

Space-filling lattice designs for computer experiments

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Outline

- Computer experiments and designs
- New type of lattice design
- Nested structure
- Application in predictive science



Many processes are investigated using computational models

- Many scientific applications use **deterministic** mathematical models to describe physical systems
- To understand how inputs to the computer code impact the system, scientists adjust the inputs to computer simulators and observe the response
- The computer models frequently:
 1. require solutions to PDEs or use finite element analyses
 2. have high dimensional inputs
 3. have outputs which are complex functions of the inputs
 4. require a large amounts of computing time
 5. have features from some of the above



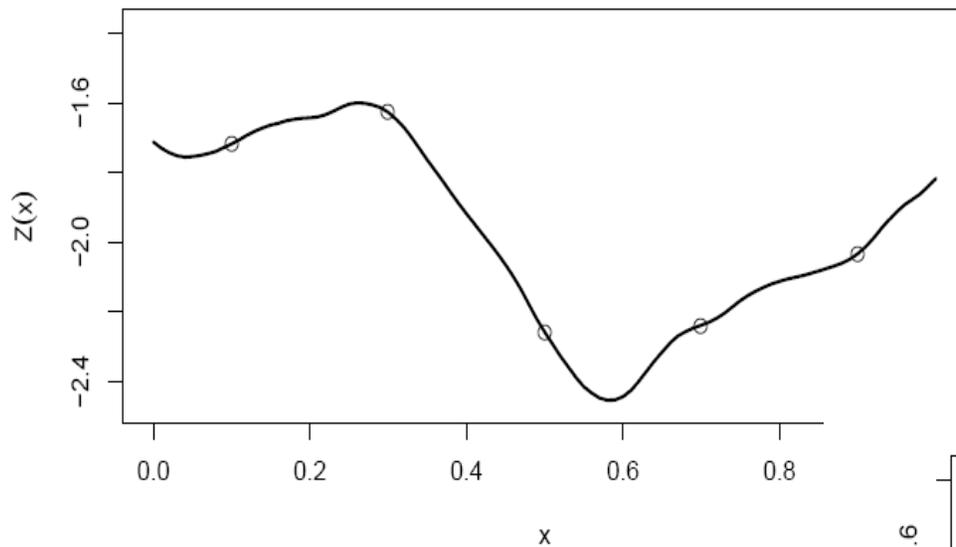
Use Gaussian processes for emulating computer model output

- Gaussian processes have proven effective for emulating computer model output (Sacks et al., 1989) and also data mining
- Emulating computer model output
 - output varies smoothly with input changes
 - output is essentially noise free
 - passes through the observed response
 - GP's outperform other modeling approaches in this arena



Why use a GP for emulation?

