

to (Maybe) Money Pocket Science



By Rob Barnes and Siu Fai Hsu



AGENDA

- ・ 1. What We Do
- 2. The "Rocket Science"
- **当 3. Our NBA Kalman Filters**
- 4. How We're Utilising Databricks
- 🐞 5. How You Make Money
- 👸 6. Summary





What We Do



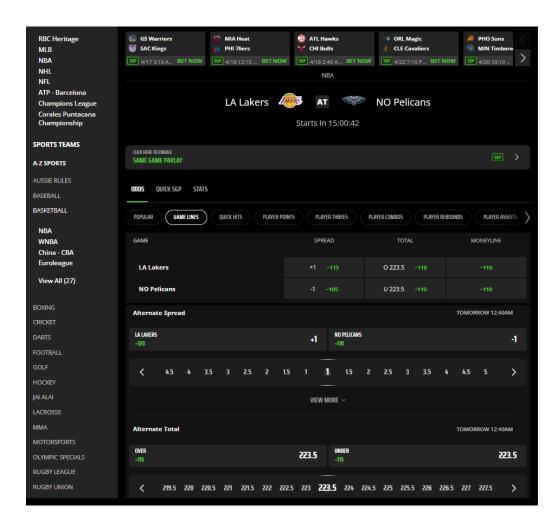


What We Do

We are part of the Sports Data Science Team at DraftKings, responsible for setting the prices customers see on our sportsbook.

Our complex monte carlo simulation based pricing engines are powered by multiple ML models and generate probabilities for various sports events, such as:

- Final game scores
- Timing of the next touchdown
- The next team to score a 3-pointer





Business Problem



WHAT

At the end of 2022 we were asked to produce models that are capable of producing NBA player props markets (over/under player gets x points/assists etc)



HOW

Produce player centric NBA models that are able to understand the value of players

Produce several new player based features and models that to help us produce props markets

Step change our MLOps practices by utilizing Databricks to produce robust feature and model pipelines and monitor our models.









KALMAN FILTERS

Kalman Filters are an **algorithm** that uses a series of measurements to produce an estimate of an unknown variable.

They are designed to work in scenarios where high degrees of **uncertainty** are present, both in the variable you're attempting to estimate and the the method being used to measure.

Kalman Filters also account for **uncertainty in the estimate** which gets updated alongside the
estimate after every measurement.

Estimate Update

$$\hat{oldsymbol{x}}_{n,n} = \hat{oldsymbol{x}}_{n,n-1} + oldsymbol{K}_n(oldsymbol{z}_n - oldsymbol{H}\hat{oldsymbol{x}}_{n,n-1})$$

Kalman Gain

$$oldsymbol{K}_n = oldsymbol{P}_{n,n-1}oldsymbol{H}^Tig(oldsymbol{H}oldsymbol{P}_{n,n-1}oldsymbol{H}^T+oldsymbol{R}_nig)^{-1}$$

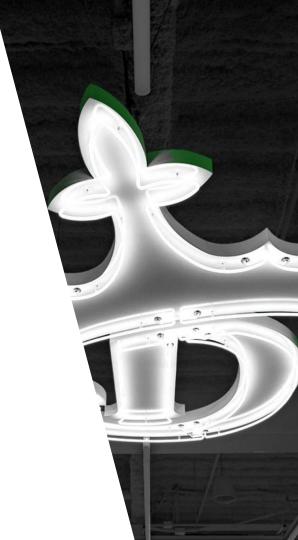
Covariance Update

$$\boldsymbol{P}_{n,n} = \left(\boldsymbol{I} - \boldsymbol{K}_{n} \boldsymbol{H}\right) \boldsymbol{P}_{n,n-1} (\boldsymbol{I} - \boldsymbol{K}_{n} \boldsymbol{H})^{T} + \boldsymbol{K}_{n} \boldsymbol{R}_{n} \boldsymbol{K}_{n}^{T}$$



Applications

ZZ.	Auto Pilot
	Portfolio Optimization
\$	Financial Time Series Prediction
6-6	GPS
8.9	Rocket Science
: Q:	And Many More!!!



