



# Analyzing the Modern NFL Running Back

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# Dataset: Kaggle's "Big Data Bowl"

## NFL's "Next Gen Stats"

- Taken from Kaggle for a competition to predict yards gained
- "Running" or "rushing" plays from 2017- mid 2019
  - Hand-off was made to either a Running Back (RB), Wide Receiver (WR) or Tight End (TE)
- Contains rich information:
  - Player pre-snap X,Y coordinates
    - Offensive/Defensive Formations
  - Player acceleration and speed
  - Play direction and orientation
  - Play result



					Games		Rushing						
Rk	Player	Tm	Age	Pos	G	GS	Att	Yds	TD	1D	Lng	Y/A	Y/G
1	<a href="#">Derrick Henry</a> *	<a href="#">TEN</a>	25	RB	15	15	303	1540	16	73	74	5.1	102.7
2	<a href="#">Nick Chubb</a> *	<a href="#">CLE</a>	24	RB	16	16	298	1494	8	62	88	5.0	93.4
3	<a href="#">Christian McCaffrey</a> *+	<a href="#">CAR</a>	23	RB	16	16	287	1387	15	57	84	4.8	86.7
4	<a href="#">Ezekiel Elliott</a> *	<a href="#">DAL</a>	24	RB	16	16	301	1357	12	78	33	4.5	84.8
5	<a href="#">Chris Carson</a>	<a href="#">SEA</a>	25	RB	15	15	278	1230	7	75	59	4.4	82.0
6	<a href="#">Lamar Jackson</a> *+	<a href="#">BAL</a>	22	QB	15	15	176	1206	7	71	47	6.9	80.4
7	<a href="#">Leonard Fournette</a>	<a href="#">JAX</a>	24	RB	15	15	265	1152	3	55	81	4.3	76.8
8	<a href="#">Josh Jacobs</a>	<a href="#">OAK</a>	21	RB	13	13	242	1150	7	53	51	4.8	88.5
9	<a href="#">Joe Mixon</a>	<a href="#">CIN</a>	23	RB	16	15	278	1137	5	56	41	4.1	71.1
10	<a href="#">Dalvin Cook</a> *	<a href="#">MIN</a>	24	RB	14	14	250	1135	13	60	75	4.5	81.1
11	<a href="#">Marlon Mack</a>	<a href="#">IND</a>	23	RB	14	12	247	1091	8	67	63	4.4	77.9
12	<a href="#">Aaron Jones</a>	<a href="#">GNB</a>	25	RB	16	16	236	1084	16	55	56	4.6	67.8
13	<a href="#">Carlos Hyde</a>	<a href="#">HOU</a>	29	RB	16	14	245	1070	6	55	58	4.4	66.9
14	<a href="#">Mark Ingram</a> *	<a href="#">BAL</a>	30	RB	15	15	202	1018	10	53	53	5.0	67.9
15	<a href="#">Phillip Lindsay</a>	<a href="#">DEN</a>	25	RB	16	16	224	1011	7	38	40	4.5	63.2
16	<a href="#">Saquon Barkley</a>	<a href="#">NYG</a>	22	RB	13	13	217	1003	6	45	68	4.6	77.2
17	<a href="#">Sony Michel</a>	<a href="#">NWE</a>	24	RB	16	14	247	912	7	55	26	3.7	57.0
18	<a href="#">Adrian Peterson</a>	<a href="#">WAS</a>	34	RB	15	15	211	898	5	41	32	4.3	59.9
19	<a href="#">David Montgomery</a>	<a href="#">CHI</a>	22	RB	16	8	242	889	6	50	55	3.7	55.6
20	<a href="#">Todd Gurley</a>	<a href="#">LAR</a>	25	RB	15	15	223	857	12	51	25	3.8	57.1

# What do we already know about running backs?

Predominately, they “run” or “rush” with the ball to gain yardage to score touchdowns.

Rushing Yards & Touchdowns seem to be primary drivers for Pro Bowl selection

# Dataset allows for us to visualize the *how*

Proposed tasks both explore and present new information for analysts and the public alike

## 1: RB Speed

Who are consistently the fastest RBs?

