

# Contract Year Effect in the NBA

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# Introduction

- Contract Effects
  - Work Ethic
  - Overall Goal
- Contract Year Phenomenon
  - Players underperforming
  - Forced effort
- Performance in that year
  - Sign a long term contract



# Literature Review

- Holmstrom (1982) and Gibbons & Murphy (1992) discuss career concerns
  - Applicable to superstars, not applicable to worse players who may mostly care about money
- Contract year phenomenon is a principal-agent problem
  - Incentive incompatible contracts create shirking behavior and moral hazard
    - Alchian and Demsetz (1972), Holmstrom (1979), Ress (1994)
- Sports-specific research yields conflicting results of opportunistic behavior
  - Lehn (182), Scogins (1993), Stiroh (2007), Paulsen (2018) find evidence of opportunistic behavior
  - Krautmann (1990), Maxcy et al. (2002, Berri & Krautmann (2006) do not

# Data Collection - Salary and Advanced Stats

- Scraped Salary Data and advanced stats from basketballreference.com for all current NBA players
  - Scraper visits the basketballreference page for every current NBA player and scrapes relevant statistics
- Focus on the 2016-2017 to 2019-2020 regular seasons

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Season	Age	Tm	Lg	Pos	G	MP	PER	TS%	3PAr	FTr	ORB%	DRB%	TRB%	AST%	STL%	BLK%	TOV%	USG%	OWS	DWS	WS	WS/48	OBPM	DBPM
<a href="#">2013-14</a>	19	<a href="#">MIL</a>	<a href="#">NBA</a>	SF	77	1897	10.8	.518	.282	.483	4.6	16.3	10.2	12.1	1.7	2.6	19.4	15.0	0.1	1.1	1.2	.031	-2.4	-0.1
<a href="#">2014-15</a>	20	<a href="#">MIL</a>	<a href="#">NBA</a>	SG	81	2541	14.8	.552	.056	.445	4.5	19.7	12.2	13.1	1.5	2.8	15.6	19.6	2.2	4.0	6.2	.117	-1.2	1.2
<a href="#">2015-16</a>	21	<a href="#">MIL</a>	<a href="#">NBA</a>	PG	80	2823	18.8	.566	.108	.404	4.6	20.0	12.4	20.0	1.7	3.4	14.8	22.3	4.2	2.9	7.1	.121	1.3	0.7
<a href="#">2016-17</a> *	22	<a href="#">MIL</a>	<a href="#">NBA</a>	SF	80	2845	26.1	.599	.143	.486	5.9	22.6	14.3	26.6	2.3	4.7	13.3	28.3	7.9	4.5	12.4	.210	4.9	2.4
<a href="#">2017-18</a> *	23	<a href="#">MIL</a>	<a href="#">NBA</a>	PF	75	2756	27.3	.598	.100	.457	6.7	25.3	16.0	23.7	2.0	3.3	11.7	31.2	8.3	3.6	11.9	.207	5.3	0.9
<a href="#">2018-19</a> *	24	<a href="#">MIL</a>	<a href="#">NBA</a>	PF	72	2358	<b>30.9</b>	.644	.163	.550	7.3	30.0	19.3	30.3	1.8	3.9	14.8	32.3	8.9	5.5	14.4	<b>.292</b>	6.3	<b>4.1</b>
<a href="#">2019-20</a> *	25	<a href="#">MIL</a>	<a href="#">NBA</a>	PF	57	1763	<b>31.6</b>	.608	.238	.500	7.8	34.5	22.0	33.9	1.5	2.9	13.1	<b>37.4</b>	5.6	<b>4.8</b>	10.4	<b>.282</b>	7.4	<b>4.1</b>
Career			<b>NBA</b>		<b>522</b>	<b>16983</b>	<b>22.9</b>	<b>.594</b>	<b>.147</b>	<b>.478</b>	<b>5.8</b>	<b>23.7</b>	<b>15.0</b>	<b>22.7</b>	<b>1.8</b>	<b>3.4</b>	<b>14.1</b>	<b>26.5</b>	<b>37.2</b>	<b>26.4</b>	<b>63.6</b>	<b>.180</b>	<b>3.1</b>	<b>1.8</b>