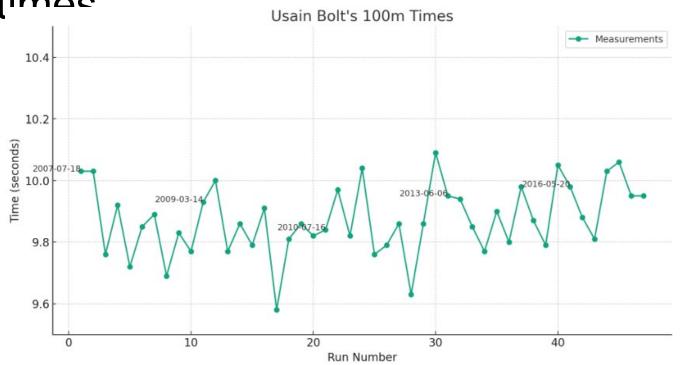


Doesn't Use Kalman Filters



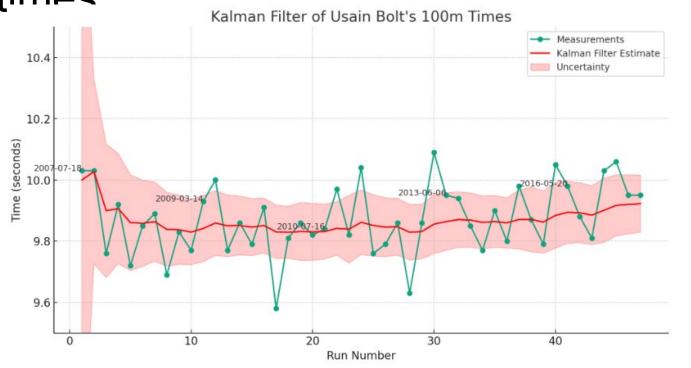


A Simple Example - Usain Bolt's 100





A Simple Example - Usain Bolt's 100 times





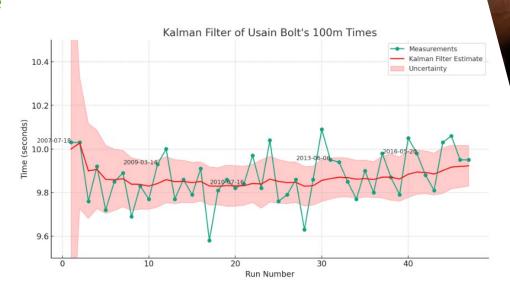
A Simple Example - The Parameters

Starting Expectation

Starting Uncertainty

Measurement Noise

Drift





Kalman Filters as a Rating

System

Kalman Filters can be used for situations where there are multiple players/teams participating, and only a subset of players participate in any given observations

,	Rating System	Elo	Glicko	Trueskill	Kalman Filters
	Uncertainty Value	No	Yes	Yes	Yes
	Measurement Noise	No	No	Yes	Yes
	Updates	Only involved players	Only involved players	Only involved players	All entities/players



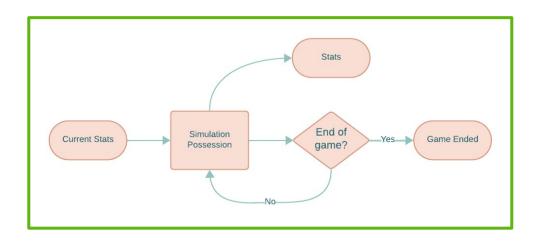






Simulation Engine

At DraftKings, our sports engines are built using Monte Carlo Simulation



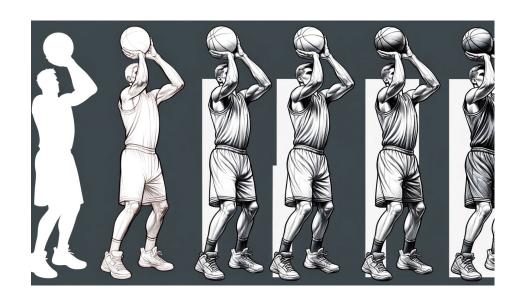




Enrichment Probability Models

During possession simulation, details are enriched sequentially based on probability models.

- Given a three pointer made, who has shoot the three?
- Is the three pointer being assisted?
- If yes, who has passed the ball?





