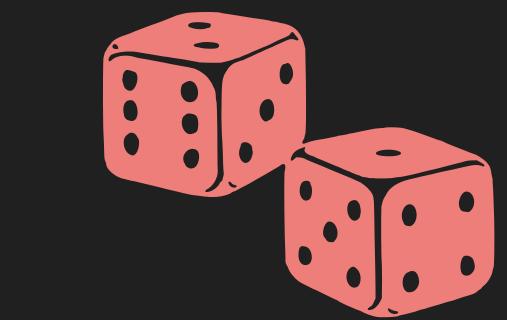


March Madness and Gambling Mac McLean, Santhosh Rajendran & Ariana Arenson



BACKGROUND

- March Madness Tournament is an exciting time for athletes, fans, and sports betters
- Approximately \$9 billion wagered during 2021 March Madness
- Over/Under betting is popular form of betting predicting the total number of combined points based on a line

Our Mission: To see if there is a way to gain an edge in predicting the total number of points in a game through multiple linear regression of team's statistics

Research Questions:

- 1.Do higher season long-term statistics lead to a larger total number of points in a March Madness game?
- 2. Do shooting specific statistics (such as 3P%, FT%, etc.) correlate with the total amount of points in a March Madness game?
- 3.Do non-team statistics, such as seed, round, or region correlate with the total amount of points in a March Madness game?

Our Data:

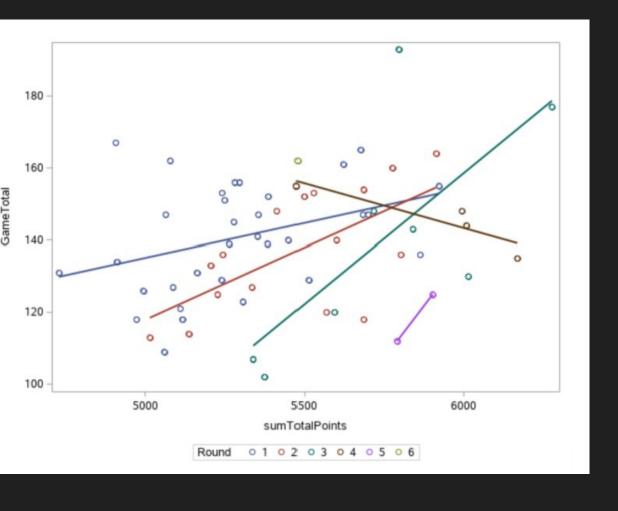
- collected from sportsreference.com, which is the most reliable source for all kinds of sports statistics data
- combined the team stats with a historical dataset from dataworld.com that has the results of all of the games from the 2019 tournament

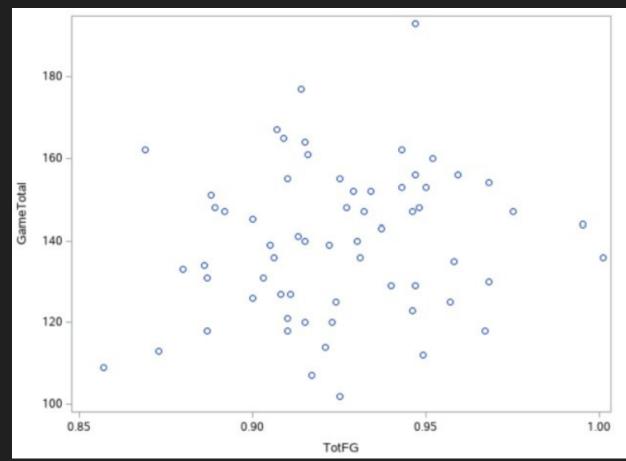
REGRESSION ASSUMPTIONS

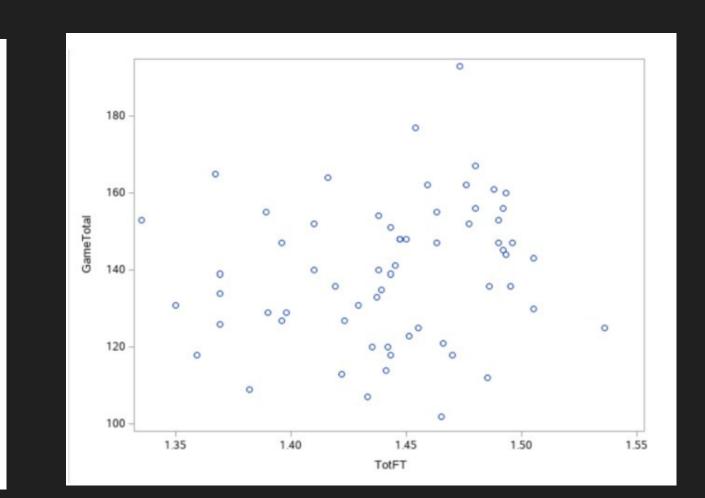
• focus on the sum of team stats for each game