# Data Collection - Speed and Box-Out Data



- Scraped Box out and Hustle Data from stats.nba.com for all current NBA players

PLAYER	TEAM	AGE	GP	MIN	BOX OUTS	OFF BOX OUTS	DEF BOX OUTS	TEAM REB ON BOX OUTS	PLAYER REB ON BOX OUTS	% BOX OUTS OFF	% BOX OUTS DEF	% TEAM REB WHEN BOX OUT
Aaron Gordon	ORL	23	78	33.8	2.4	0.2	2.2	1.1	0.4	9.1	90.9	85.6
Aaron Holiday	IND	22	50	12.9	0.5	0.0	0.5	0.3	0.0	0.0	100	92.9
Abdel Nader	OKC	25	61	11.4	0.8	0.0	0.8	0.4	0.2	0.0	100	79.4
Al Horford	BOS	33	68	29.0	6.5	0.5	6.0	3.3	1.1	7.4	92.6	86.0
Al-Farouq Aminu	POR	28	81	28.3	3.6	0.3	3.3	1.6	0.7	8.0	92.0	87.5
Alan Williams	BKN	26	5	5.1	0.8	0.2	0.6	0.0	0.0	25.0	75.0	0.0
Alec Burks	SAC	27	64	21.5	0.5	0.0	0.5	0.3	0.1	0.0	100	95.0
Alex Abrines	OKC	25	31	19.0	1.4	0.0	1.4	0.7	0.1	0.0	100	74.2
Alex Caruso	LAL	25	25	21.2	1.2	0.0	1.2	0.8	0.2	0.0	100	86.4
Alex Len	ATL	26	77	20.1	5.7	0.8	5.0	2.9	0.9	13.4	86.6	89.3
Alex Poythress	ATL	25	21	14.5	2.5	0.3	2.2	1.5	0.7	11.3	88.7	88.6

# Data Collection - Speed and Box-Out Data



## 545 Contracts [Share & more ▼](#) [Glossary](#)

Rk	Player	Tm	Salary						Signed Using	Guaranteed
			2019-20	2020-21	2021-22	2022-23	2023-24	2024-25		
1	<a href="#">Stephen Curry</a>	<a href="#">GSW</a>	\$40,231,758	\$43,006,362	\$45,780,966				Bird Rights	\$129,019,086
2	<a href="#">Chris Paul</a>	<a href="#">OKC</a>	\$38,506,482	\$41,358,814	\$44,211,146				Bird Rights	\$79,865,296
3	<a href="#">Russell Westbrook</a>	<a href="#">HOU</a>	\$38,178,000	\$41,006,000	\$43,848,000	\$46,662,000			Bird Rights	\$123,032,000
4	<a href="#">John Wall</a>	<a href="#">WAS</a>	\$37,800,000	\$40,824,000	\$43,848,000	\$46,872,000			Bird Rights	\$122,472,000
5	<a href="#">James Harden</a>	<a href="#">HOU</a>	\$37,800,000	\$40,824,000	\$43,848,000	\$46,872,000			Bird Rights	\$122,472,000
6	<a href="#">LeBron James</a>	<a href="#">LAL</a>	\$37,436,858	\$39,219,565	\$41,002,273				Cap Space	\$76,656,423
7	<a href="#">Kevin Durant</a>	<a href="#">BRK</a>	\$37,199,000	\$39,058,950	\$40,918,900	\$42,778,850			Sign and Trade	\$117,176,850
8	<a href="#">Blake Griffin</a>	<a href="#">DET</a>	\$34,234,964	\$36,595,996	\$38,957,028				Bird Rights	\$70,830,960
9	<a href="#">Kyle Lowry</a>	<a href="#">TOR</a>	\$33,296,296	\$30,000,000					Bird Rights	\$63,296,296
10	<a href="#">Paul George</a>	<a href="#">LAC</a>	\$33,005,556	\$35,450,412	\$37,895,268				Maximum Salary	\$68,455,968
11	<a href="#">Klay Thompson</a>	<a href="#">GSW</a>	\$32,742,000	\$35,361,360	\$37,980,720	\$40,600,080	\$43,219,440		Bird Rights	\$189,903,600
12	<a href="#">Jimmy Butler</a>	<a href="#">MIA</a>	\$32,742,000	\$34,379,100	\$36,016,200	\$37,653,300			Sign and Trade	\$103,137,300
13	<a href="#">Kemba Walker</a>	<a href="#">BOS</a>	\$32,742,000	\$34,379,100	\$36,016,200	\$37,653,300			Sign and Trade	\$103,137,300
14	<a href="#">Kawhi Leonard</a>	<a href="#">LAC</a>	\$32,742,000	\$34,379,100	\$36,016,200				Cap Space	\$67,121,100
15	<a href="#">Gordon Hayward</a>	<a href="#">BOS</a>	\$32,700,690	\$34,187,085					Cap space	\$32,700,690
16	<a href="#">Mike Conley</a>	<a href="#">UTA</a>	\$32,511,623	\$34,504,132					Cap Space	\$54,938,006
17	<a href="#">Kyrie Irving</a>	<a href="#">BRK</a>	\$31,742,000	\$33,329,100	\$34,916,200	\$36,503,300			Cap space	\$136,490,600
18	<a href="#">Tobias Harris</a>	<a href="#">PHI</a>	\$31,034,483	\$33,517,241	\$36,000,000	\$38,482,759	\$40,965,517		Bird Rights	\$180,000,000
19	<a href="#">Khris Middleton</a>	<a href="#">MIL</a>	\$30,603,448	\$33,051,724	\$35,500,000	\$37,948,276	\$40,396,552		Bird Rights	\$137,103,448
20	<a href="#">Paul Millsap</a>	<a href="#">DEN</a>	\$30,500,000						Cap space	\$30,500,000