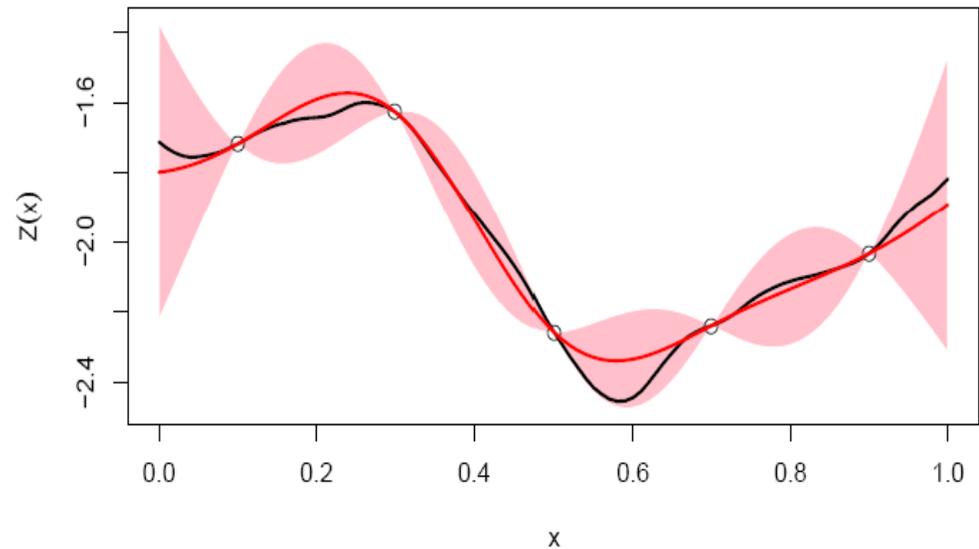
True function and observations



$$y(x) = \mu + z(x)$$

$$E(z(x)) = 0; \quad \text{var}(z(x)) = \sigma^2$$

$$\text{cov}(z(x), z(x')) = \sigma^2 R(x, x')$$



Applications of interest



- Upcoming space based cosmology missions promise exquisite measurements of the large-scale structure distribution of the Universe (e.g., including weak lensing, baryon acoustic oscillations, clusters of galaxies, and redshift space distortions)
- Currently exploring an 8-dimensional input space that, when combined with observations, should shed light into the initial conditions of the Universe and also the nature of dark energy



Applications of interest



- Will be running about 100 simulations that should take between 1 and 2 years to complete ... **can run several of these in sequence**



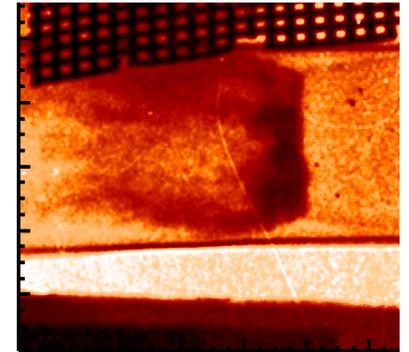
Applications of interest



- Will be running about 100 simulations that should take between 1 and 2 years to complete ... **can run several of these in sequence**
- Can investigate the response in intermediate stages while other simulations are running



Applications of interest



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Target Coord. X (μm)

- At the Center for Radiative Shock Hydrodynamics (CRASH), computational models were employed to simulate features of radiative shocks
- The CRASH codes consisted of high and low fidelity models
- It was helpful to run the high and low fidelity codes with the same inputs to explore the discrepancy between the two models (e.g., Qian and Wu, 2008)
- The low fidelity code was run at far more input settings (high fidelity design was **nested** within the low fidelity design)

Design for computer experiments

- Johnson et al. (1990) and others (e.g., Kunsch et al., 2005) demonstrate that designs with good *space-filling* properties are essential for prediction using GPs
- Latin hypercube designs (McKay et al, 1989) and other variants (Tang, 1993) have proven popular
- Designs based on Cartesian lattices have also been proposed (Beattie and Lin, 2004; Qian and Ai, 2010)
- Single state lattice designs have been discussed (Bates et al., 1996; Pronzato and Müller, 2012)
- **Here, a new type of lattice design is proposed**



Would like our designs to have specific properties

1. Would like n -run designs where each design point is a d -dimensional input vector to the computer model
2. In our setting would like experiment designs (D) with good d -dimensional space-filling properties
3. Would like the designs to have the nesting property
 - Important for applications where good **intermediate-stage designs** are required, as well as the **final experiment design**
 - Important for applications with high- and low-fidelity simulators where the **high-fidelity simulator design is a sub-set of the larger, low-fidelity simulator design**



Suggestion ...

- Use a lattice
- For one-stage designs, can use already computed lattices (Conway and Sloane, 1999) that have good space filling properties
- Not quite as easy as you might think ...
- Is more challenging for our setting where nesting is required



Notation and definitions

- A **point lattice** is an infinite, discrete set of points in \mathbb{R}^d that is constructed from integer multiples of a set of basis vectors in the columns of a $d \times d$ **generating matrix**, \mathbf{G} ,

$$\Lambda(\mathbf{G}) = \mathbf{G}\mathbb{Z}^d = \{\mathbf{G}\mathbf{k} : \mathbf{k} \in \mathbb{Z}^d\} \subset \mathbb{R}^d$$

