



# Picking Intro

School of Information Studies  
Syracuse University

# Patterns in Games and Prices

- Obtain
- Scrub
- Explore
- Model
- iNterpret

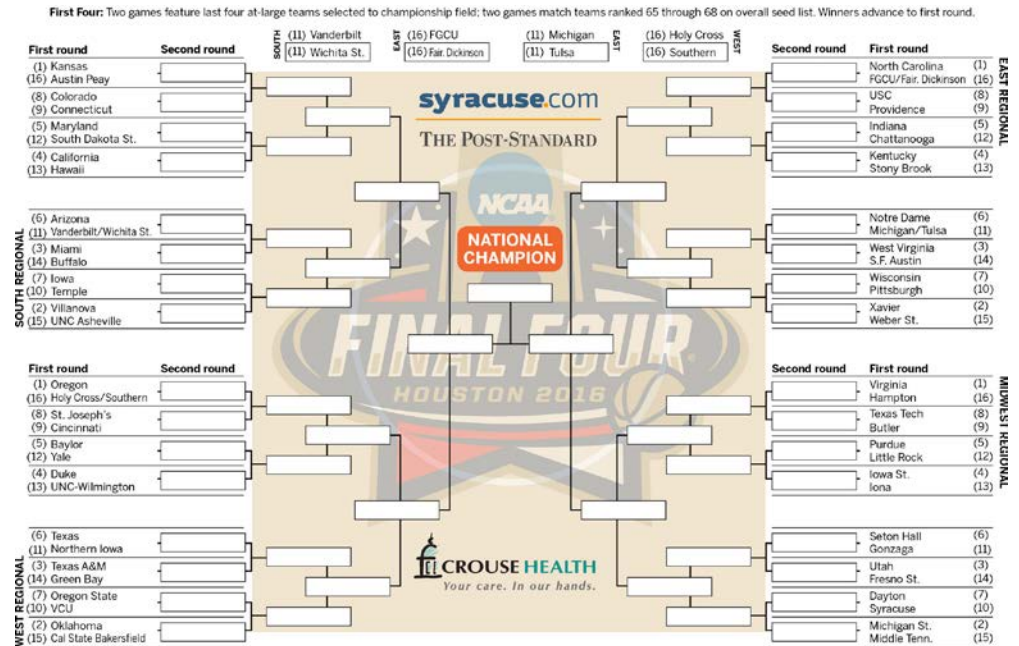




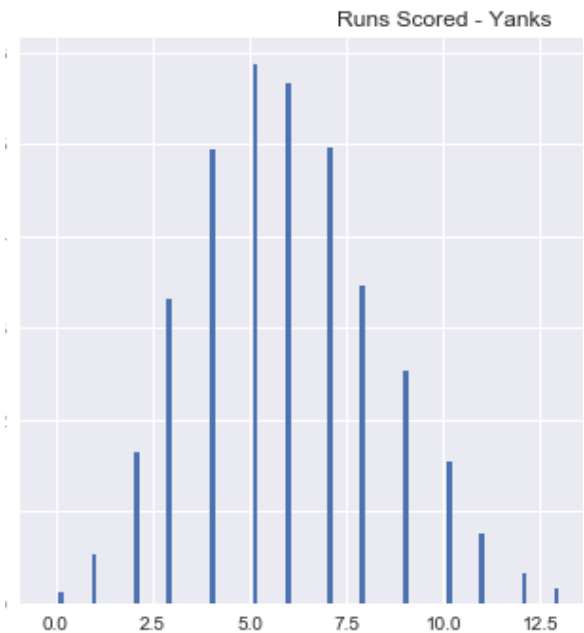
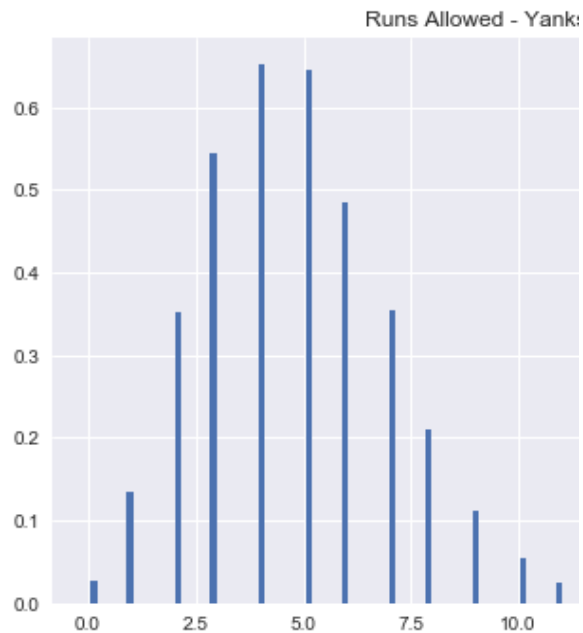
# Our Challenge This Week?



# Using Distributions to Pick a Winner



# But How?



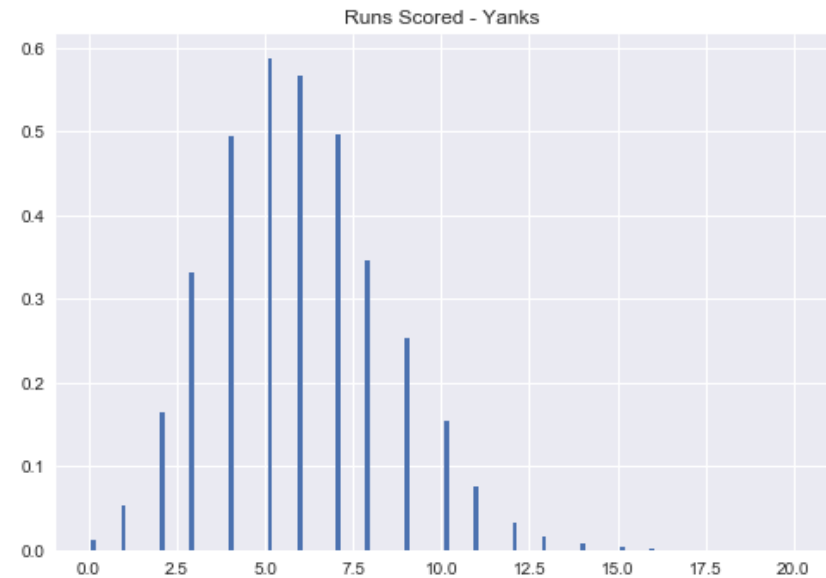
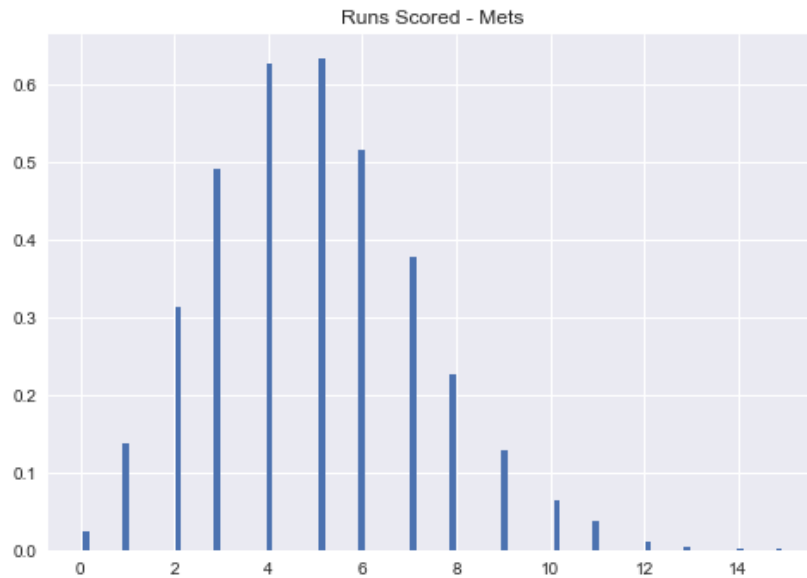