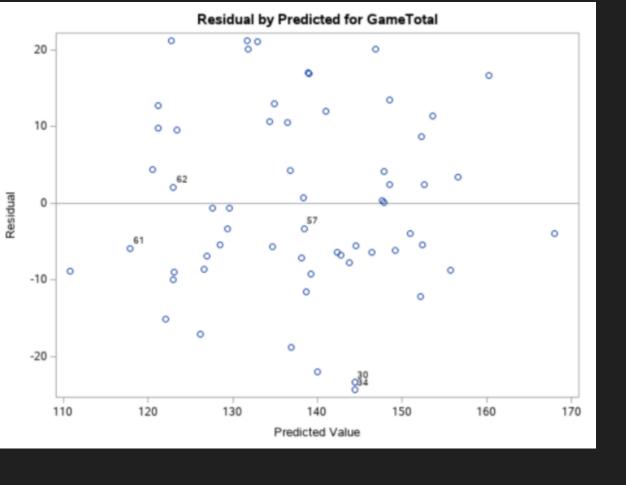
Variables: Round, Region Name, Seed1, Seed2, GameTotal, sumTotalPoints, sumPointsAllowed, fgPer, threeptPer, ftPer, sumRebounds, sumAssists

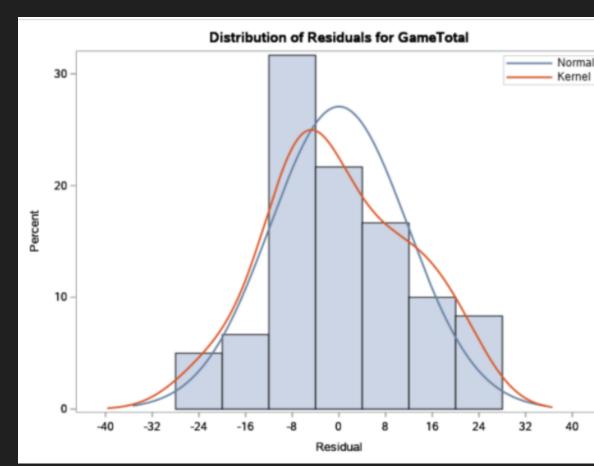
EXPLORATORY DATA ANALYSIS

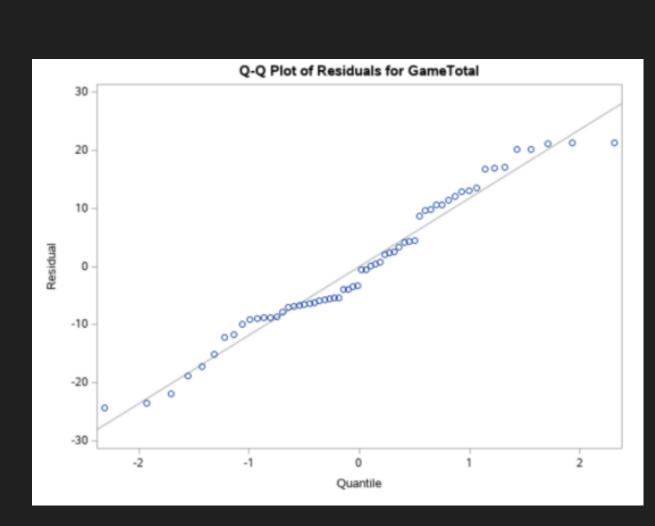












STEP 1: Add Quantitative Predictors

Initial: GameTotal = $\beta_0 + \beta_1$ fgPer+ β_2 ftPer+ β_3 sumTotalPoints+ β_4 threePtPer + β_5 sumAssists + β_6 sumRebounds + β_7 sumPointsAllowed + β_8 sumSeed

Final: GameTotal = $\beta_0 + \beta_1$ threePtPer+ β_2 sumRebounds+ β_3 sumPointsAllowed

FINAL MODEL FOR STEP 1

Adj. R-Squared: .4361 P-value: <.0001

Root MSE: 12.655

STEP 2: Add Qualitative Predictors

Initial: GameTotal = $\beta_0 + \beta_1$ threePtPer+ β_2 sumRebounds+ β_3 sumPointsAllowed+ β_4 Round + β_5 East + β_6 South + β_7 West + β_8 Midwest + β_9 Seed1 + β_{10} Seed2

Final: GameTotal = $\beta_0 + \beta_1$ threePtPer+ β_2 sumRebounds+ β_3 sumPointsAllowed+ β_4 Round + β_5 Seed2