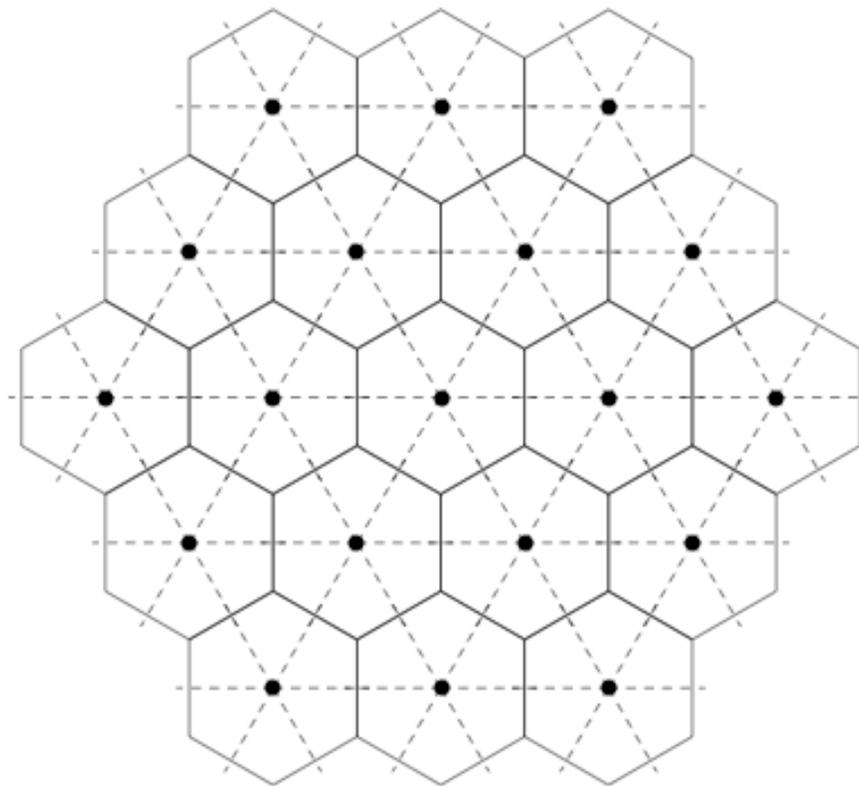
- A **lattice design** $D(\Lambda, \mathcal{M}, \mathbf{p}) = \mathcal{M} \cap \{\Lambda + \mathbf{p}\}$ is the intersection of a point lattice and region $\mathcal{M} \subset \mathbb{R}^d$ that is shifted by a vector, \mathbf{p}
- See Conway and Sloane, 1999 or Patterson, 1954

Fun facts about lattices

- As a linear transformation of the integers, so lattices inherit their abelian group structure
 - ... this implies that the neighborhood around each lattice point is the same
 - This region is also called the **Voroi cell**
- The space between the lattice points are described by the **fundamental parallelepiped**



Fun facts about lattices



Example

- Factorial design (Cartesian lattice):
- Have d inputs with levels $\mathbf{S} = (S_1, S_2, \dots, S_d)$
- Here $\mathbf{G} = \mathbf{I}_d$ and the lattice is $\Lambda(\mathbf{G}) = \mathbf{G}\mathbb{Z}^d = \{\mathbf{G}\mathbf{k} : \mathbf{k} \in \mathbb{Z}^d\} \subset \mathbb{R}^d$
- Region of interest is $[0,1)^d$ scaled by $\text{diag}(\mathbf{S})$



We are looking for specific designs

- A sequence of designs, is said to be nested if

$$D_l \subseteq D_{l+1} \text{ for all } l \in \mathbb{Z}.$$

- Increasing l is called a **refinement** and decreasing l is called **coarsening**



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THE END



More notation and definitions

- A **dilation matrix**, $\mathbf{K} \in \mathbb{Z}^{d \times d}$, with $|\det \mathbf{K}| = \beta > 1$, applied to a lattice forms a nested sequence of lattices, $\Lambda_{l-1} \subset \Lambda_l$, via $\Lambda_{l-1} = \Lambda_l \mathbf{K}$