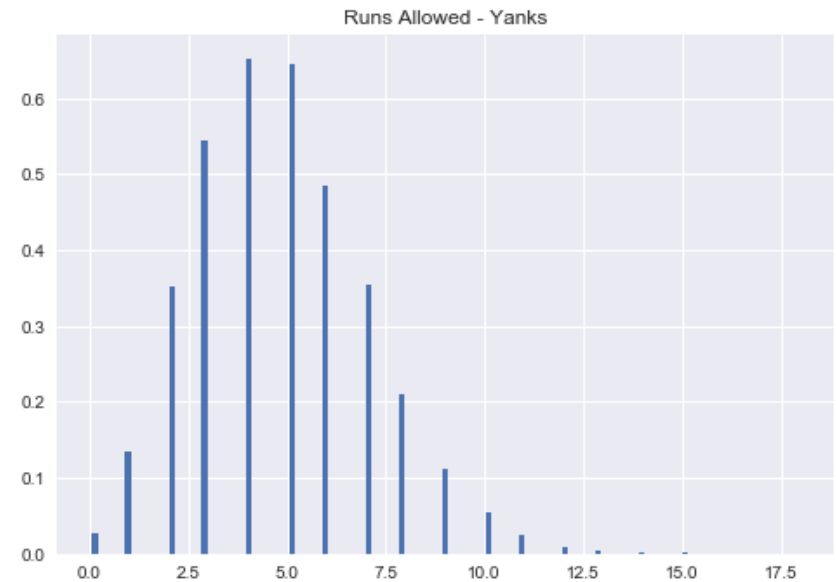
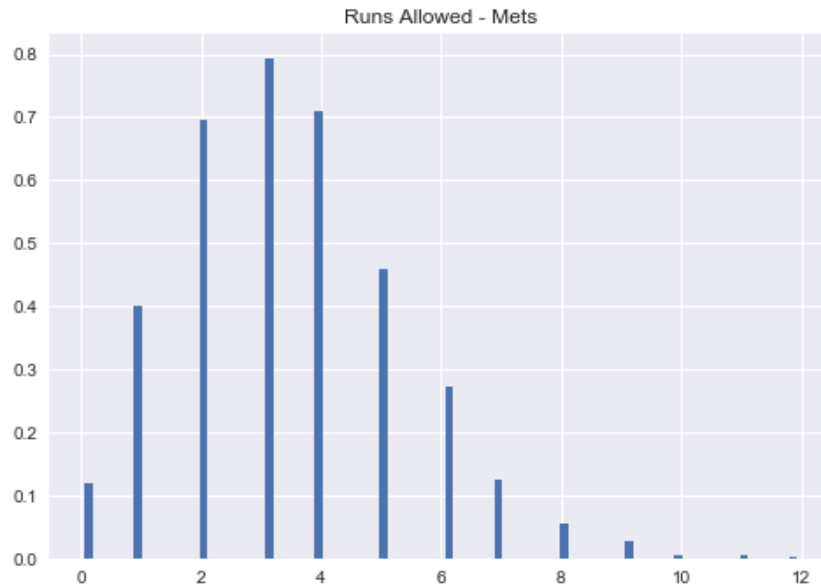
# Data Review

School of Information Studies  
Syracuse University

# Runs Scored



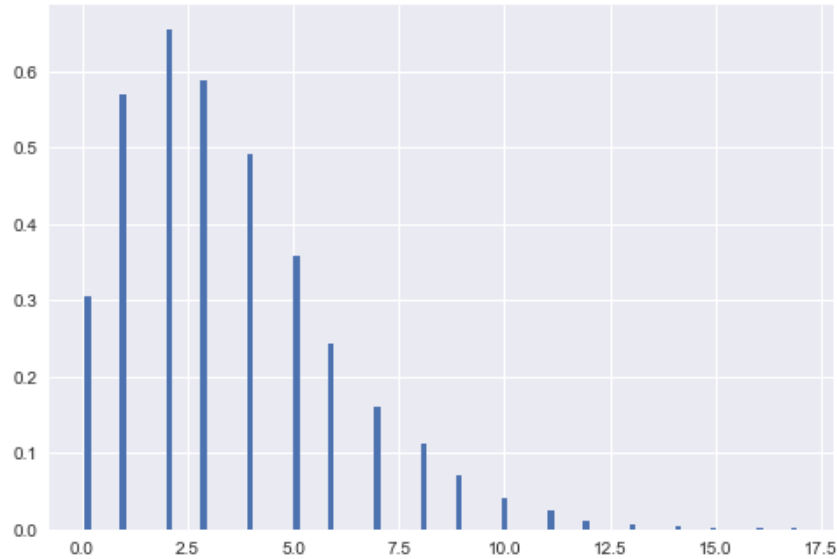
# Runs Allowed



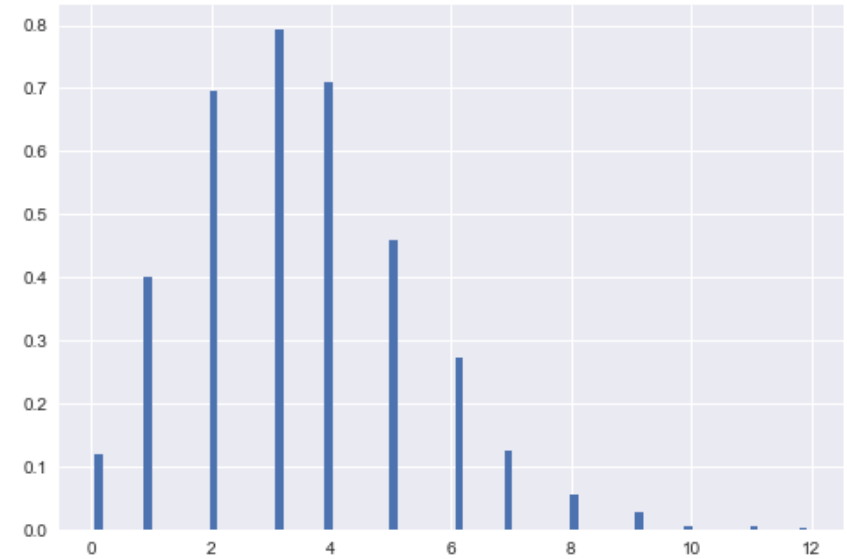


# Scoring Distributions

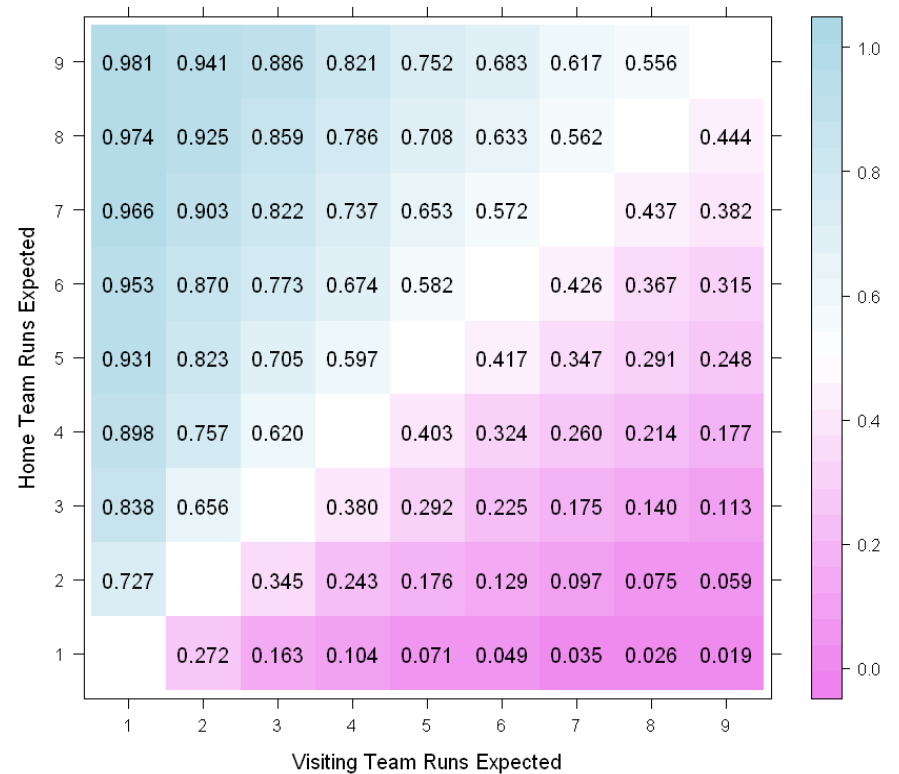
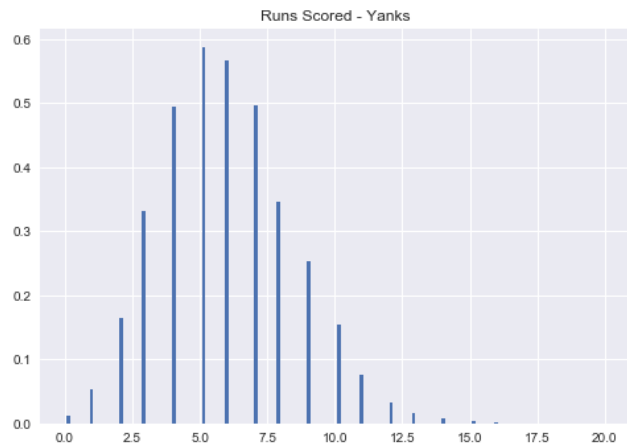
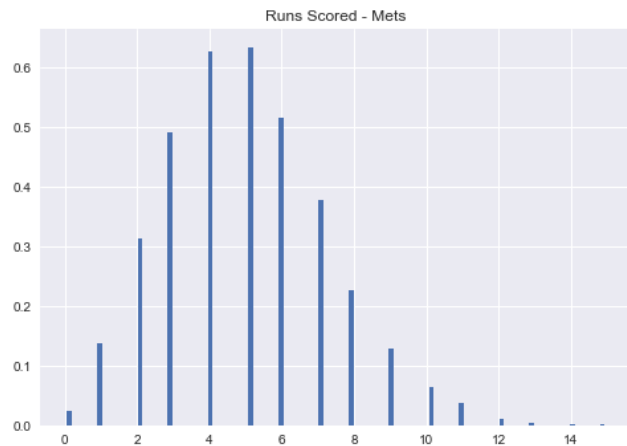
Runs Allowed - Mets