# Model and Analysis

Challenges in estimating the contract year effect fall under three categories.

- Player effort cannot be observed, and existing metrics such as distance covered on the field per minute may only be loosely correlated with player effort.
- The effort put in by a particular player may also correlate with other players' performance. Since basketball is a team game, effort put in by one player may synergize with effort put in by other players.
- Heterogeneity in different individuals: unobserved variables such as skill.



# Controlling for individual fixed effects

We use the regression model

$$y_{it} = \alpha_0 + \sum_{j=1}^N \alpha_j \mathbb{1}\{i = j\} + \mathbb{1}\{i_t \in \text{Contract}\} \beta + c'_{it} \gamma + u_{it}$$

Identifying assumption: unobservable effects soaked up by the individual indicator variable that simultaneously affect the outcome player statistics and the explanatory variable and covariates are time-invariant.





# Controlling for team fixed effects

We use the same regression model

$$y_{it} = \alpha_0 + \sum_{j=1}^N \alpha_j \mathbb{1}\{i = j\} + x_{it}\beta + c'_{it}\gamma + u_{it}$$

This time  $x_{it}$  is the weighted number of players on a contract year.

The dataset is grouped by team and the average weighted values of each variable is computed.

# Double LASSO: individual & team fixed effects



When combining datasets from three different sources, the amount of observations decrease drastically and creates a curse of dimensionality problem. To circumvent this issue we employ double LASSO.

# Double LASSO: individual & team fixed effects



1. We first fit a standard Lasso regression:

$$y_{it} = \alpha_0 + \sum_{j=1}^N \alpha_j \mathbb{1}\{i = j\} + \mathbb{1}\{i_t \in \text{Contract}\} \beta + c'_{it} \gamma + u_{it}$$

2. Drop the zero coefficient variables, then do another Lasso regression:

$$\mathbb{1}\{i_t \in \text{Contract}\} = \alpha_0^L + \sum_{j=1}^N \alpha_j^L \mathbb{1}\{i = j\} + c'_{it} \gamma^L + u_{it}$$

3. Do a regular OLS:

$$y_{it} = \alpha_0^O + \sum_{j=1}^N \alpha_j^O + \mathbb{1}\{i_t \in \text{Contract}\} \beta^O + z'_{it} \gamma^O + u_{it}$$

# Double LASSO: individual & team fixed effects



The strategy is similar for team fixed effects.



# Individual Fixed Effect OLS

Table: Using Average Seconds per Dribble as the Dependent Variable

	<i>Dependent variable:</i>
	Average Seconds per Dribble
Contract Year	-0.0247 (0.0368)
Average Minutes Played	0.011** (0.00369)
Current Salary	0.000* (0.000)
Constant	5.117*** (0.664)
Player Fixed Effects	Yes
Year Fixed Effects	Yes
Observations	1,057
R <sup>2</sup>	0.974
Adjusted R <sup>2</sup>	0.952
Residual Std. Error	1.609 (df = 569)
F Statistic	44.844*** (16, 487, 569)

Questions?





# References