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- Sub-sampling a lattice such that the chosen sample is **also a lattice** is performed by right-multiplying an integer dilation matrix onto G



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- For this setting, an **admissible dilation matrix** is one where (a) \mathbf{K} is a dilation matrix; (b) magnitude of all eigen-values of \mathbf{K} are larger than 1; and (c) $\det \mathbf{K} = \alpha^d$, where α is the eigen-value for \mathbf{K}

Theoretical results we can prove

- In $[0,1)^d$, the expected number of lattice points is $1/\det \mathbf{G}$
- \mathbf{K} must be an integer matrix
- For refinement (i.e., l goes up) the volume of the fundamental parallelepiped decreases by $|\det \mathbf{K}| = \beta$
- Can show that to get a nested lattice $\beta > 1$, thus best refinement is to half the volume between points as the run-size is doubled



Why do all this fancy stuff?

- Dyadic sub-sampling is impossible for Cartesian lattices with $d > 2$
- For Cartesian lattices, number of lattice points grows exponentially with dimension
- **The main idea is to:**
 - use more general, non-diagonal generators, \mathbf{G}
 - allow for sub-sampling rates that are 2, 3, 5, ... (2 is most useful)
 - Equivalently, means we can start with a coarse lattice and find new ones with 2, 3, 5 ... times more points



Why do all this fancy stuff?

- **Benefits:**
 - Sometimes can use general bases for known best packing or covering lattices, leading to a direct construction of maxi-min or mini-max designs, respectively
 - can consider virtually any run size
 - allows a sequence of designs that can be used in practical applications



How do we find designs

- Assume that the design region is $[0,1)^d$
- n is the experiment run size
- Looking for a non-singular lattice generating matrix G that, when sub-sampled by a dilation matrix K with reduction rate $\beta = |\det K|$
- Need to find
 1. K
 2. G
 3. **Shift, rotation and scaling** to fit n points in the design region



How do we find designs

- Need some more theory:
- Restrict attention to designs where sub-sampled lattice is a scaled or rotated version of the original lattice
- **GK=QG**
- Preserves nice geometric properties
- Imposes restrictions on **K**



How do we find designs... more theory

- The restriction ($\mathbf{G}\mathbf{K}=\mathbf{Q}\mathbf{G}$) implies the choice of \mathbf{G} up to rotation and scale ... reason is that this implies that \mathbf{K} and \mathbf{Q} have same characteristic polynomial $\det(\mathbf{K} - \lambda \mathbf{I})$
- Can prove that: (i) for even d , there are 5 different \mathbf{K} ; and (ii) for odd d there is only 1 \mathbf{K}
- We can still warp these \mathbf{G} and we do so to optimize a desirable property (e.g., mini-max, maxi-min, correlation between columns of the design matrix)
- Finally, \mathbf{G} is scaled so that $\mathbf{G}^* = \det \mathbf{G} = 1/n \dots$

How do we find designs

- Finally, we can use the generating matrix \mathbf{G}^* and \mathbf{K} to construct our lattice design
- However, the number of points in the region of interest is only expected to be n
- So, we randomly rotate \mathbf{G}^* and also shift the lattice to achieve the desired run-size in $[0,1)^d$