Runs Allowed - Yanks



# Picking Winners



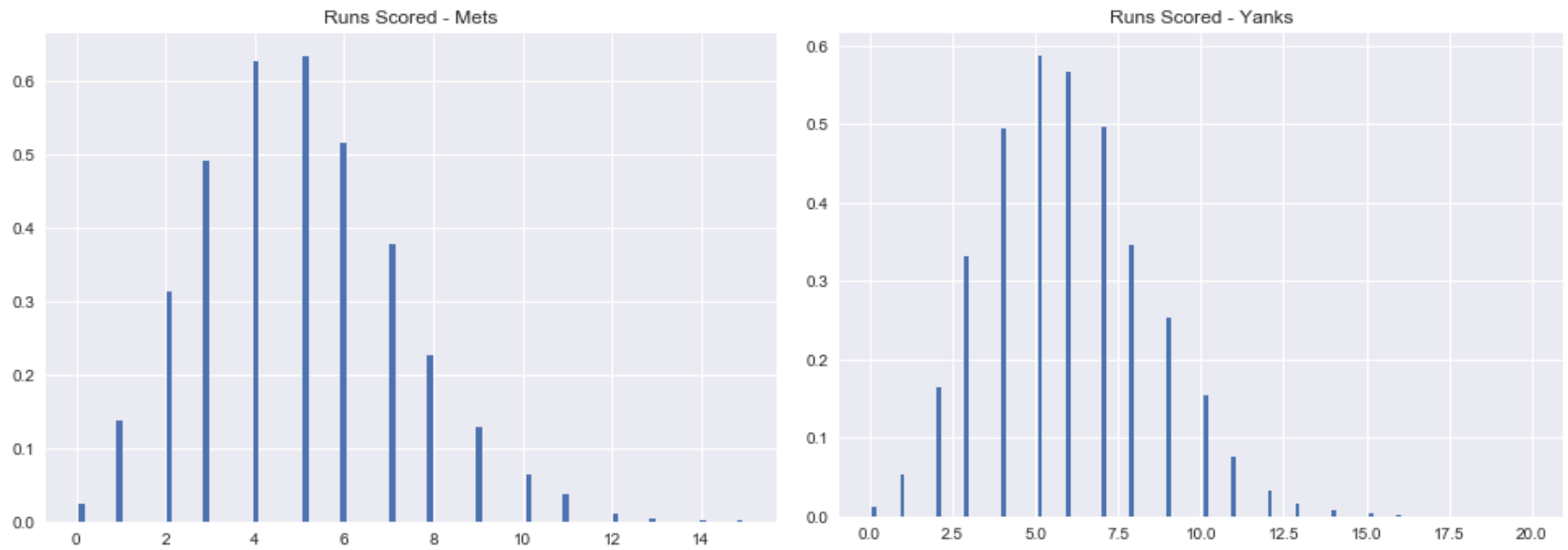


# Poisson Distribution

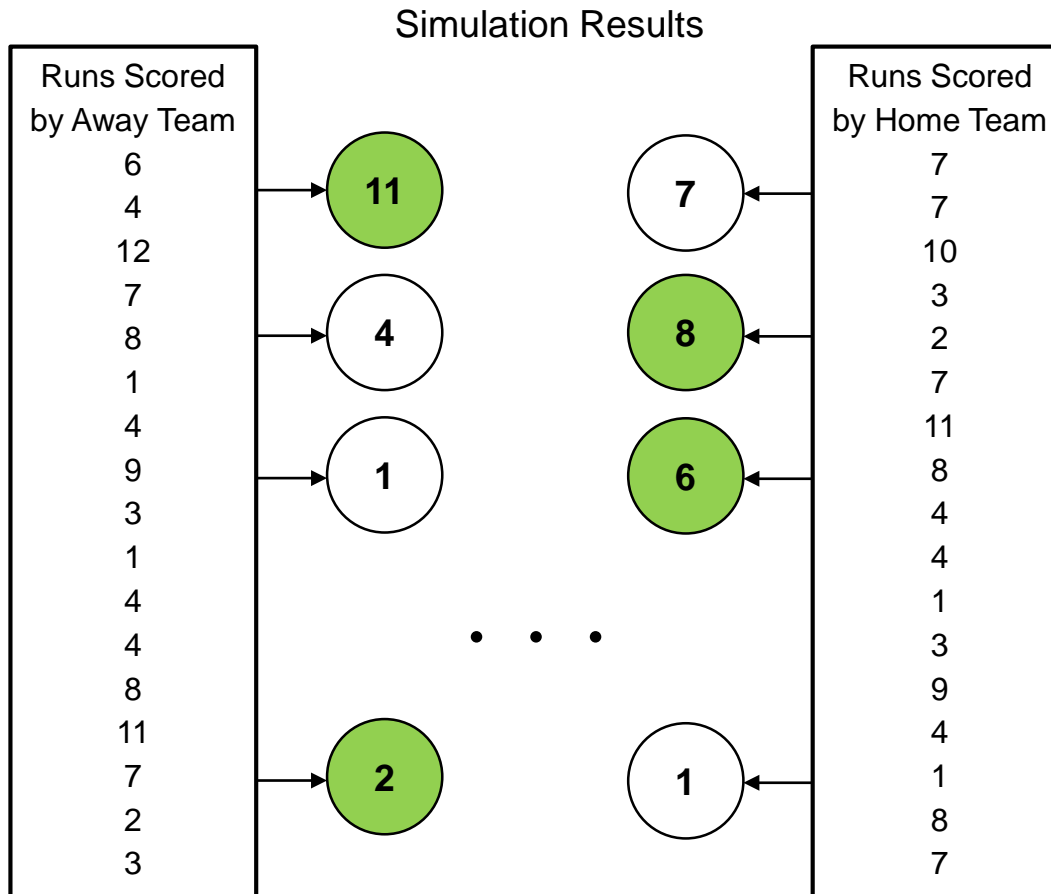
School of Information Studies  
Syracuse University



