Predicting Outcomes of games

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Problem Overview

Sports books present often present 3 main ways to bet on the outcomes of games. Take this example from sportsbook.draftkings.com:

TOMORROW	POINTISPREAD	TOTAL POINTS	MONEYLINE
7:209M MIA Dolphins		O 48 -110	+138
JAX Jaguars		U 48 -110	-157

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We focus here on the moneyline and the point spread.

The money line corresponds to odds which corresponds to winning probabilities. We can check these winning probabilities, estimate our own, and wager accordingly.

According to the moneyline $\it L = MIA + 138$ suggests MIA has a win probability of

$$\mathbb{P}(\mathsf{MIA}) = \frac{100}{100 + L} = \frac{100}{100 + 138} \approx 0.4202$$

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Wait a minute!

$$\mathbb{P}(\mathsf{JAX}) + \mathbb{P}(\mathsf{MIA}) \approx 1.0311 > 1$$

We can normalize these probabilities to get what the book considers as fair probabilities without vigorish:

$$\begin{split} \mathbb{P}(\mathsf{MIA}) &= \frac{\mathbb{P}_{\mathsf{vig}}(\mathsf{MIA})}{\mathbb{P}_{\mathsf{vig}}(\mathsf{MIA}) + \mathbb{P}_{\mathsf{vig}}(\mathsf{JAX})} \approx \frac{0.4202}{1.0311} = 0.4075. \\ \mathbb{P}(\mathsf{JAX}) &= \frac{\mathbb{P}_{\mathsf{vig}}(\mathsf{JAX})}{\mathbb{P}_{\mathsf{vig}}(\mathsf{MIA}) + \mathbb{P}_{\mathsf{vig}}(\mathsf{JAX})} \approx \frac{0.6109}{1.0311} = 0.5925. \end{split}$$

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According to the book's probabilities, either way is a losing bet. In fact, the loss for each bet might be the same if not for a rounding error. The probabilities estimated by the book might not be accurate, since they are set for the purposes of avoiding risk. If more accurate win probabilities can be estimated, then we can exploit market inefficiencies.

Vigourish

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The extra probability is called overround and lets us compute the vigourish.

$$V=\frac{O}{1+O}.$$

In our case,

$$V = \frac{0.0311}{1.0311} = 0.0302.$$

This means the sports book expects to make \$3.02 for every \$100 wagered on this particular money line.

We will now consider a couple of simple models to determine our own win probabilities.

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- I Compute the average points scored and average points allowed for each team.
- 2 Compute the standard deviation for points scored by each team.
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- 4 Repeat 10,000 times.

Monte Carlo Results

In Excel, we found $\mathbb{P}(\mathsf{MIA}) = 0.2701$ and $\mathbb{P}(\mathsf{JAX}) = 0.7299$. If we bet \$100 on MIA, we get \$138 for winning, so according to our model, our expected winnings are

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$$\mathbb{E}[W_{\mathsf{JAX}}] = 0.7299 \cdot \$63.69 + 0.2701 \cdot (-\$100) = \$19.48$$

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- Might want to consider data from games that do not involve the two teams.
- Simulations would be easier in R or Python.

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Caution: We cannot estimate individual team effects using Im() since the design matrix is not full column rank.

Linear Regression Results

From our regression model we found that

$$\mathbb{E}[Y] = \hat{\beta}_{\mathsf{JAX}} - \hat{\beta}_{\mathsf{MIA}} + \hat{\beta}_1 = 5.3749.$$

$$\mathbb{P}(MIA) = 0.3293 \text{ and } \mathbb{P}(JAX) = 0.6706.$$

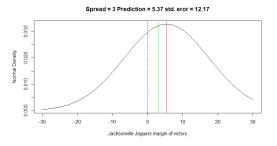


Figure: Probability density for JAX margin of victory according to the regression model.

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$$\mathbb{E}[W_{\mathsf{JAX}}] = 0.6706 \cdot \$63.69 + 3293 \cdot (-\$100) = \$9.78$$

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- Estimates for differences in team effects will improve as we see more games.
- Might consider a combination of the Monte Carlo and Regression Model.
- The regression model should also give some insight on the point spread.

We already determined from the regression model that $\mathbb{E}[W_{\text{JAX}}] = \9.78 if we bet on the moneyline. What if we bet on the spread?

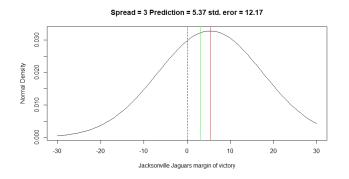


Figure: Probability density for JAX margin of victory.

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We very close to \$9.78, but still less. Bet on the money line.

Questions?

[1] Lloyd Danzig. An Intro to Quantitative Modeling for Sports Bettors (in Excel).

https://medium.com/@lloyddanzig/quantitativesports-betting-6976e1ceaf0f

[2] Lloyd Danzig. Bookmaking Economics: Vigorish & Overround. https://medium.com/analytics-vidhya/bookmaking-economics-8710d25a42a5

[3] Hal Stern. On the Probability of Winning a Football Game. *The American Statistician*. (1991): 179–183.