

MLB Betting Model

Ryan Pedersen

Background Info

- Professional and Amateur
 Sports Protection Act
 wasted in 1992 that made
 sports gambling illegal in
 all states except Nevada
- Federally legalized in 2018
 by the Supreme Court
- Sports betting has grown massively in popularity the past few years.



Legal landscape as of August 11, 2022



Betting Lines Example

			Runline	Win	Total
8/16/22	Philadelphia Phillies Kyle Gibson (R)	40.0	-1.5 (-120)	-180	0 9.0 (-115)
3:40 PM	+ 110 Bets > Cincinnati Reds T.J. Zeuch (R)	+1.5 (EVEN)	+148	U 9.0 (-105)	
8/16/22	San Diego Padres Sean Manaea (L)		-1.5 (+115)	-155	0 7.0 (-125)
3:40 PM	Miami Marlins Edward Cabrera (R)	+ 110 Bets >	+1.5 (-135)	+130	U 7.0 (+105)
8/16/22	Boston Red Sox Nick Pivetta (R)		-1.5 (+120)	-145	0 8.5 (-105)
4:05 PM	Pittsburgh Pirates Mitch Keller (R)	+ 111 Bets >	+1.5 (-140)	+122	U 8.5 (-115)

Process

01

Data Acquisition

How all of the data was gathered

02

Modeling

Type of model used, features

03

Utilization

What to do with the predictions





Data Sources



Fangraphs

Used pybaseball wrapper to pull game logs, batting and pitching stats



MLB.com

Scraped mlb.com probable pitchers page to get daily games and the starting pitchers.



Betting sites

Used the-odds-api to pull odds from various betting sites



Sample of Statistics Used

Stat name	Description	Player type		
ВА	Hits divided by at-bats	Batter		
wRC+	Weighted runs created plus	Batter		
ERA	Earned Runs allowed per 9 innings	Pitcher		
WHIP	Walks plus hits per inning pitched	Pitcher		



Modeling Info

- Model chosen: Linear Regression
- Random Forest also considered
- 21 features used
- Use game logs from Fangraphs to train model



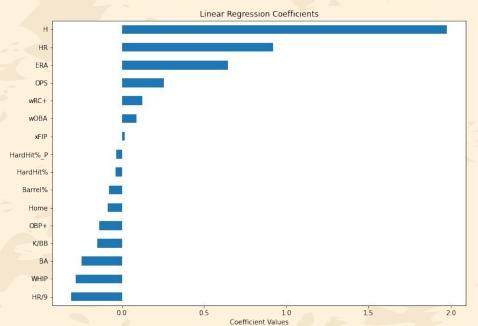






	M	od	lel	M	let	ric	S
							Linear Re

Metric	Value
Training r2	0.69
Testing r2	0.71
Testing MSE	2.95

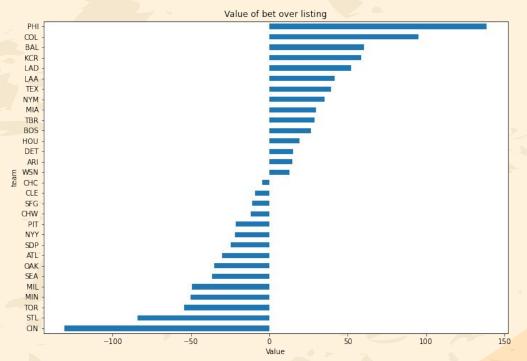








- Scrape daily games and probable pitchers from mlb.com
- Build prediction dataframe
- Get predicted scores
- Generate odds
- Compare to sportsbook odds pulled from the-odds-api







Conclusion and Next Steps

- Time is needed to fully understand the performance of the model relative to bookmakers
- Early returns are good, sample size too small.

- Generate daily predictions
- Assess model performance over the course of a month
- Add more features
- Player value coefficient

Thanks!

Any questions?

CREDITS: This presentation template was created by Slidesgo, including icons by Flaticon and infographics & images by Freepik

Please keep this slide for attribution