# Simulating Runs Scored



# Sports Simulation



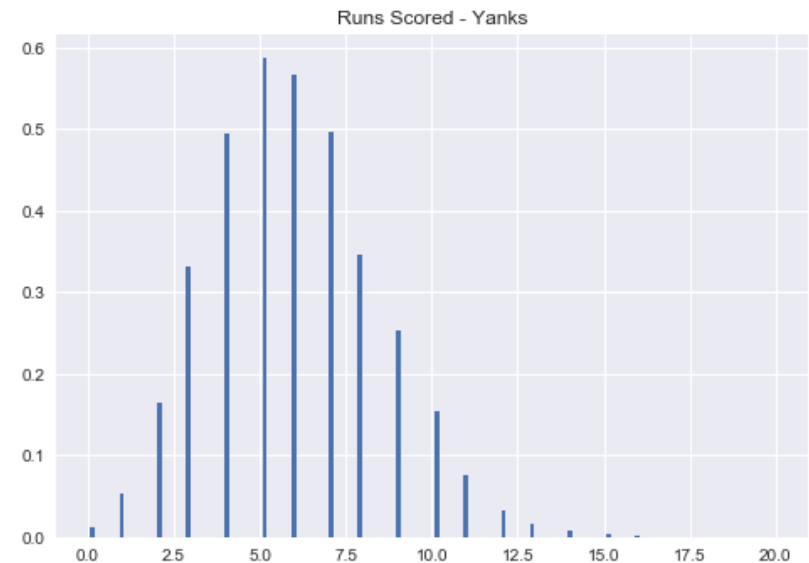
Source: Adapted from Miller (2005).

# Poisson Distribution

- Good approximation for count responses

$$P(Y = y) = \frac{e^{-\mu} \mu^y}{y!}$$

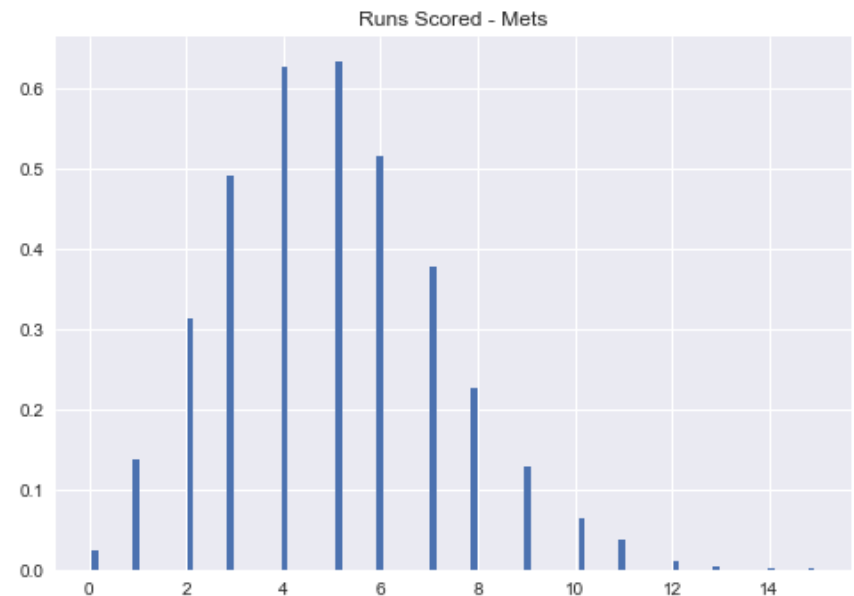
- Occurrence of events during certain time interval
- Arrival rate problems





# Poisson Distribution (cont.)

```
MetAwayScore =  
np.random.poisson(4.97, 10000)  
  
MetAwayDefend =  
np.random.poisson(3.45, 10000)  
  
YankHomeScore =  
np.random.poisson(5.97, 10000)  
  
YankHomeDefend =  
np.random.poisson(4.84, 10000)  
  
plt.hist(MetAwayScore, bins='auto',  
rwidth = .5, normed=True)  
  
plt.title("Runs Scored – Mets")  
plt.show()
```

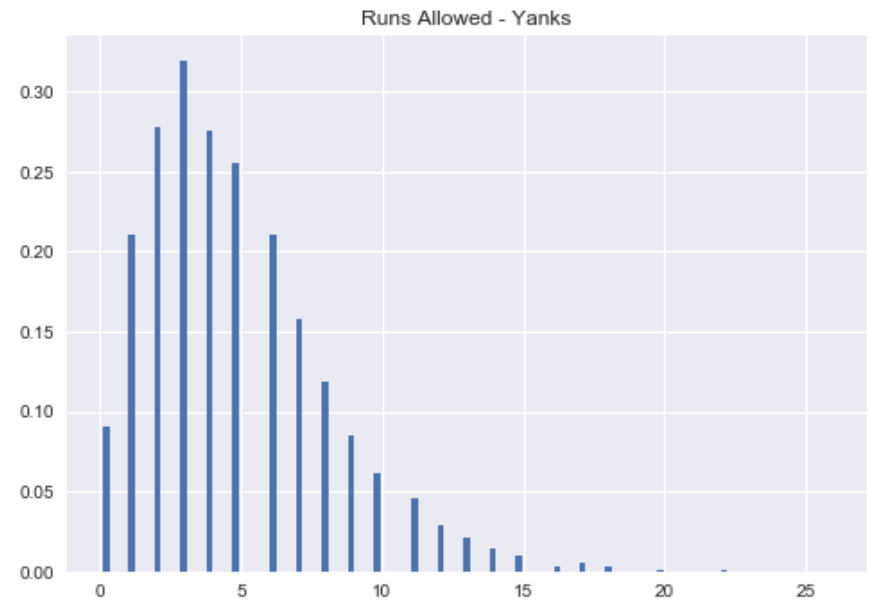
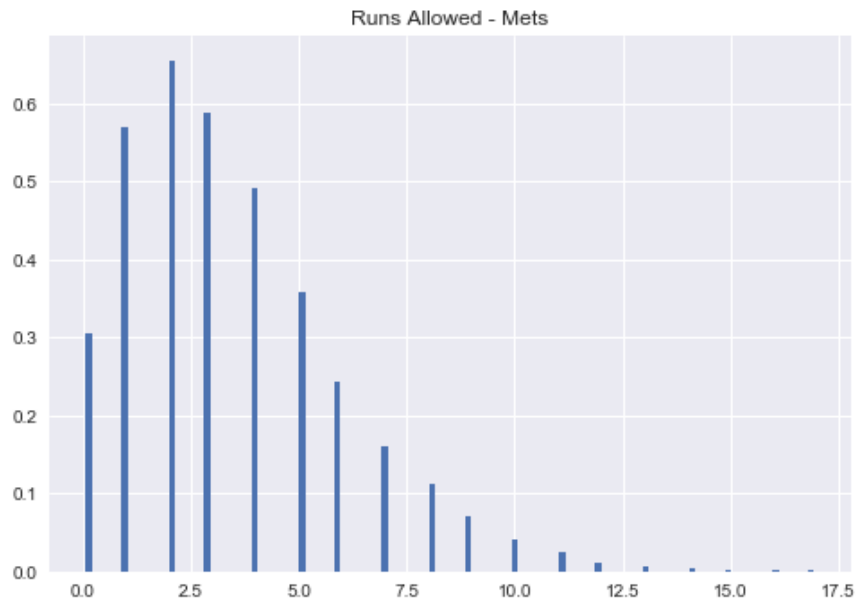




