

# Using Simulation to Present Complex Topics

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## Introduction: Why Simulate?

# Why Simulate?

## Common Questions:

- Is it worthwhile to teach programming?
- Wouldn't writing a simulation be hard?
- Why not just use theorems and proofs and other such nonsense?

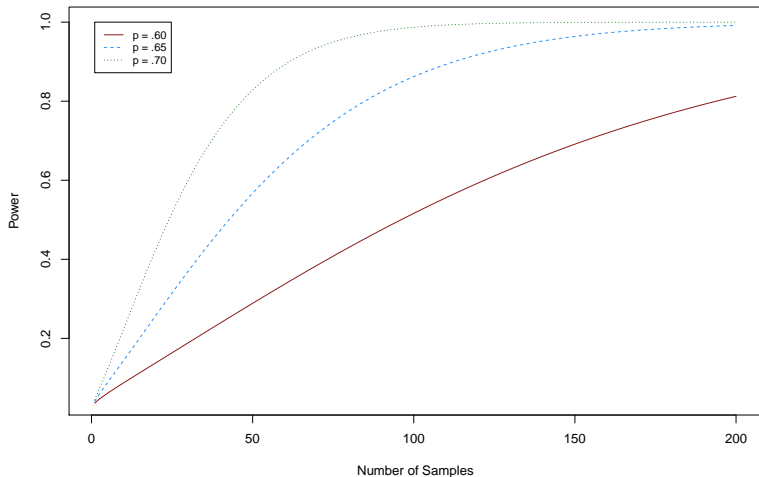
# Why We Love Math

## The Big Questions

- Why do we love math?
- What drives students to love math?
- How can we utilize this information to enhance student's learning?

# Why We Love Math: GRAPHS

Power versus Number of Samples for Sign Test



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Why We Love Math: FUNctions and Equations!

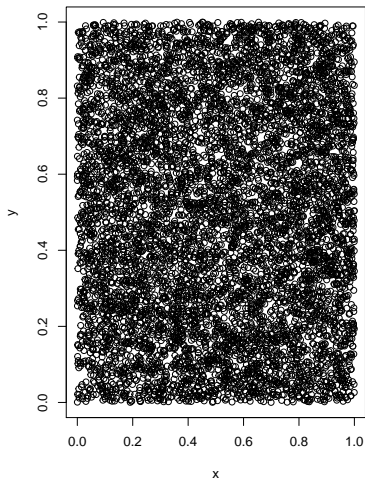
# Senior Research Project

Assignment: Find the probability that a percolating cluster appears in Mat'ern II Point Process distributed  $\text{Poisson}(\lambda)$

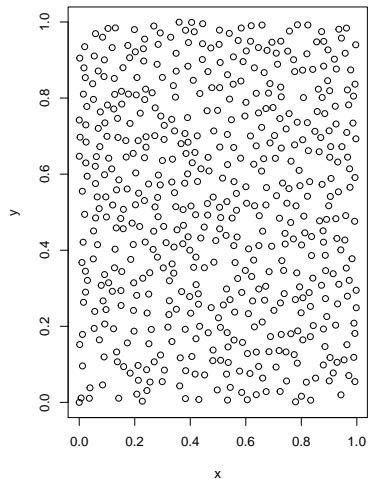
- What does it mean?
- How do I solve this?
- What's a Mat'ern?
- What's a percolating cluster?