FINAL MODEL FOR STEP 2

Adj. R-squared: .4697

P-value: <.0001 Root MSE: 12.157

STEP 3: Add Interactions

$$\label{eq:first-seed-seed} \begin{split} & \text{Initial: GameTotal} = \beta_0 + \beta_1 \\ & \text{threePtPer} + \beta_2 \\ & \text{sumRebounds} + \beta_3 \\ & \text{sumPointsAllowed} + \beta_4 \\ & \text{threePtPer} * \\ & \text{Round} + \beta_5 \\ & \text{threePtPer} * \\ & \text{sumRebounds} * \\ & \text{sumRebounds} * \\ & \text{Seed2} \\ & + \beta_{10} \\ & \text{sumRebounds} * \\ & \text{Round} + \beta_{11} \\ & \text{sumRebounds} * \\ & \text{sumPointsAllowed} * \\ & \text{sumPointsAllo$$

to be Significant

No Interactions Were Found

Final Model

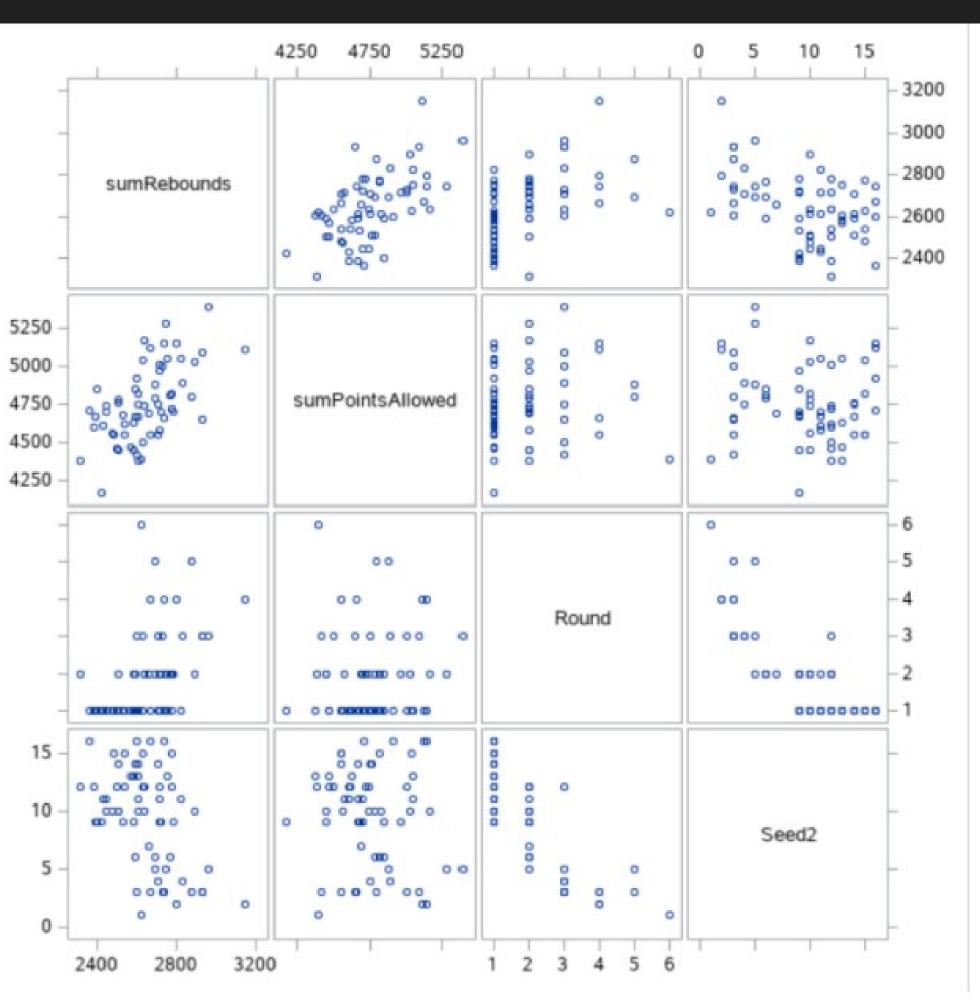
Final Overall: GameTotal = $\beta_0 + \beta_1$ sumRebounds+ β_2 sumPointsAllowed+ β_3 Round + β_4 Seed2

FINAL MODEL STATISTICS

Adj. R-Squared: .4897

P-value: <.0001 Root MSE: 12.207

Multicollinearity



Variable Screening

| Summary of Stepwise Selection | | | | | | | | |
|-------------------------------|---------------------|---------------------|-------------------|---------------------|-------------------|--------|---------|--------|
| Step | Variable Entered | Variable Removed | Number Vars In | Partial R-Square | Model R-Square | C(p) | F Value | Pr > F |
| 1 | sumPointsAllowed | | 1 | 0.2537 | 0.2537 | 8.4901 | 20.74 | <.0001 |
| 2 | sumRebounds | | 2 | 0.0507 | 0.3044 | 5.9048 | 4.37 | 0.0407 |
| 3 | Seed2 | | 3 | 0.0420 | 0.3464 | 4.1104 | 3.79 | 0.0564 |