# Negative Binomial Distribution

School of Information Studies  
Syracuse University

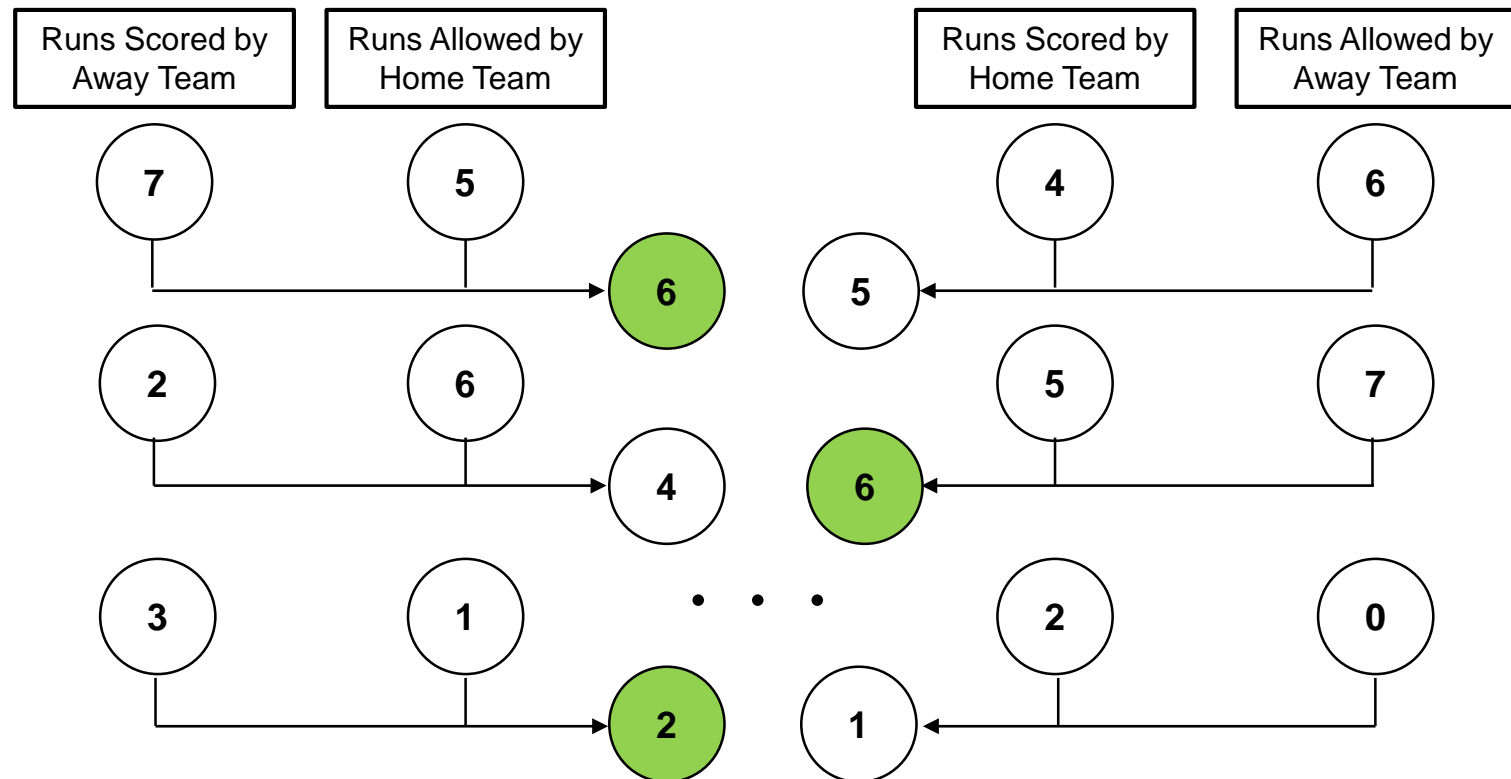
# Simulating Runs Allowed





# Sports Simulation

## Simulation Results



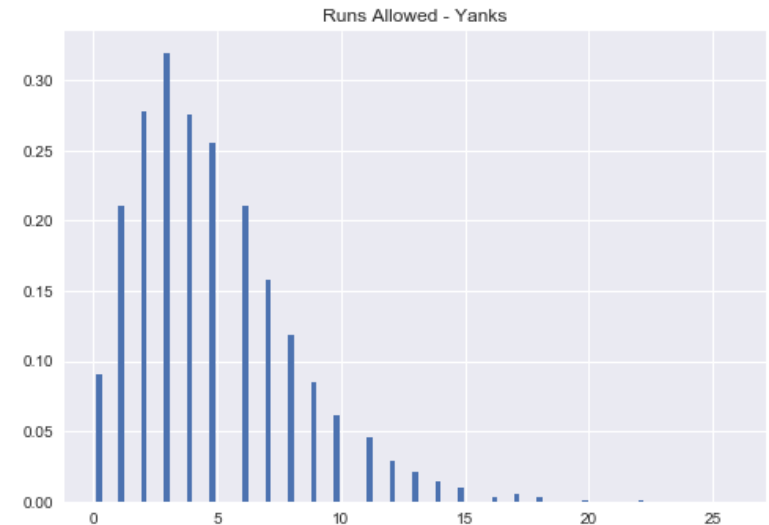
Source: Adapted from Miller (2005).

# Negative Binomial

- Alternative approximation for count responses

$$P(Z = z) = \binom{z-1}{k-1} p^k (1-p)^{z-k}$$

- Generalization of Poisson distribution
- Rare event problems

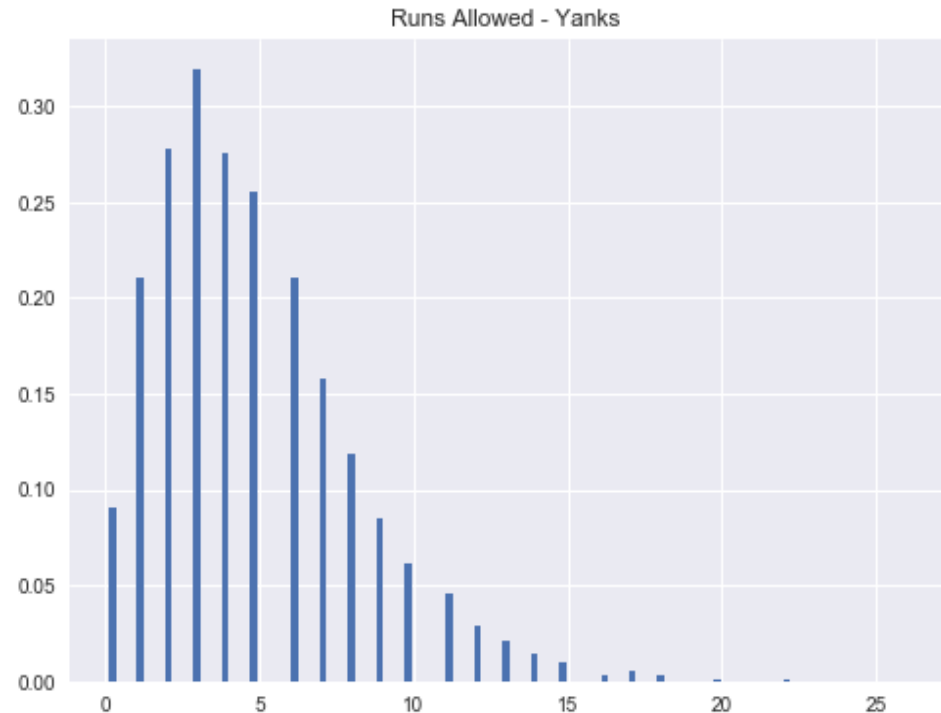


# Negative Binomial

- Alternative approximation for count responses

$$P(Z = z) = \binom{z-1}{k-1} p^k (1-p)^{z-k}$$

- Generalization of Poisson distribution
- Rare event problems



# Negative Binomial (cont.)

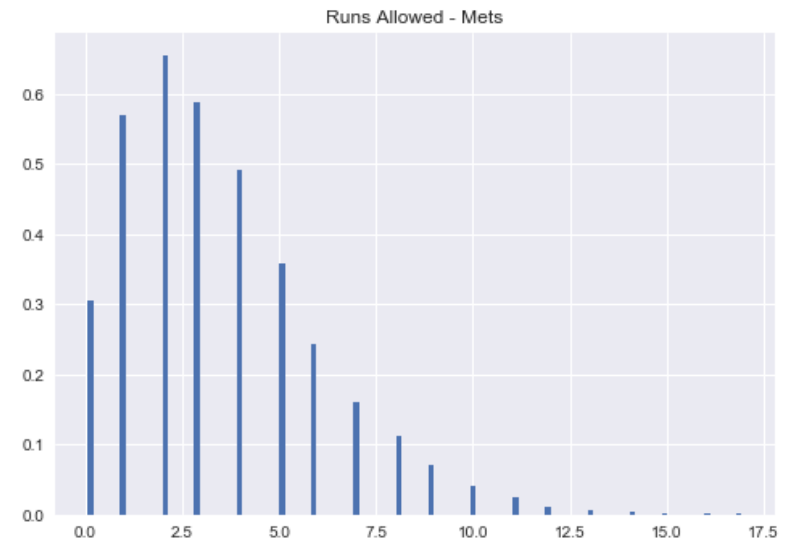
```
MetAwayScore =  
np.random.negative_binomial(4, mas, 10000)
```

```
MetAwayDefend =  
np.random.negative_binomial(4, mad, 10000)
```

```
YankHomeScore =  
np.random.negative_binomial(4, yhs, 10000)
```

```
YankHomeDefend =  
np.random.negative_binomial(4, yhd, 10000)
```

```
plt.hist(MetAwayScore, bins='auto', rwidth = .5,  
normed=True)  
plt.title("Runs Scored – Mets")  
plt.show()
```