# Senior Research Project

**Poisson Process**



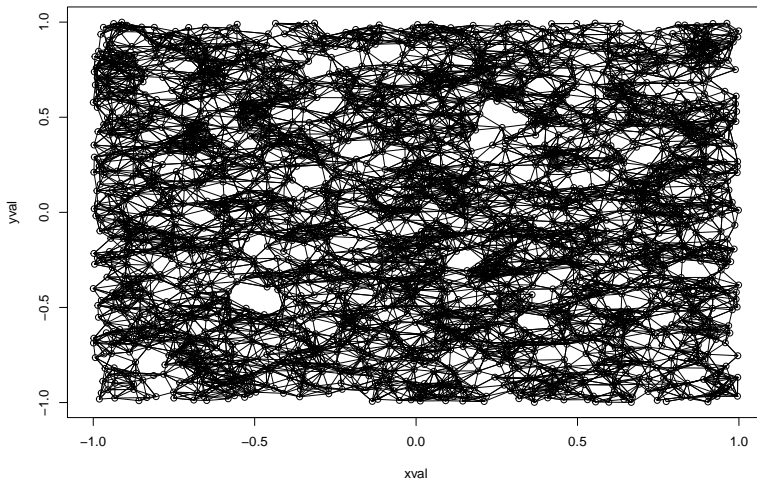
**Matern II**





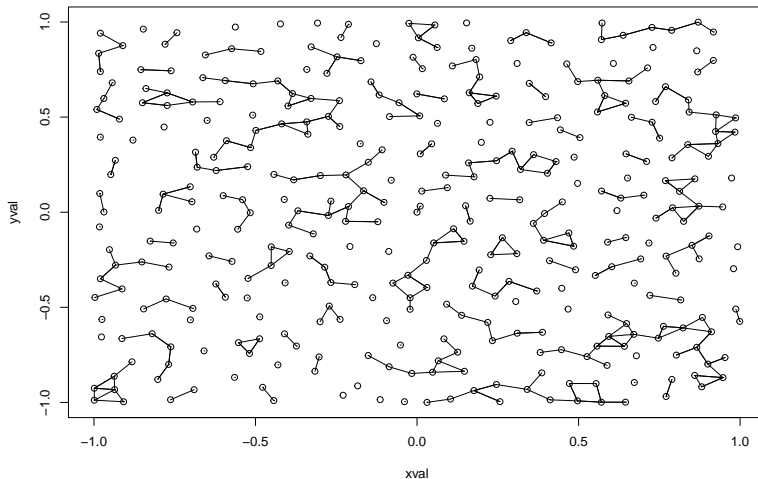
# Senior Research Project

Matern II,  $a = (4/5)r$



# Senior Research Project

Matern II,  $a = (4/5)r$



# The “Traditional” Statistics Course

- Measures of Center
- Sampling
- Regression
- Hypothesis Testing

# More Intense Problems

- Non-Normal Distributions
  - Poisson
  - Binomial
  - Exponential
- Central Limit Theorem

## The “Traditional” Statistics Course

# Measures of Center

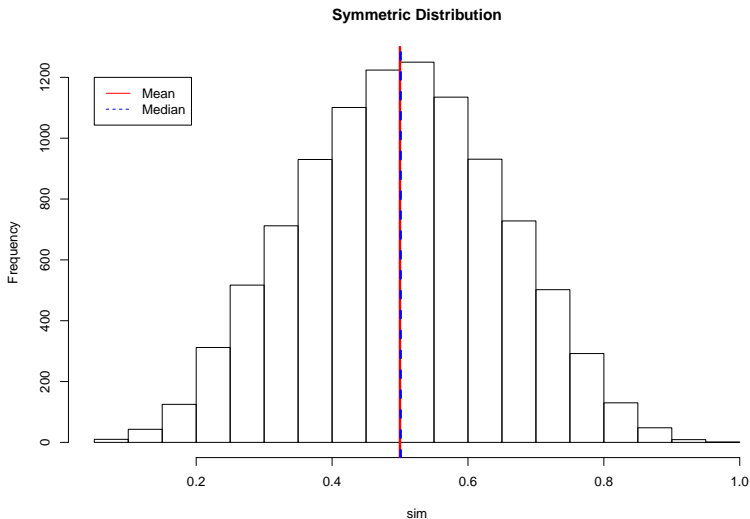
- Measures of center usually taught in respect to a symmetric distribution, left skewed distribution, and a right skewed distribution.
- How can we improve this information through simulation?
- More specifically, how can we show this without having data on hand?