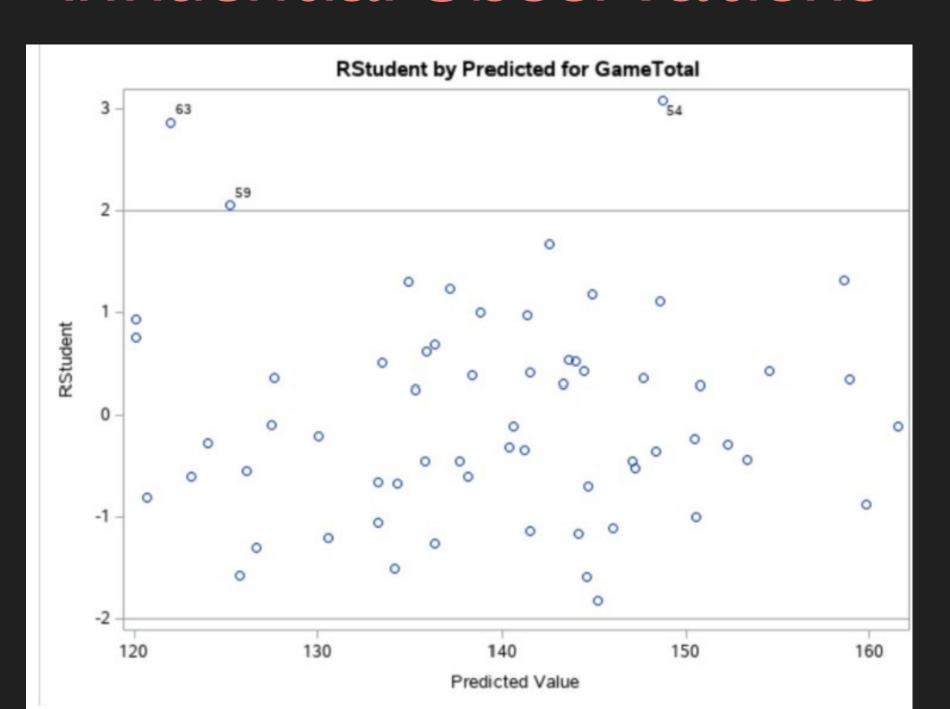
Stepwise Selection

- Resulted in sumPointsAllowed, sumRebounds, and Seed2
 - All significant parameters used in final model
- SLEntry and SLStay values of 0.15

Multicollinearity Check

- VIF numbers are all below 10
- Highest correlation is .57 between rebounds and opponent total.
- Overall multicollinearity shouldn't be an issue.

Influential Observations



54, 59, and 63
are outliers
and were
removed

Conclusions:

Prediction Equation:

 $\widehat{GameTotal} = -14.93 - .022sumRebounds + .05sumPointsAllowed - 8.72Round - 1.62Seed2$

Interpretation:

The Round that the game is the largest parameter and has a lot weight in predicting the total score. The strongest positive parameter was that of the sumPointsAllowed

Usefulness for Predictions:

- the model is significant and somewhat useful for predicting the total number of points in a given

 March Madness game
- the R-squared value is only .51, meaning that only
 51% of the variation in the data can be explained by
 the model

Examples of Model Usage:

UVA vs. Ohio, first round game from 2021 tournament

Over/Under: 130

Actual Total Points: 120

Predicted Total Points: 118.92

Residual: 1.08

Low residual of 1.08 and the model also correctly predicted the under for this game (the application of our model)

Future Use/Limitations:

The scope of the data is only from the 2019 tournament. With more years of data, like from the 2000s for example, the model's accuracy may be improved.

The game constantly changes in terms of strategies, so having this year's data from the 2021 tournament would also likely improve the model.

We could also improve this research by using more advanced statistics, like offensive and defensive efficiency or offensive and defensive ratings.

WORKS CITED

Blinder, Alan. "The N.C.A.A. Tournament's Pandemic Playbook." The New York Times, The New York Times, 15 Mar. 2021, www.nytimes.com/2021/03/15/sports/ncaabasketball/ncaa-covid-march-madness.html.

Chiusano, Anthony. "How Free Throw Percentages Factor into March Madness Success." NCAA.com, NCAA.com, 3 Aug. 2018, www.ncaa.com/news/basketball-men/article/2018-08-02/how-team-free-throw-percentages-factor-march-madness-success.

Steinberg, Leigh. "March Madness' Popularity Off The Charts." Forbes, Forbes Magazine, 17 Mar. 2016, www.forbes.com/sites/leighsteinberg/2016/03/16/march-madness-popularity-off-the-charts/?sh=14a7651b2084.