# Applications

School of Information Studies  
Syracuse University

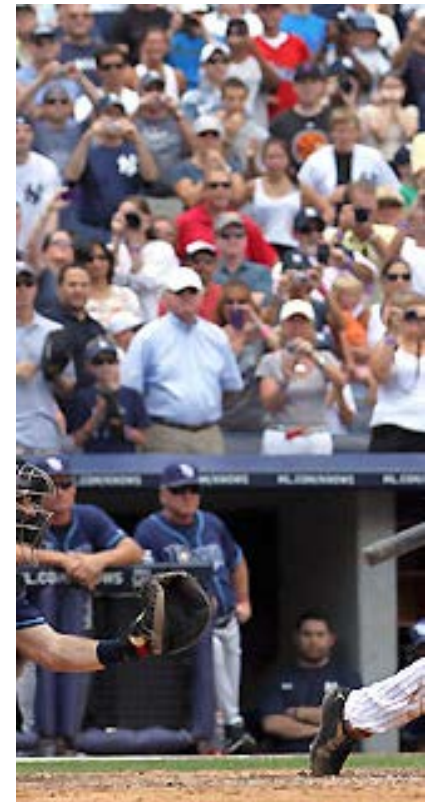


# Additional Applications



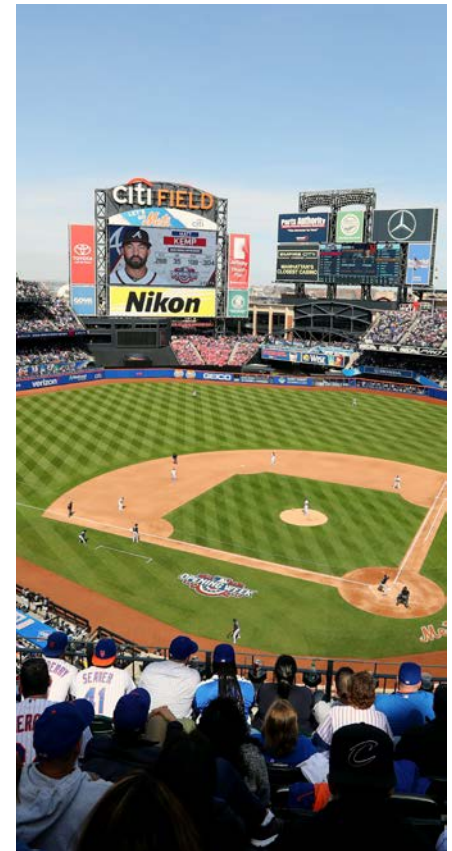
# Baseball Prospectus

- Predicting performance before the season
  - Nate Silver
  - PECOTA
- Variations
  - Military war games
  - Film releases
  - Associate performance



# *Moneyball* Problem

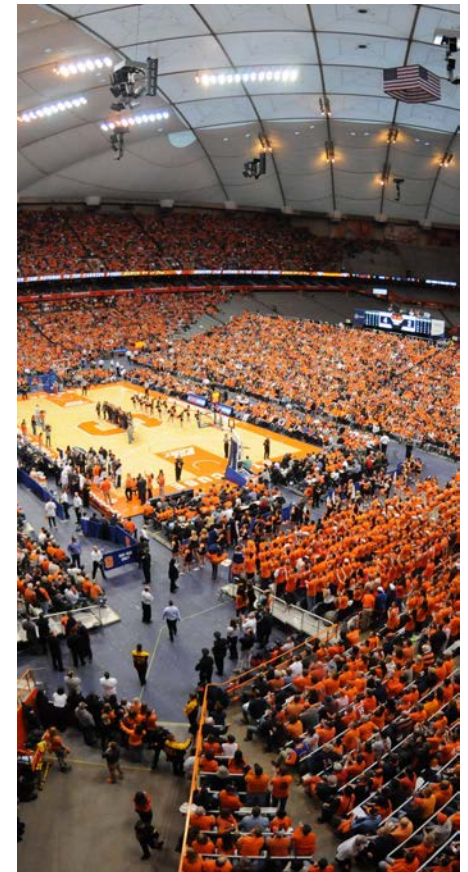
- Calculating the value of a player
  - Billy Beane
  - Individual summary stats applied to team performance
- Variations
  - Associate performance
  - Client conversion
  - Customer lifetime value





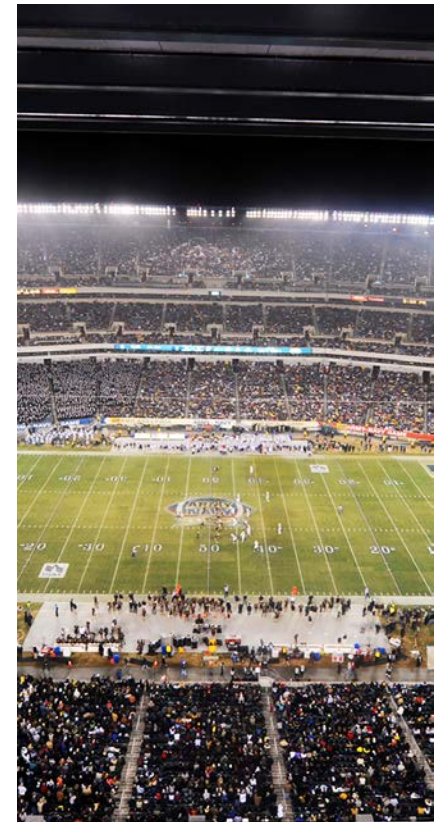
# Coaching Problems

- Utilizing player performance
  - Microanalysis of game
  - Sabermetrics
- Variations
  - Football
  - Basketball



# Bowl Championship Series

- Predicting team performance against unknown opponent
  - Strength of schedule
  - Ensemble approaches
- Variations
  - March Madness
  - Product deployment
  - Recommendation engine



# Billy Waters Problem

- Predicting the winning team in the next game?
  - Human expertise
  - Simulation
- Variations
  - March Madness
  - Film release
  - Product deployment