Kalman Filters as Probability Model



Simple

$$P(X) = \frac{1}{1 + e^{-\mu}}$$

- A rating for every team/player
- Simple transformation from ratings to probabilities



Team / Player Specific

Do not need additional features



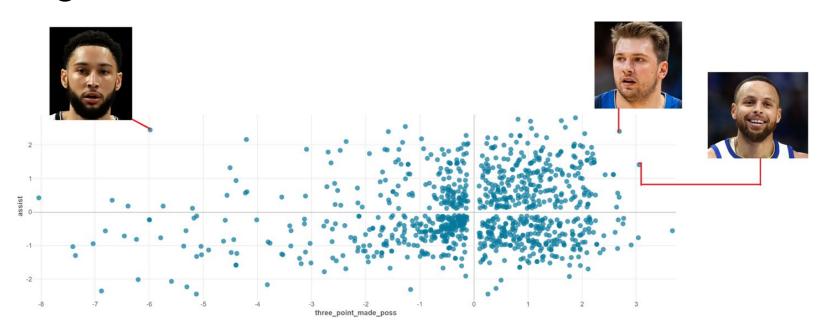
Easy to update

Simple matrix operations to update the ratings after the game





Three Pointer Ratings vs Assists Ratings









"The most valuable commodity I know of is information"

- A quote from 1987 film Wall Street





Before & After Databricks

Before

- Ratings are trained locally and stored in a csv
- Need to do a deployment to update the ratings
- Basically only updates every year
- Not feasible for fast moving player ratings

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ast_ratings_3_0	Microsoft Excel Comma Separat	2 KB
ast_ratings_3_1	rosoft Excel Comma Separat	2 KB
blk_ratings_0_0	Microsoft Excel Comma e., rat	2 KB
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After!



Workflow are set up in the databricks to update dataset and ratings daily



Dataset and ratings can be reproduced easily



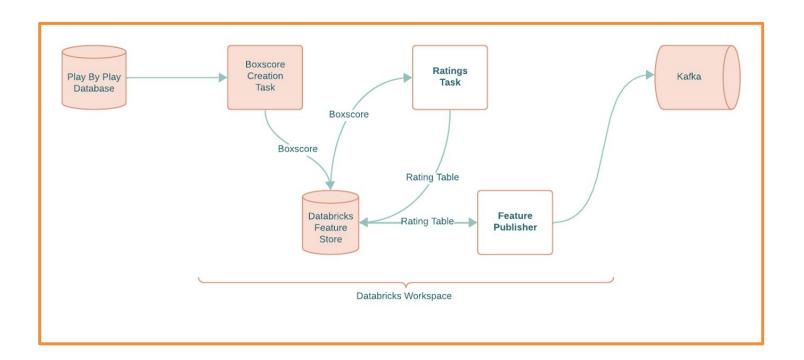
The ratings for calculation is sent downstream automatically



Able to backtest the model performance using historical ratings



Architecture





Rating Tasks





Feature Publisher

CDC

The Feature Publisher task reads the CDC (change data capture) table

> • Figure out which rows have been changed

KAFKA

The change is then sent downstream to kafka

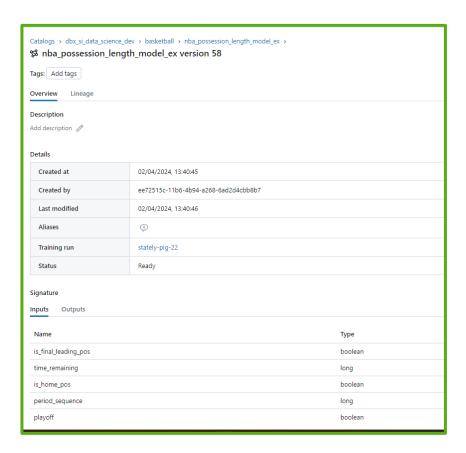
- The updated team/player ratings are mapped to different games
- Form calculation requests by combing ratings with other information



Model Injection

The models are trained on the updated dataset by scheduled workflows.

When the engine pod is started up, it will grab the latest version of the model from the model registry.

