## **Emulators and Design**

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https://github.com/andrewcparnell/intro\_emulators

#### Introduction

- In this part of the course we will cover how to choose the input values at which to run our simulator
- ► Recall that the simulator is slow to run and we can only afford a small number of runs
- We need to choose the 'best possible' input values to run the simulator
- ► The values that we choose will become the inputs (features) to our machine learning model; the outputs will become the targets
- ▶ How do we choose these best values?

# Choosing the total number of runs of the simulator

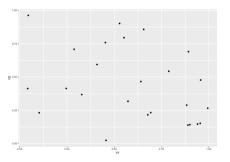
- ► Choosing the total number of runs is not really a mathematical/statistical problem, but it needs to be as large as possible
- ▶ It depends on the speed of the simulator, and how much patience/super computer time we have to spend on running it
- With some assumptions about the variability of the simulator surface it is possible to work out the expected uncertainty in the emulator for a given number of runs, but in the end it always comes down to doing as many as you possibly can

Choosing which input values to run the simulator at is a much harder problem. Here are some ideas. . .

#### Bad idea number 1 - random values

We could just choose random values across the input space:

```
N_sim <- 25
N_input <- 2
X <- matrix(runif(N_sim * N_input), ncol = N_input, nrow = N_sim)</pre>
```

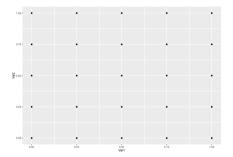


This is bad because we might miss lunch chunks of the input space; and it gets worse in higher dimensions

#### Bad idea number 2 - grids

Alternatively we could grid up the input space to cover the full region

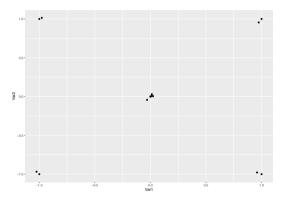
```
x <- seq(0, 1, length = sqrt(N_sim))
X <- expand.grid(x, x)</pre>
```



... but this is also bad because (a) one or more of the variables might not be important (so harder to identify non-linear effects), and (b) we might have better information about the likely values of the variables

## Bad idea number 3 - traditional design

Here's a Central-Composite (similar to Box-Behnken) design



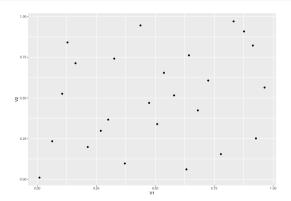
 $\dots$  but since the simulator is deterministic there is no point in running it at the same input points more than once

## A better idea - Latin hypercubes

A better method is a Latin hypercube design

library(lhs)

X <- maximinLHS(N\_sim, N\_input)</pre>

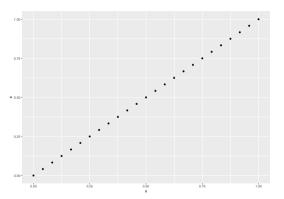


Think of dividing up the input space into horizontal and vertical bands, and picking one one value that covers each row and each column

7 / 14

## Bad Latin hypercubes

Actually, that idea doesn't work very well, because this is also a valid Latin hypercube design:

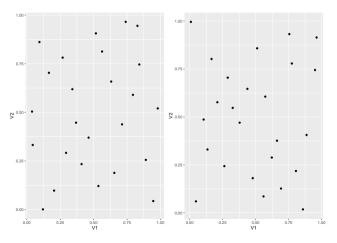


We want a design that is both a Latin hypercube and fills as much of the space as possible

## Good Latin hypercubes

Try to opimise the sample by finding a Latin Hypercube sample that maximises the minimum distance between design points

We can generate lots of these with the function maximinLHS:



## Designs for example 1

▶ Recall example 1: our simple sine wave with 2 inputs:

```
f <- function(x1, x2) {
  return(10 * sin(pi * x1 * x2))
}</pre>
```

Suppose we are willing to run this simulator 20 times. Create the design with:

```
n_grid <- 20
n_inputs <- 2
initial_grid <- maximinLHS(n_grid, n_inputs)</pre>
```

## Designs for example 2

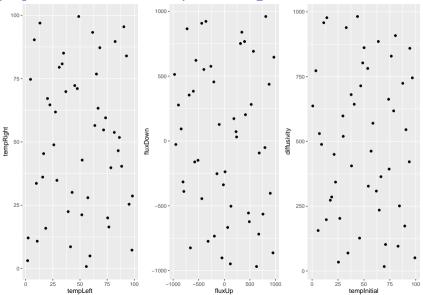
Recall example 2 is our 2D temperature diffusion with 6 inputs

Suppose we are willing to run this 50 times

```
n_grid <- 50
n_inputs <- 6
initial_grid <- maximinLHS(n_grid, n_inputs)</pre>
```

... harder to plot this as now in 6 dimensions

## Some 2D projections of the example 2 design



## Where to get more information on design

There are many possibilities when it comes to design in emulation:

- More advanced ways of spreading out the design and ensuring you do not miss important parts of the space
- More advanced ways of taking into account knowledge about the likely values of the inputs. For example, we might be able to guess a probability distribution for the input variables and use this to 'target' values in the design
- See Chapter 5 of The Design and Analysis of Computer Experiments for more detailed discussion
- We will stick to using Maximin Latin hypercube samples for our emulator

## Summary

- We need to choose how many total runs we can afford
- Lots of traditional design ideas do not work well for emulator design
- ► Remember that the simulator is deterministic not much point running at the same values twice! And try to avoid gridded values if you can
- Using Maximin Latin hypercube samples a good default choice
- Next: building the emulator...