# Picking II Intro

School of Information Studies  
Syracuse University



# Patterns in Games and Prices

- Obtain
- Scrub
- Explore
- Model
- iNterpret



# Our Challenge Now?

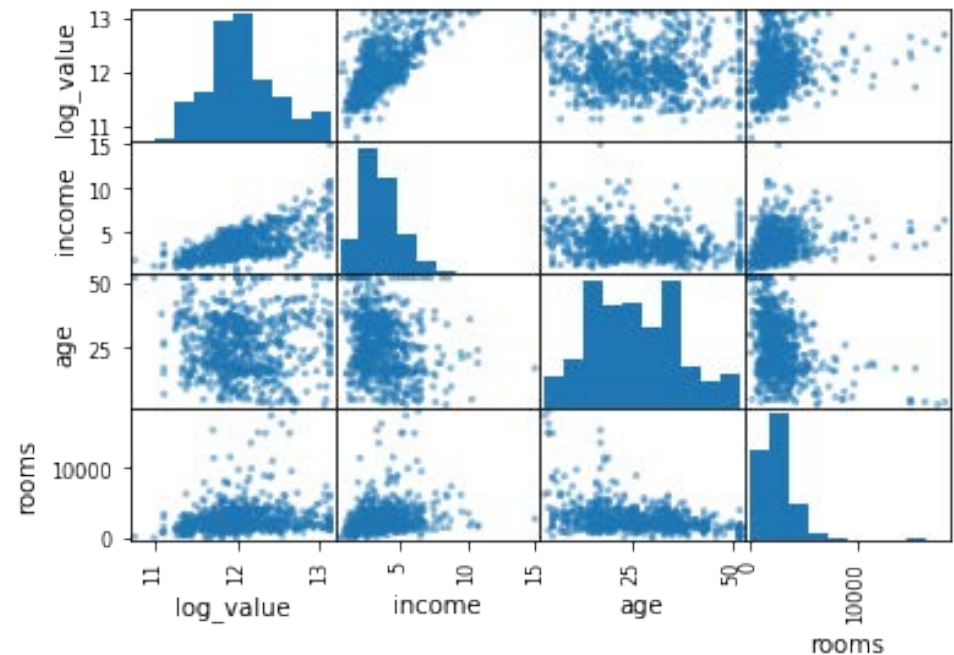
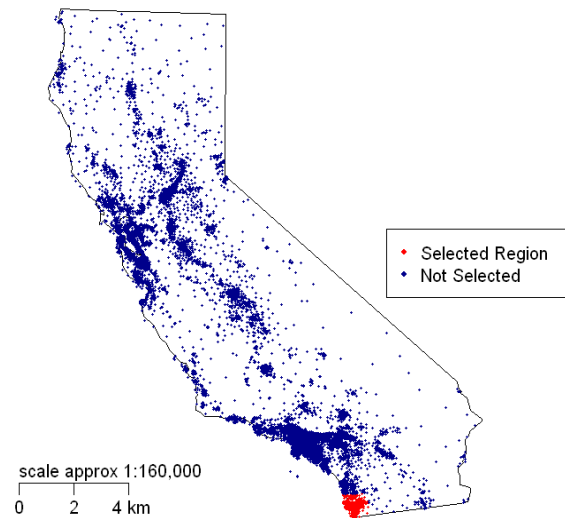


# Modeling Housing Prices





# But How?



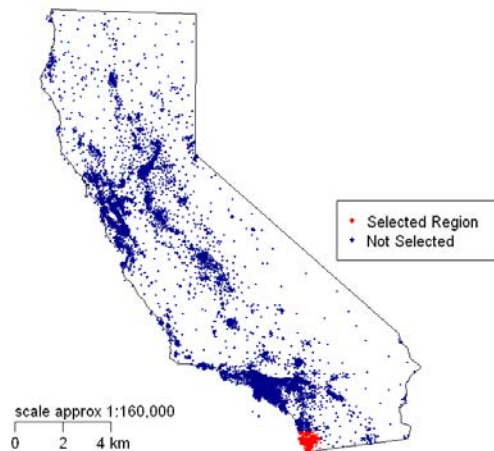




# Data Review II

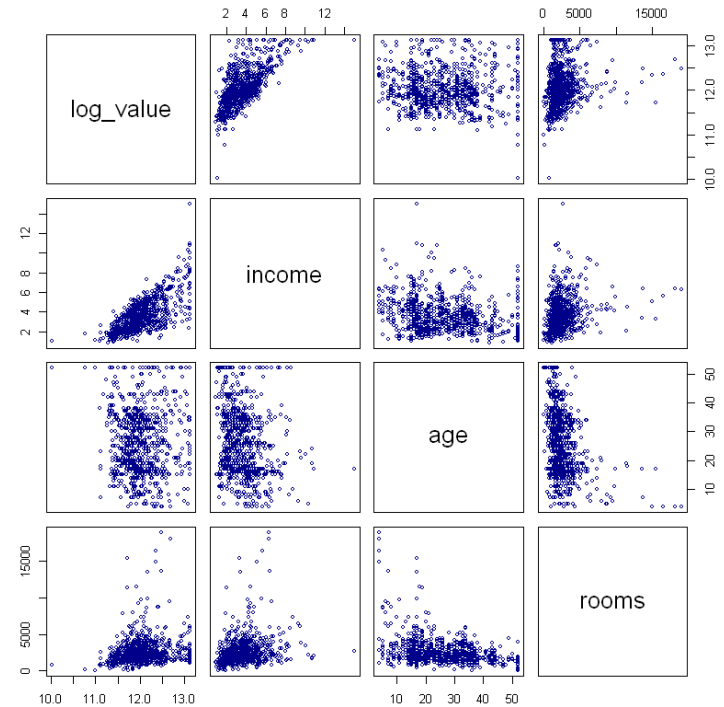
School of Information Studies  
Syracuse University