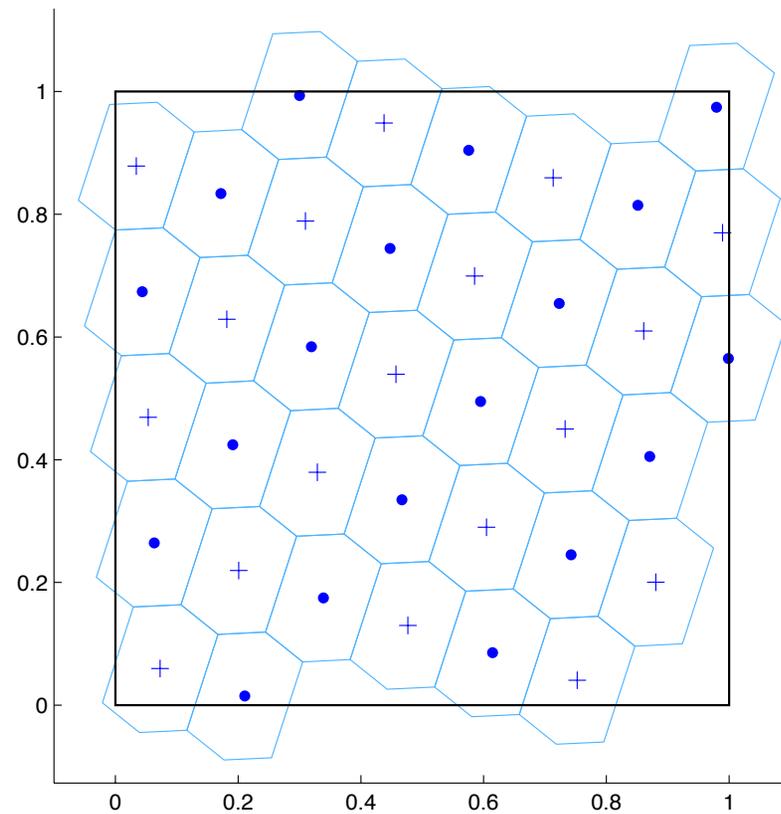
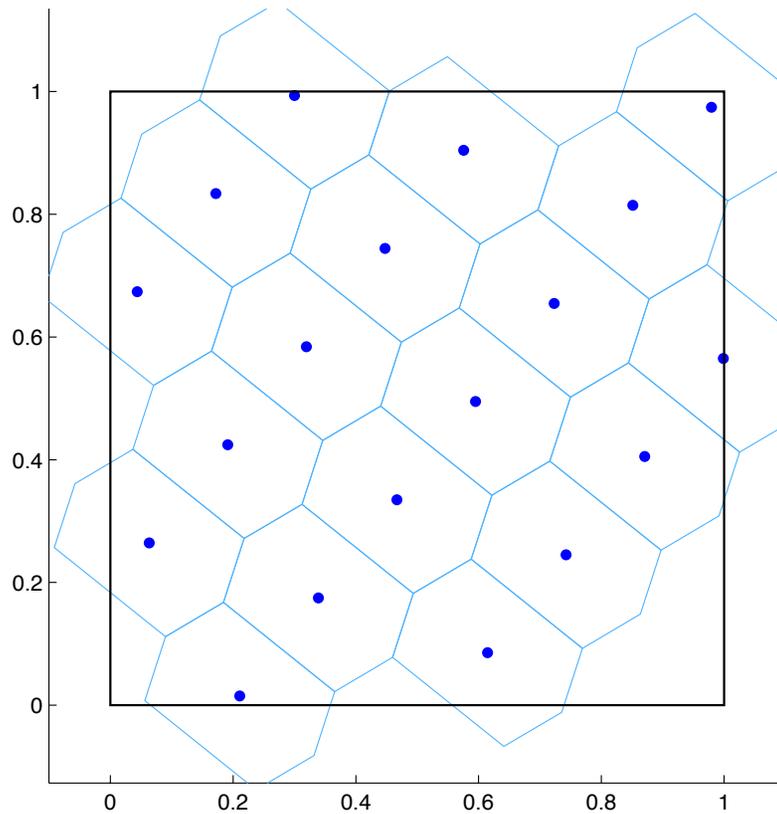


Summary

1. For even and odd d , can write out formulas for \mathbf{K}
2. Can compute $\mathbf{G} = \mathbf{J}_d \mathbf{P}^{-1}$
3. **Shift, rotation and scaling** to fit n points in the design region



Example ($d=2$; $n_1=19$; $n_2=2n_1$);



Re-cap

1. Have proposed a new type of lattice design that is useful in a variety of applications
2. Can be used to find designs with good space filling properties
3. Can find large designs from small ones
4. We are pre-computing good bases for different d



Thanks to all of you



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