## Scribe Report - April 28, 2014

## 1. BaseballBets.CSV

- First we plot the Home Team Victory Win (1 or 0) and the point spread
  - Noted that the better odds with the greater point spread
- Second we will place the data into buckets
  - Put the victory into buckets based on the point spread
- Review: Fitted values
  - $\widehat{y}_i = B_0 + B_1 x_i$
  - Where  $\hat{y}_i$  fitted value is the conditional expected value
- Expected value for binary outcome is the weighted average of the expected probabilities
  - For a binary outcome

$x_i$	$w_i$
0: Loss	Nothing
1: Win	$B_0$
	$+ B_1 x_i$

- $E(Y|X) = B_0 + B_1x_i + O(-)$ 
  - $= B_0 + B_1 x_i$
- Conditional expected value of a binary outcome is exactly the probability of getting a 1
- Fit the model with the with the binary outcome for victory we see that the home team wins 52% of the time with a 2% increase with every point spread increase
- o Fit the simple linear model
- o Issues:
  - Probabilities are against the rules of mathematics
  - Problem model on the right side can take any real number but it should only be between 0 and 1 – probabilities are between 0 and 1
    - $Pr(y_i = 1 | x_i)$  should be on (0,1)
    - But  $B_0 + B_1 x_i$  can only be a real number
  - We need our regression function to have a "speed limit" so it does not go past 0 and 1
  - Make an S shaped curve so the probabilities are not outside 0 and 1
- Logistic Regression

- $\Pr(y_i = 1 \mid x_i) = \frac{e^{(B_0 + B_1 x_i)}}{1 + e^{(B_0 + B_1 x_i)}}$   $= g(B_0 + B_1 x_i) \leftarrow \text{speed limit function or "link function"}$
- where g(s) =  $\frac{e^{(s)}}{1+e^{(s)}}$
- Use the link function so that way you cannot go above 1 or below 0

## o In R:

- glm ← generalized linear function
- family ← binomial
- When imposed using the link function you get an S shaped curve

# A simple fix: fit a logistic regression model instead glm1 = glm(homewin~spread, data=bballbets, family=binomial)