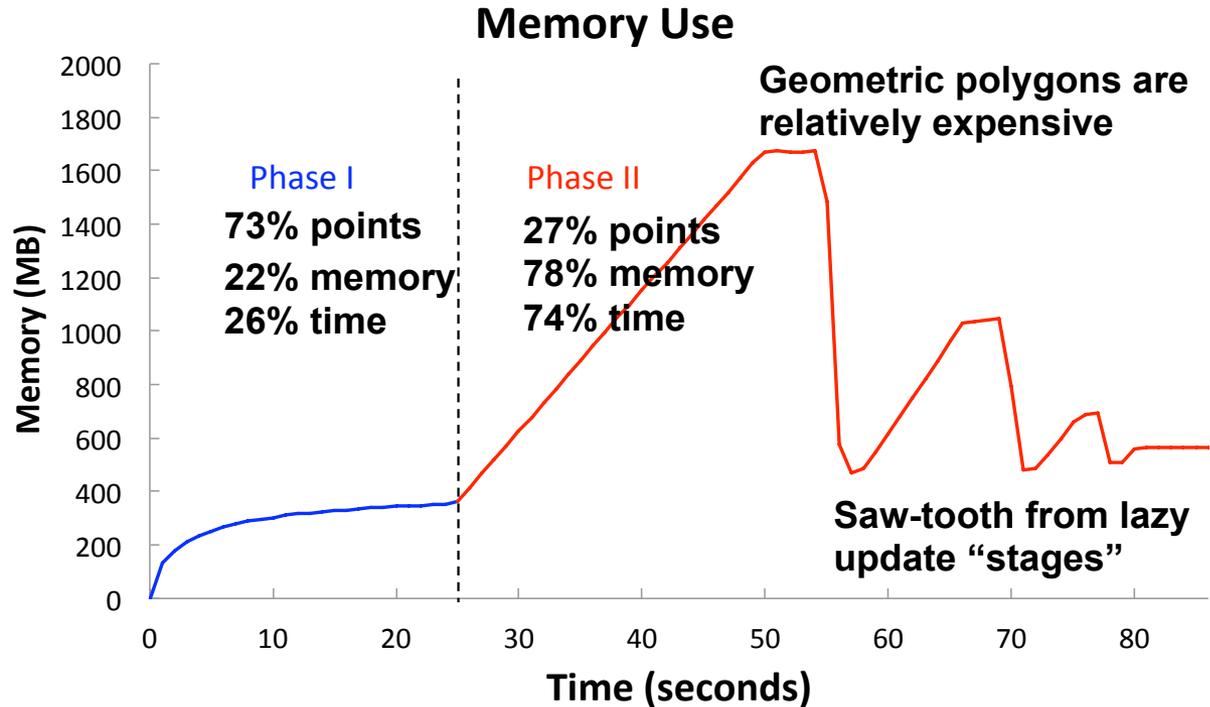
# Runtime – Why we do Phase I



- Phase I
  - 73% of points
  - 26% of runtime



# Serial Memory Use





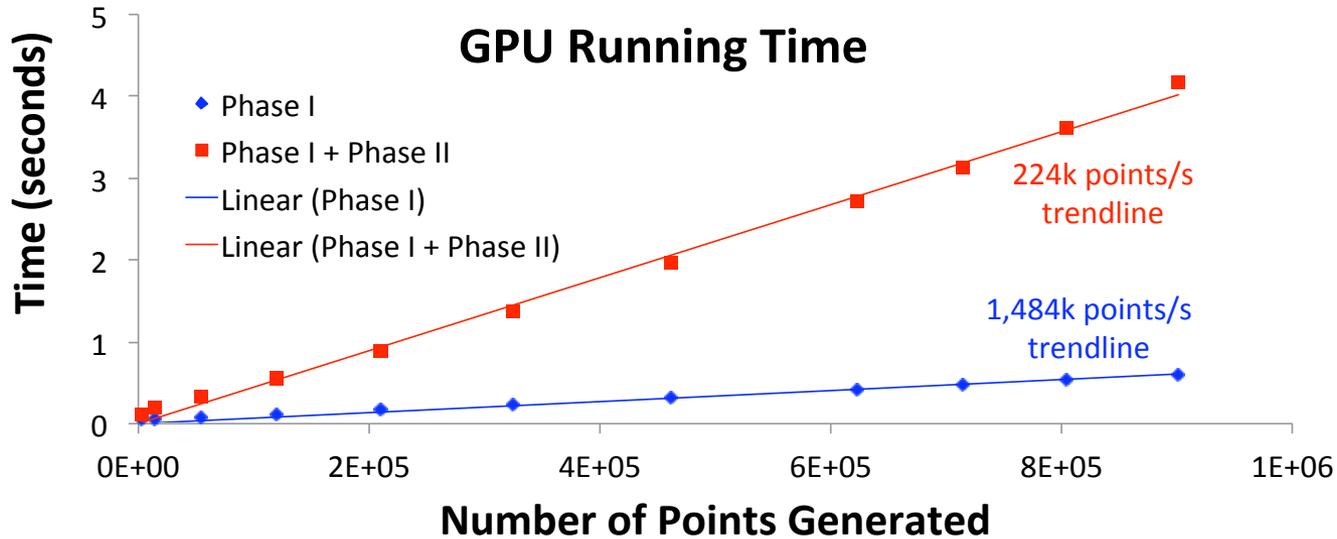
# GPU Algorithm

---

Points generated in parallel, conflicts resolved in an unbiased way

- Point buffers: candidate and final
- Phase I
  - Iterate: **synchronize at start of iteration**
    - Generate  $|C|/5$  candidate points
    - Square states: empty, test, accepted, done
      - Done = Point from prior iterations
      - Test = Point doesn't conflict with nearby "done" points, **compute in parallel**