# Housing Data

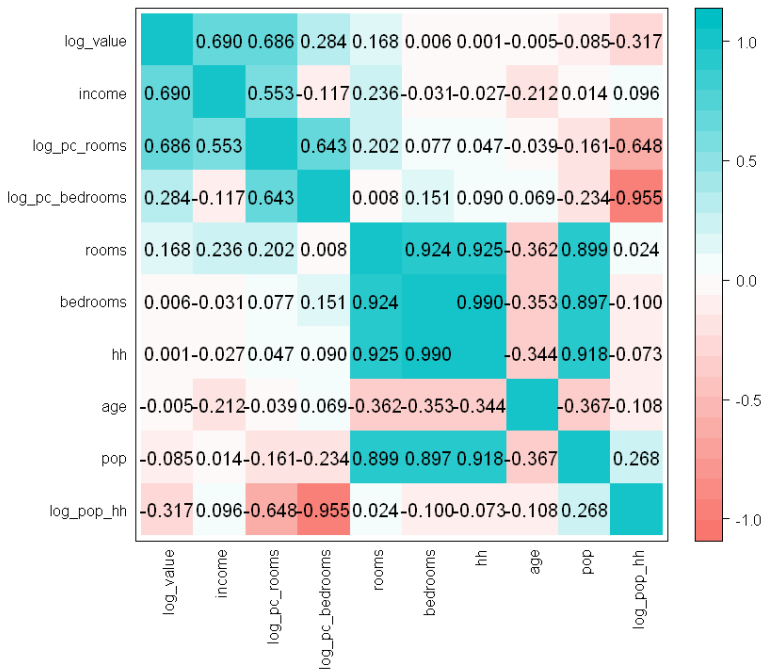




# Feature Correlation

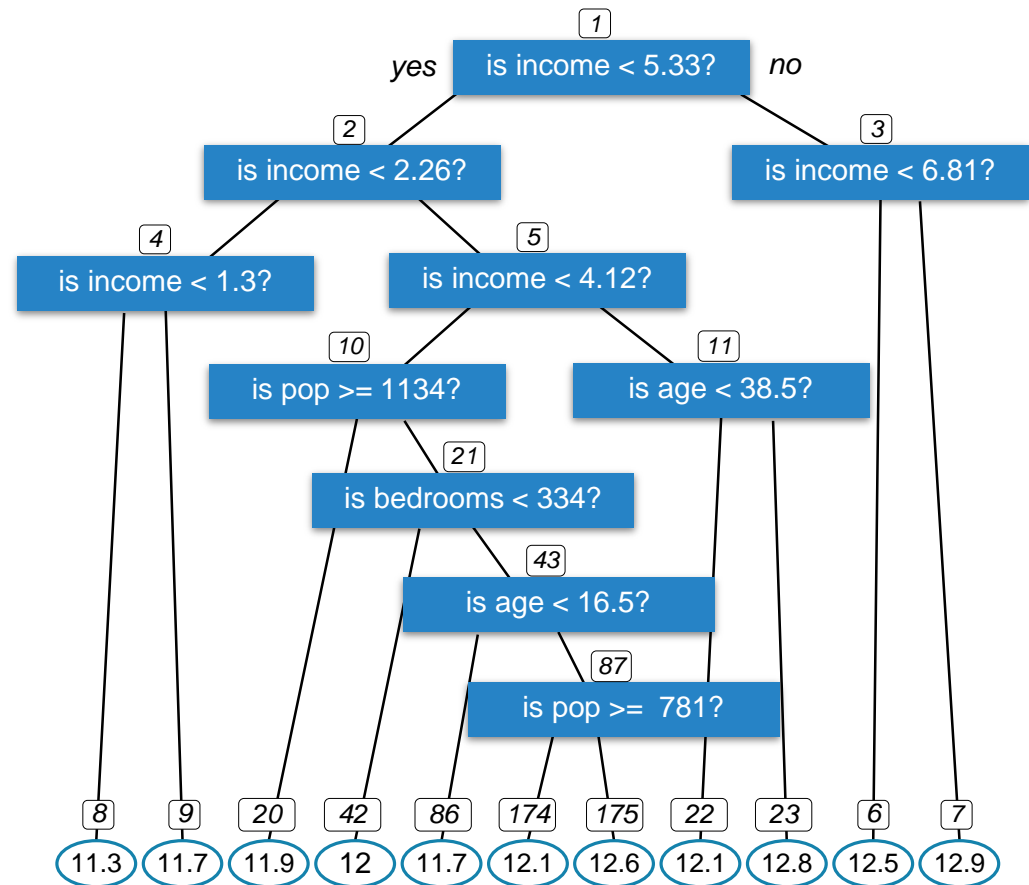


# Feature Correlation (cont.)





# Picking Values



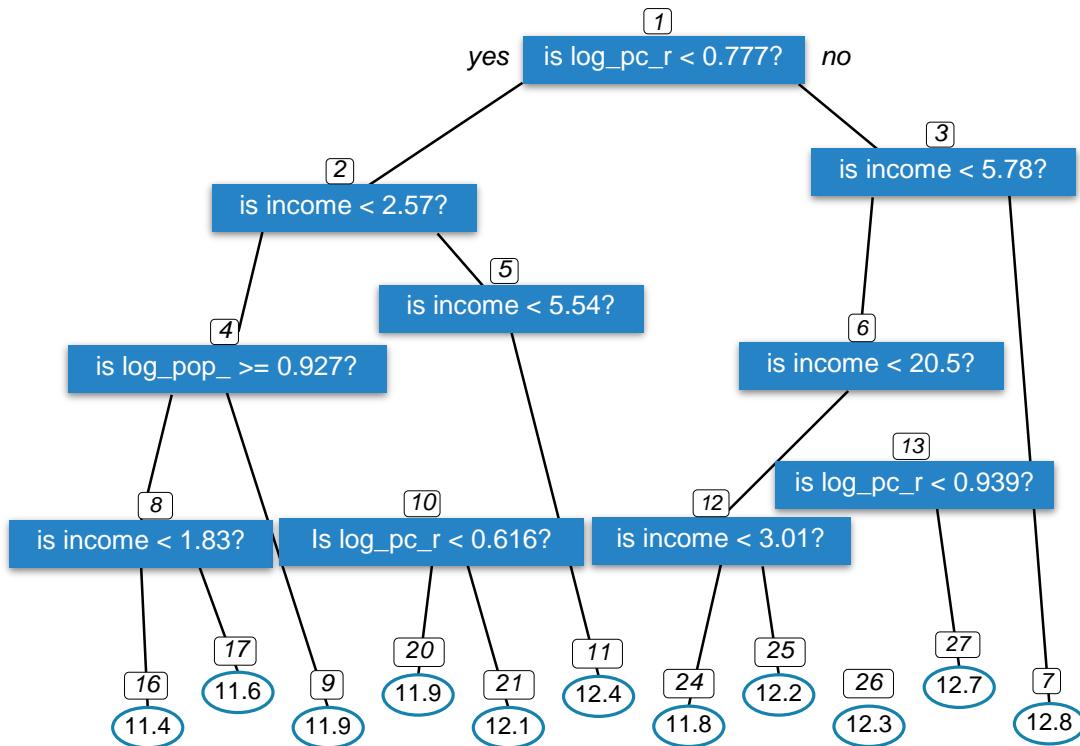


# Trees Forests

School of Information Studies  
Syracuse University

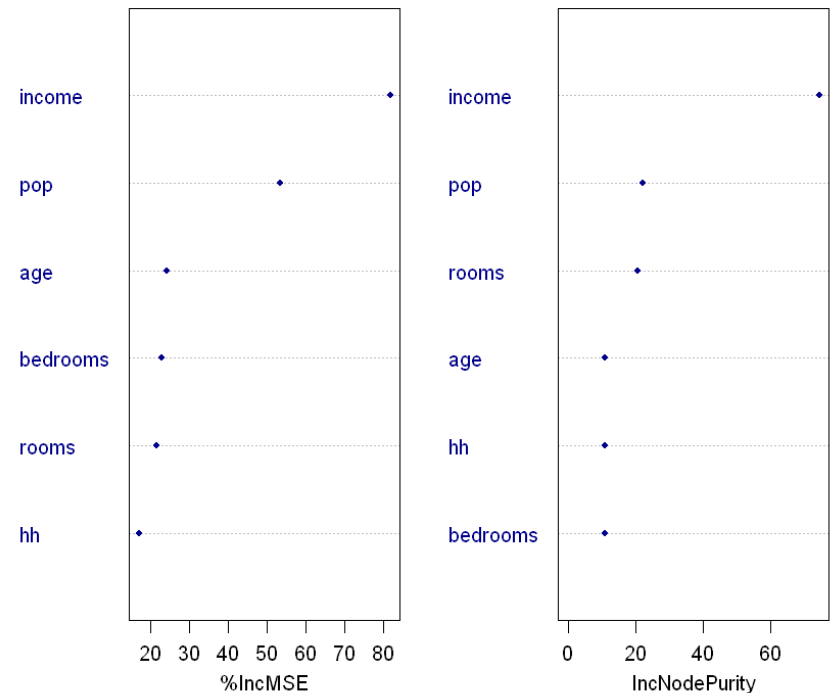


# Picking a Tree in the Forest



# Decision Trees

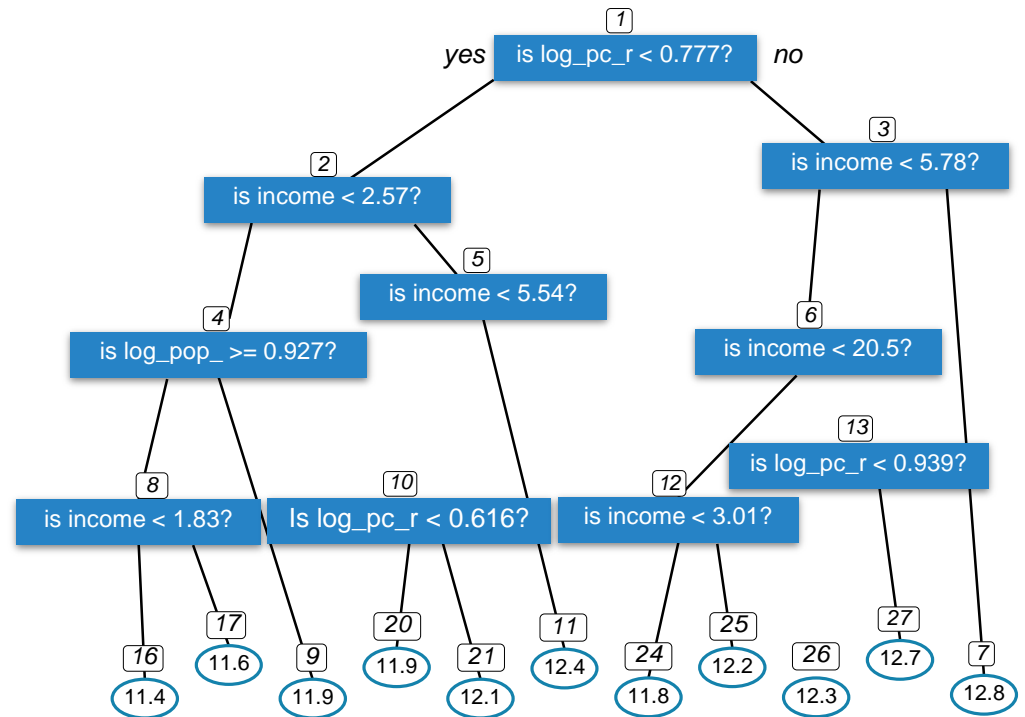
- Key advantage is interpretability
- Partition the space into simple regions to achieve best fit
- Pruning methods control the size of the tree





# Random Forests

- Ensemble method using multiple decision trees
- Recursive partitioning on the training set
- Effective with large number of explanatory variables



# Random Forests (cont.)

- Provides interpretability through use of one tree from set
- Significant difference in performance between train and test indicates overfitting
- Individual explanatory variables can still be inferred