# Measures of Center on a Symmetric Distribution

## The Simulation

```
sim <- rbeta(10000,5,5)
hist(sim, main = "Symmetric Distribution")
abline(v = mean(sim), col = "red", lwd = 3)
abline(v = median(sim), col = "blue", lwd = 3, lty = 2)
legend(0.05, 1200, c("Mean","Median"), col = c("red", "blue"))
```

# Measures of Center on a Symmetric Distribution

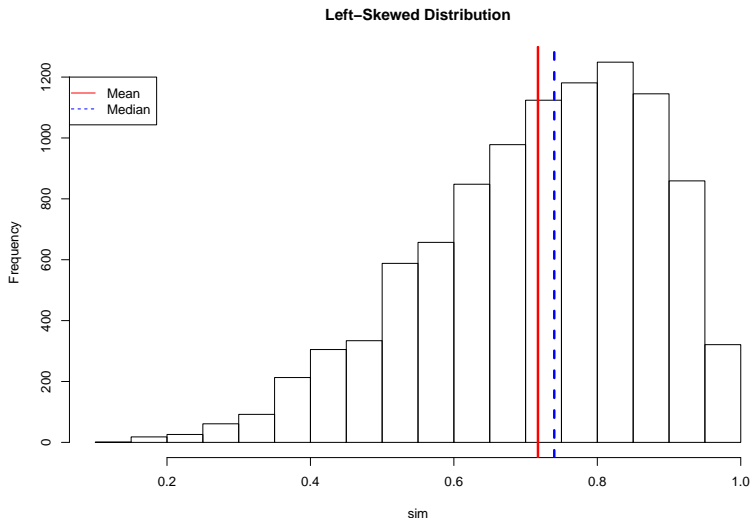




## Measures of Center on a Left-Skewed Distribution

```
sim <- rbeta(10000,5,2)
hist(sim, main = "Left-Skewed Distribution")
abline(v = mean(sim), col = "red", lwd = 3)
abline(v = median(sim), col = "blue", lwd = 3, lty = 2)
legend(0.05, 1200, c("Mean","Median"), col = c("red", "blue"))
```

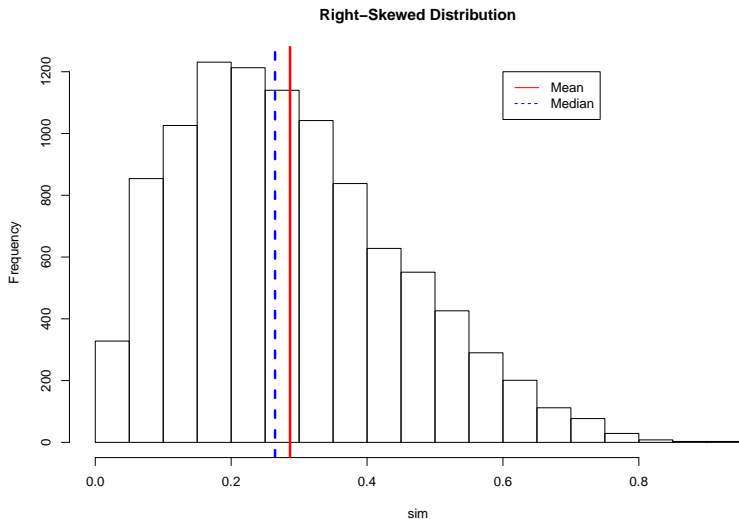
# Measures of Center on a Left-Skewed Distribution



## Measures of Center on a Right-Skewed Distribution

```
sim <- rbeta(10000,2,5)
hist(sim, main = "Right-Skewed Distribution")
abline(v = mean(sim), col = "red", lwd = 3)
abline(v = median(sim), col = "blue", lwd = 3, lty = 2)
legend(0.05, 1200, c("Mean", "Median"), col = c("red", "blue"))
```

# Measures of Center on a Right-Skewed Distribution



# Sampling in R

The true beauty of simulation lies in sampling (which is the core of most statistics).

- R can simulate from any notable distribution (and some non-notable!)
- R can then randomly sample from any of these distributions
- We can show elements of sampling empirically to students through these two easily understandable parts of R

# Sampling Examples

```
sim <- rnorm(10000)
cat("Mean of Simulation is:", mean(sim), "\n")
```

```
## Mean of Simulation is: 0.006586892
```

```
sim_sample <- sample(sim, 1000)
cat("Mean of Sim Sample is:", mean(sim_sample), "\n")
```

```
## Mean of Sim Sample is: -0.008006949
```

```
cat("Difference between Sample and Pop Mean:",
    abs(mean(sim) - mean(sim_sample)), "\n")
```

```
## Difference between Sample and Pop Mean: 0.01459384
```

## Elevating the Statistics Course

## Conclusion



# Conclusion

# Acknowledgements