      - Accepted = Point is **earlier** (id) than conflicting "test" points, **compute in parallel**
    - Migrate accepted points to done, otherwise remove
- Phase II
  - Construct polygons, **compute in parallel**
    - Squares "rejected" if covered by prior disks, has no polygon, no work to do
    - Split polygons into triangles
  - Proceed as Phase I, with triangles playing role of squares

# GPU Performance



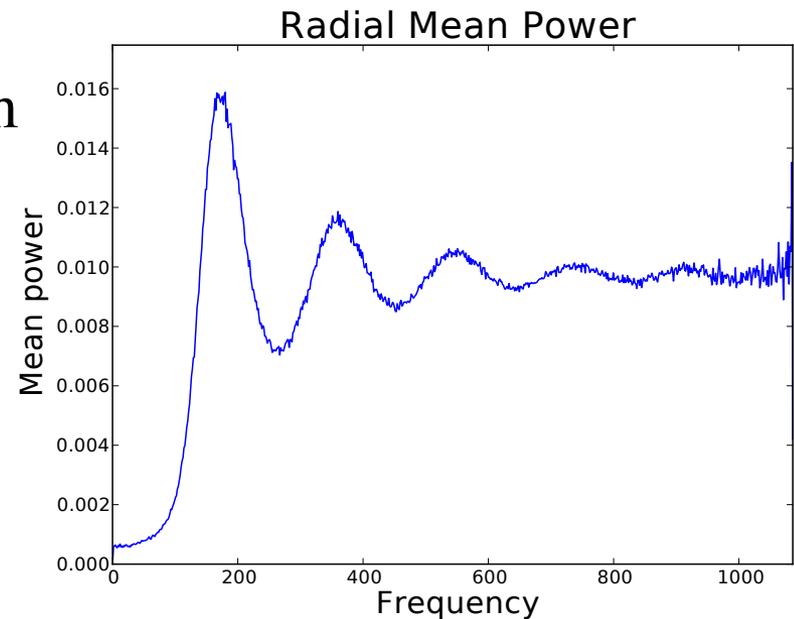
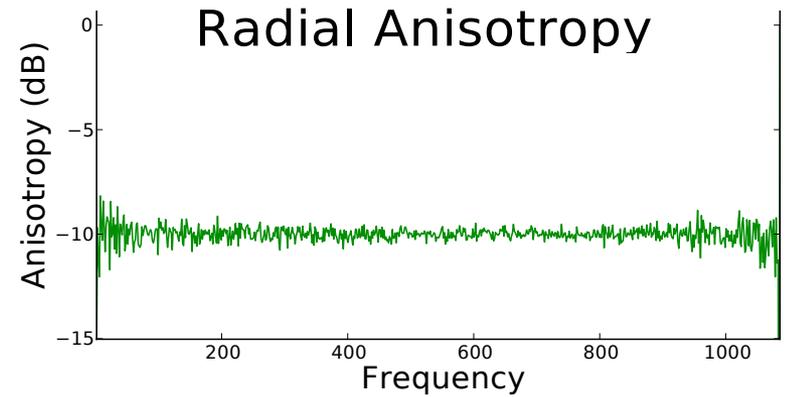
NVIDIA GTX 460

2.4x speedup over serial (6.7x memory bandwidth)

