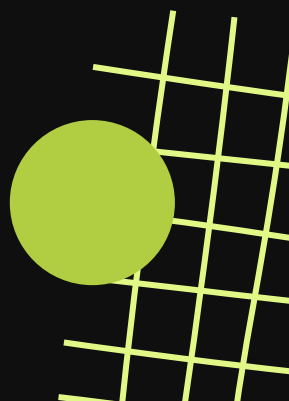




# Tennis Server Quality Metric

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# Contents

**01**

**Motivation**

**02**

**Objectives**

**03**

**Data/EDA**

**04**

**Serving Metrics**

**05**

**Evaluation**

**06**

**Discussion**



# Motivation

- Existing research focuses extensively on match-level prediction (mostly ELO models)
- Basic serving statistics exist, but they tell an incomplete story of server dominance
- Room in the literature for an evaluation of servers in men's and women's tennis



# Research Objectives



**Group players by serve style:** Clustering by serve speed, aces, location entropy, and modal serve location



**Model serve outcomes:** Using server statistics and clusters to predict point outcomes



**Construct server quality metrics:** Creating numerical scores to evaluate servers.

**Clustering**

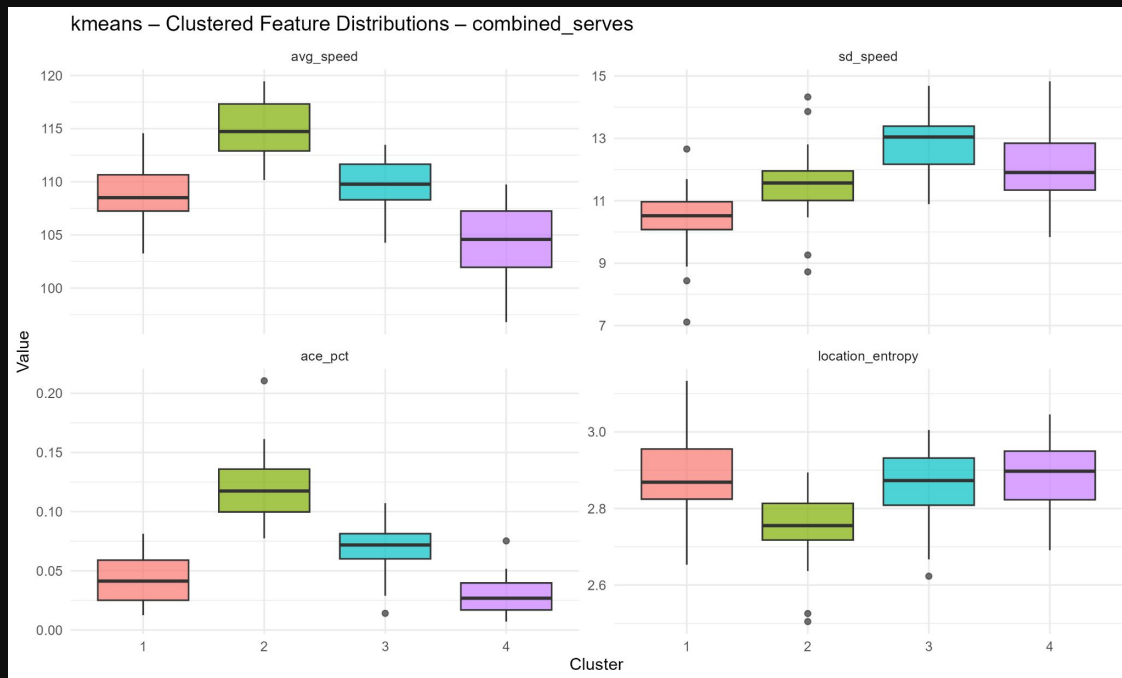


**Modeling**



**Server Metrics**

# K-Means Clustering: Speed, Aces, Entropy



1	cluster	avg_speed	sd_speed
2	1	108.960916920795	10.4145155732768
3	2	114.815266722397	11.5776122609671
4	3	109.746322772352	12.9167511071142
5	4	104.32923706341	12.0743616614806

1	cluster	ace_pct	location_entropy
2	1	0.0433578436481444	2.872178922517
3	2	0.119837475337068	2.75073665561808
4	3	0.0706959470167434	2.86197448159623
5	4	0.0295168777819073	2.8866707644689

Data: Wimbledon Males, 2021-24, from [Jeff Sackmann's point-by-point grand slam data](#)

# Server Quality Metrics

## Results from Prediction Models

- General form of models:  $(Win\_pct \text{ or } serve\_efficiency) \sim (avg\_speed + sd\_speed + location\_entropy + factor(modal\_location) + factor(cluster) + \text{interactions of cluster with all other variables})$
- 3 models to predict win\_pct or serve\_efficiency: linear regression, random forest, and XGBoost
- Server metric = **scaled(prediction) + scaled(residual)**

Separate metrics using different data (Wimbledon/U.S. Open, males/females)  
All metrics scaled to be between [-1, 1]

# Server Quality Metrics

## Weighted Average of Server Statistics

- Formula:  $\sum [(Importance\ Score) * (player\ statistic)]$
- Player statistics: speed (average and sd), entropy, modal serve location (one-hot encoded)
- Importance scores from random forest model that predicts outcome variable (serve efficiency or win percentage)

Separate metrics using different data (Wimbledon/U.S. Open, males/females)  
All metrics scaled to be between [-1, 1]

# Server Quality Metrics

## Weighted Elo

- Baseline metric: server's weighted Elo (wElo)
- Minimum wElo = 1500 in our datasets

Separate metrics using different data (Wimbledon/U.S. Open, males/females)  
All metrics scaled to be between  $[-1, 1]$



# Outcomes for Prediction Models

## Serve Efficiency

- Formula: *(# points won as server where rally count  $\leq 3$ ) / (total # points played as server)*
- How often a serve leads **rapidly** to a point win
- Might undervalue players with good serves who didn't win points quickly

## (Point) Win Percentage

- Formula: *(# points won as server) / (total # points played as server)*
- How often a serve leads to a point win, regardless of rally count
- More confounded by other factors (e.g., opponent skill; other hits during rally)

# Out-of-sample Predictive Performance

## Serve Efficiency (Wimbledon males)

1	cor	p_value	rmse	avg_metric	n	Model
2	0.502909180219369	9.11273224652082e-05	0.98798057154534	-9.2886557031849e-18	55	performance_lm
3	0.524026062162999	4.02518734147486e-05	0.966767672442521	5.20417042793042e-18	55	performance_rf
4	0.521279498615177	4.49032935644374e-05	0.969552981810417	1.0857791938273e-17	55	performance_xgb
5	0.514238774755903	5.91794182818541e-05	0.976656749290078	1.18631430436687e-17	55	weighted_avg
6	0.0214043186327044	0.875575490603796	0.990804165032057	-1.85565751597868e-15	56	avg_welo

Metrics based on 2021-24 data; tested on serve efficiency from 2018-19 data

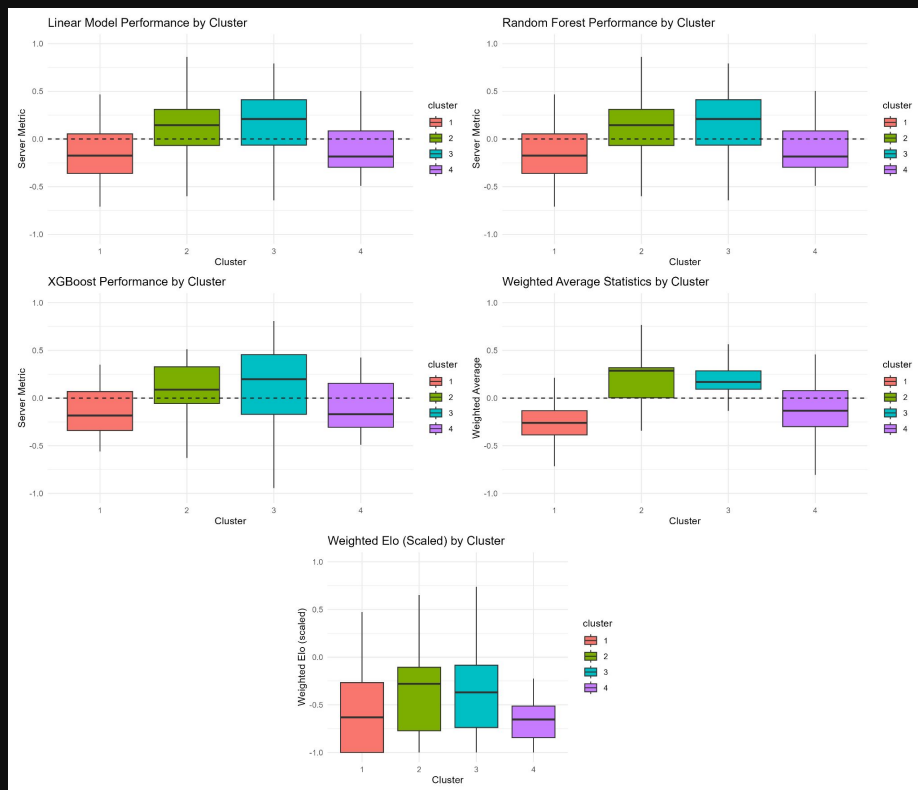
# Out-of-sample Predictive Performance

## Win Percentage (Wimbledon males)

1	cor	p_value	rmse	avg_metric	n	Model
2	0.309249505097055	0.0215953931066117	1.16463848038404	2.11517987468536e-17	55	performance_lm
3	0.298792693380145	0.0266984441446279	1.17342071130787	1.83895473028621e-17	55	performance_rf
4	0.311730967811898	0.0205136945428977	1.16254466562337	5.50183320619462e-18	55	performance_xgb
5	0.13516046428811	0.325175695971142	1.30316167877752	-1.93185114370144e-17	55	weighted_avg
6	0.284334647347935	0.0336846431777086	0.950126490239438	-1.85565751597868e-15	56	avg_welo

Metrics based on 2021-24 data; tested on win percentage from 2018-19 data

# Distn of Server Quality Per Cluster



Server metrics based on  
win\_pct prediction models for  
Wimbledon males data





# Alexander Zverev



## Serve Details

### Average Serve Speed

119.448 mph

### SD Serve Speed

11.783 mph

### Ace Percentage

0.116

### Location Entropy

2.833

### Modal Serve Location

WC - DNCTL

(center; not close to line)

### Cluster

2



## Metrics (based on serve efficiency; scaled from [-1, 1])

### RF performance metric

0.887

### Weighted Stats

0.767


### wElo (scaled)

-0.006

- Placed in "Powerhouse" server cluster
- Known for high speed, powerful serve



# Future Work

- Adding serve spin and placement data to the metrics (if there is data)
  - Ace percentage as an outcome variable
  - Generate actionable insights for players (e.g., personalized training plan for certain serve styles and opponents)
  - Server metric that updates live throughout a match
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