## 2: RB Direction

How well do RBs rush to both sides of the center?

## 3: Defenses

How do defensive formations affect the RB?

## 4: All together

Can we see a rich combination of data for RB rushes?

## 5: Analysis

Can we visually propose an alternative hypothesis for Pro Bowl consideration?



## Pre-requisite steps

01

### Reduce Dataset

Create rush-level dataset by dropping peripheral snap-level information

02

### Create rush vectors

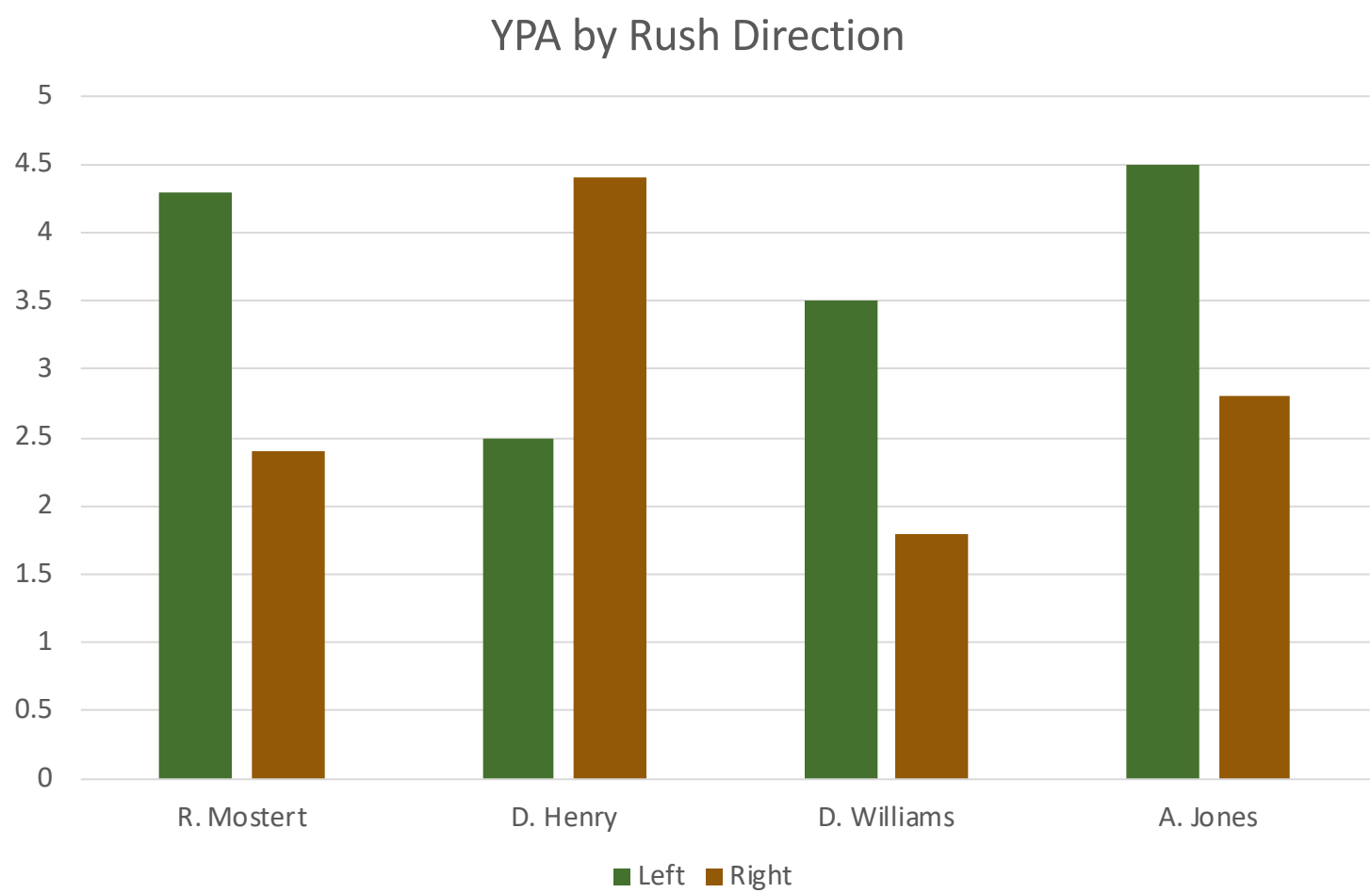
Leverage X, Y coordinates, speed, direction, etc. to develop rush vectors to plot