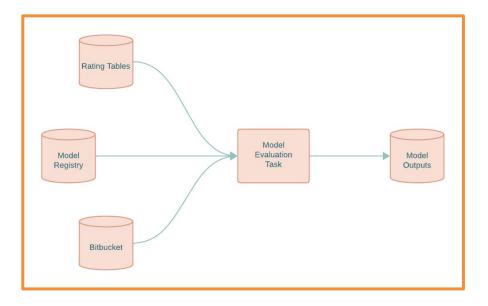




Model Backtesting

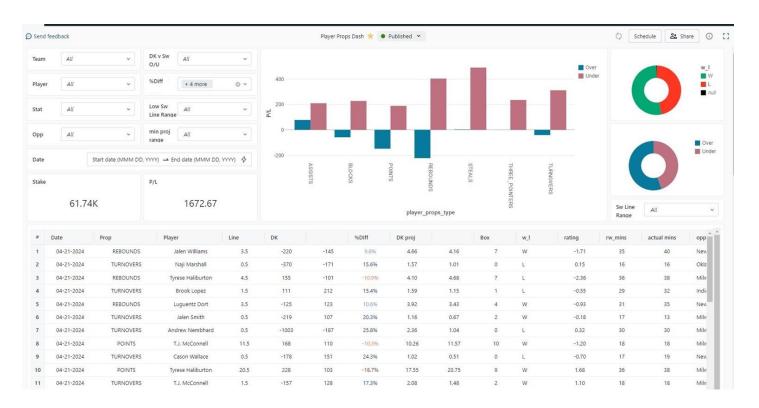
Historical KF ratings

- Historical kf team/player ratings right before the game starts
- Avoid data leakage





Historical Dashboard









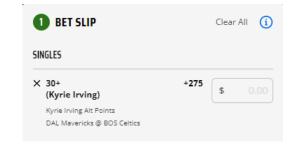


Betting Strategy

NOW LET'S MAKE SOME MONEY

WINNING INGREDIENTS

- Bookmaker offers to return 3.75 times your stake.
- Model suggests **30%** of the time you will win.
- Expected return is 30% x 3.75 = **1.125**





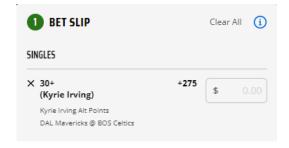


Betting Strategy

NOW LET'S MAKE SOME MONEY

KELLY'S CRITERION

- Maximize the expected growth rate.
- Everytime we only bet a proportion of the bankroll
- b = (p0 1) / (0 1)





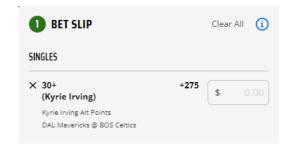


Betting Strategy

NOW LET'S MAKE SOME MONEY

EVALUATION IS KEY

- Analyse on the betting performance to optimise the betting strategy
- You only need to bet when you are confident with your prices







Sportsbook Strategy



We don't have the luxury of choosing when we bet

Need to be confident for ALL of the markets that we offer



Apply Margin

• We can control our prices



Learn from customer's bet

- Identify professional customers
- Adjust the probability of the model based on their bets











Summary

- We achieved our goal to build a model that is able to produce NBA player props
- We produced 200+ separate Kalman Filter based models all producing ratings for various different player and team attributes.
- All of the above models use features engineered through Databricks workflows
- All models are registered on Databricks and retrained and optimized through workflows running on a daily cadence
- The same practices applied here are now being adopted by other sports as they look to do similar.

https://medium.com/draftkings-engineering

Points				TO	DMORROW 12:40AM
PLAYER		OVE	R	UND	ER
Al Horford		O 6.5	-112	U 6.5	-108
Al Horford Bam Adeb Caleb Mari Derrick Wi	зуо	0 19.5	-115	U 19.5	-105
Caleb Mark	in	0 9.5	-120	U 9.5	+100
Derrick Wi	iite	O 12.5	-120	U 12.5	+100
Haywood I	lighsmith	0 6.5		U 6.5	+105
Jaime Jaqu	ez Jr.	O 14.5	-102	U 14.5	-118
Jaylen Bro	wn	0 21.5	+100	U 21.5	-120
Jayson Tat	um	O 25.5	-122	U 25.5	+102
Jrue Holida	ıy	0 9.5	-105	U 9.5	-115
Kristaps Po	orzingis	O 15.5	-142	U 15.5	+120
Jaylen Bro	c	0 8.5	-122	U 8.5	+102
Payton Pri	tchard	0 6.5	+105	U 6.5	-125
Tyler Herro	,	O 20.5	-110	U 20.5	-110



THANK YOU! THANK YOU! Any Questions?