1 million points in 1 GB RAM

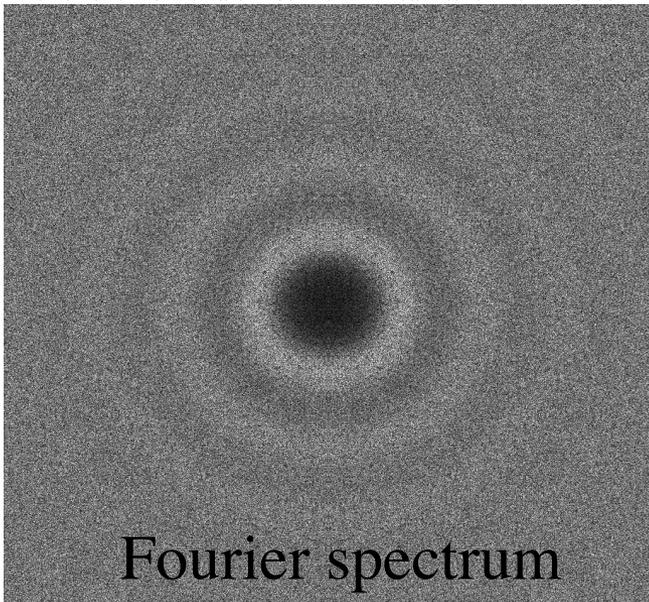
# Unbiased Parallel Sample

10k pts



Rings from  
inhibition  
radius

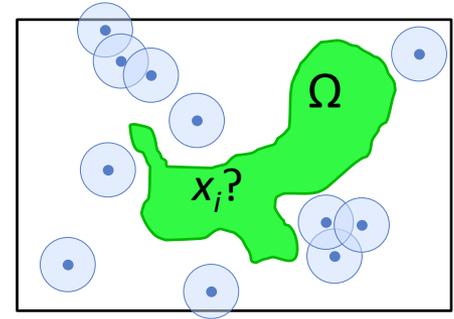
Fourier spectrum



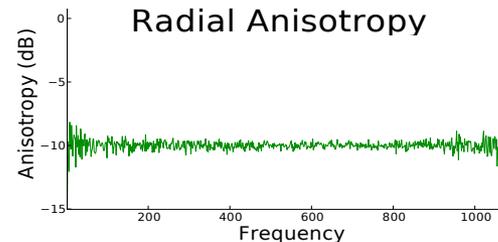
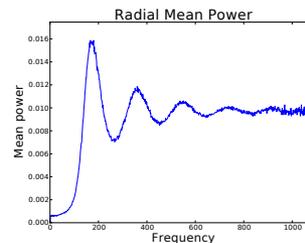
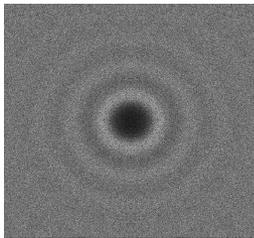
# “Unbiased” Opinion

- Unbiased as a description of (serial) process
  - insertion probability independent of location

$$P(x_i \in \Omega) \propto \text{Area}(\Omega)$$



- Unbiased as a description of outcome
  - pairwise distance spectra, blue noise



- Unbiased process leads to unbiased outcome, but so might other processes
  - Opinion: need something beyond “viewgraph norm”
  - Need metrics for “how unbiased is it”
    - Define spectrum  $S$  that is the limit distribution of unbiased sampling, and standard deviations.
    - Our process generated  $S'$ , and  $|S-S'| < 0.4 \text{ std dev}(S)$

# Synopsis of Contribution

- **Poisson-disk distributions**

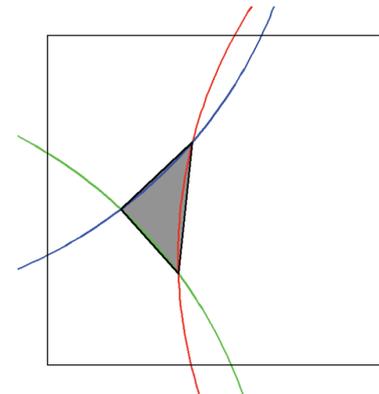
- Simple, efficient implementation
- Provable guarantees
  - Maximal
  - Unbiased
  - $O(n)$  space
  - $E(Cn + cn \log n)$  time

- **Domains**

- 2d
- Polygons with holes, non-convex

- **Algorithmic innovations**

- Two phases
  - I. fast to cover most of domain
  - II. careful to cover remainder
- Approximate uncovered “voids”, square  $\cap$  circles, with polygons. Careful weighting and selection





# Future

---

- **Extensions**
  - Could do away with polygonal approximation and weight and sample directly – every dart is a hit! (w/ Thouis Ray Jones)
- **Higher dimensions**
  - geometric primitives unappealing
  - prefer just use hypercubes
- **Thouis Ray Jones, jgt accepted paper**
  - model explicit time-of-arrival for each point
  - synchronize locally as needed
  - vs. unbiased by one dart at a time, inherently serial