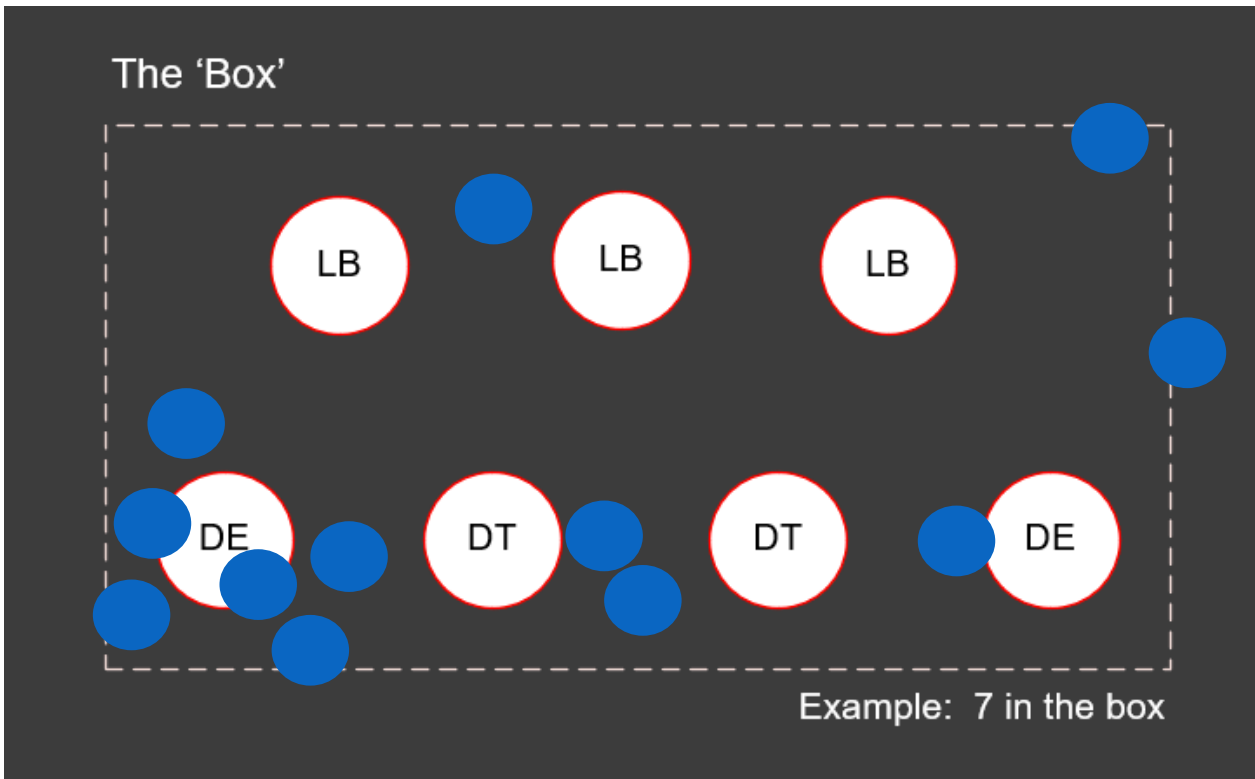
# Programmatic Solution: Leveraging filtering, sorting and movement

Some theoretical plots & sketches...

## Tasks 1 & 2:



## Task 3:



May zoom or filter for specific set or specific players. Lines and dots may move.

## Tasks 4 & 5:

