# The examination of pay discrimination based on race in the MLB

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## Intro and Motivation

The current makeup of players in the MLB in 2022:

- White 62.1%
- Hispanic or Latino 28.5%
- Black or African American 7.2%
- Asian 1.9%
- Hawaiian or Pacific Islander 0.3%
- American Indian or Alaska Native 0.1%

Been studied before and nothing has been found:

- Medoff (1975)
- Christiano (1986, 1988)
- Kahn (2000)
- King and Palmer (2006)

## Data Set Details

## At its largest:

• Hitters: 587 observations, 26 variables

• Pitchers: 286 observations, 31 variables

#### Data decisions:

- Need to be eligible to qualify for season awards
- Need to have signed a contract, can't be making league minimum

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Table	1:	Variable	Definitions

Variable

Definition

Variable	Delinion
Team	Team for that year
ln(Salary)	The Natural Log of each observations salary for that year
WAR	Wins Above Replacement
Age	Player's age that season
Service Time	Years player has played in the MLB
NonWhite	1 if player is not white, 0 otherwise
International	1 if player is not from USA, 0 otherwise
Year	year
Traditional Stats	
Hits	Number of hits collected by player
Home Runs	Number of home runs collected by player
Strikeouts	Number of strikeouts collected by player
Walks	Number of walks collected by player
AVG	Batting average by player
SLG	Slugging percentage
OBP	On base percentage
OPS	On base percentage plus slugging
Statcast Stats	
xBA	Expected batting average
xSLG	Expected slugging percentage
xwOBA	Expected weighted on base average
xOBP	Expected on base percentage
xISO	Expected isolated power
Avg EV	Average exit velo of baseball
Avg LA	Average launch angle of baseball
Sweet Spot %	Percent of at bats that are a batted-ball event with a launch angle between eight and
	32 degrees
Barrel %	Percent of at bats that are a batted ball with the perfect combination of exit velocity and
	launch angle

# First Model

 $\ln(\text{Salary}) = \beta_0 + \beta_1 * \text{WAR} + \beta_2 * \text{age} + \beta_3 * \text{NonWhite} + \beta_4 * \text{year}$ 

	10000 0. 1
	(1)
	ln_salary
war	0.0711***
	(5.19)
	0.141***
age	0.141***
	(15.22)
nonwhite	-0.00272
HOHWIH TO	(-0.05)
	(-0.03)
year	0.0220
	(1.84)
	(2133)
servicetime	
Constant	-32.99
	(-1.36)
	(-1.50)
Traditional Stats	No
Sabermetric Stats	No
Team Fixed Effects	No
N	587
$R^2$	0.2872

# Seventh Model

	(7)
	ln_salary
war	0.0358
	(1.72)
age	
1:4-	0.160**
nonwhite	-0.169**
	(-3.11)
year	
year	
servicetime	0.174***
	(19.28)
Constant	10.18***
	(5.42)
Traditional Stats	Yes
Sabermetric Stats	Yes
Sabermetric Stats	Yes
Team Fixed Effects	Yes
N	535
0	

0.6103

$\ln(\text{Salary}) = \beta_0 + \beta_1 * \text{WAR} + \beta_2 * \text{NonWhite} + \beta_3 * \text{ServiceTime}$	$me + \sum \beta_{i+3} * 1$	$\operatorname{Frad}_i + \sum \beta_{I+j+3}$	$*Sabr_i+Team F.E.$
	i	j	

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	$\ln_{\text{-}} \text{salary}$	$\ln_{-}$ salary	$\ln_{-}$ salary	$ln_salary$	$\ln_{\text{-}} \text{salary}$	$ln_salary$	ln_salary
war	0.0711***	0.0688***	0.0369*	0.0366*	0.0389	0.0438**	0.0358
	(5.19)	(5.96)	(2.41)	(2.39)	(1.92)	(2.85)	(1.72)
age	0.141***						
	(15.22)						
nonwhite	-0.00272	-0.188***	-0.176***	-0.176***	-0.150**	-0.158**	-0.169**
	(-0.05)	(-3.81)	(-3.43)	(-3.43)	(-2.77)	(-2.95)	(-3.11)
year	0.0220	0.0322**		0.00557			
	(1.84)	(3.10)		(0.45)			
servicetime		0.180***	0.175***	0.175***	0.173***	0.176***	0.174***
		(22.56)	(21.40)	(21.38)	(19.24)	(20.46)	(19.28)
Constant	-32.99	-50.43*	10.36***	-0.836	14.13***	9.500***	10.18***
	(-1.36)	(-2.41)	(6.21)	(-0.03)	(29.53)	(5.32)	(5.42)
Traditional Stats	No	No	No	No	Yes	No	Yes
Sabermetric Stats	No	No	Yes	Yes	No	Yes	Yes
Team Fixed Effects	No	No	No	No	Yes	Yes	Yes
N	587	587	587	587	535	535	535
$R^2$	0.2872	0.4685	0.5068	0.5069	0.5794	0.6005	0.6103

t statistics in parentheses

## Effect of NonWhite on In(Salary) -Hitters

<sup>\*</sup> p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001

•	

Constant									
War			(1)	(2)	(3)	(4)	(5)	(6)	(7)
Age   1.37   1			$ln_salary$	$\ln_{-}$ salary	$\ln_{-}$ salary	ln_salary	$\ln_{-}$ salary	$\ln_{\text{-}}$ salary	ln_salary
age		war	0.0689***	0.0669***	0.0333*	0.0329*	0.0322	0.0401**	0.0285
Constant			(5.02)	(5.78)	(2.18)	(2.15)	(1.58)	(2.60)	(1.37)
Year   1.15   (-1.03)   (-3.79)   (-3.81)   (-2.61)   (-2.97)   (-3.09)		age							
Constant   Fraditional Stats   No   No   No   No   Yes   Yes   Yes   No   No   No   No   No   No   No   N		international							
- Hitters	Effect of International on In(Salary)	year							
(-1.36)       (-2.40)       (6.22)       (-0.19)       (29.37)       (5.25)       (5.33)         Traditional Stats       No       No       No       No       Yes       No       Yes         Sabermetric Stats       No       No       No       Yes       Yes       Yes       Yes         Team Fixed Effects       No       No       No       No       Yes       Yes       Yes         N       587       587       587       587       535       535       535		servicetime							
Sabermetric Stats No No Yes Yes No Yes Yes Yes $\frac{\text{Team Fixed Effects}}{N} = \frac{N_0}{N_0} = \frac{N_0}{$		Constant							
$\begin{tabular}{ c c c c c c c c c c c c c c c c c c c$		Traditional Stats	No	No	No	No	Yes	No	Yes
N 587 587 587 585 535 535		Sabermetric Stats	No	No	Yes	Yes	No	Yes	Yes
N 587 587 587 585 535 535		Team Fixed Effects	No	No	No	No	Yes	Yes	Yes
$R^2$ 0.2889 0.4700 0.5090 0.5093 0.5787 0.6006 0.6102									
		$R^2$	0.2889	0.4700	0.5090	0.5093	0.5787	0.6006	0.6102

t statistics in parentheses

<sup>\*</sup> p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001

Effect of NonWhite on In(Salary) - Pitchers

				,			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	ln_salary	$\ln_{-}$ salary	$ln\_salary$	$\ln_{-}$ salary	$\ln_{-}$ salary	$ln\_salary$	ln_salary
war	0.0682**	0.0632**	0.0283	0.0283	0.00382	0.0218	0.0153
	(3.28)	(3.24)	(0.79)	(0.78)	(0.07)	(0.51)	(0.26)
age	0.108***						
	(8.72)						
nonwhite	-0.111	-0.125	-0.0457	-0.0453	-0.0494	-0.0216	-0.00828
	(-1.22)	(-1.47)	(-0.53)	(-0.52)	(-0.56)	(-0.24)	(-0.09)
year	-0.00512	0.00281		-0.00175			
•	(-0.28)	(0.16)		(-0.08)			
servicetime		0.132***	0.134***	0.134***	0.122***	0.124***	0.124***
		(11.09)	(11.21)	(11.17)	(10.00)	(10.23)	(9.74)
Constant	22.84	9.108	19.61***	23.16	14.26***	19.49***	18.92***
	(0.62)	(0.26)	(5.35)	(0.49)	(11.55)	(5.05)	(4.51)
Traditional Stats	No	No	No	No	Yes	No	Yes
Sabermetric Stats	No	No	Yes	Yes	No	Yes	Yes
Team Fixed Effects	No	No	No	No	Yes	Yes	Yes
N	286	286	286	286	251	251	251
$R^2$	0.2407	0.3290	0.3924	0.3924	0.5832	0.5737	0.5968

t statistics in parentheses

<sup>\*</sup> p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001

	war
	age
	internat
	year
Effect of International on In(Salary) - Pitchers	servicet
	Constan
	Traditio

				\ 0/			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	ln_salary	$\ln_{-}$ salary	$\ln_{\text{-}} \text{salary}$	$\ln_{\text{salary}}$	$\ln_{\text{salary}}$	ln_salary	$\ln_{\text{salary}}$
war	0.0678**	0.0631**	0.0282	0.0282	0.00332	0.0212	0.0152
	(3.27)	(3.25)	(0.79)	(0.78)	(0.06)	(0.50)	(0.25)
	0.107***						
age	0.107***						
	(8.67)						
international	-0.181	-0.172	-0.0944	-0.0942	-0.0554	-0.0297	-0.0104
	(-1.82)	(-1.84)	(-1.02)	(-1.01)	(-0.57)	(-0.31)	(-0.10)
	(-1.02)	(-1.04)	(-1.02)	(-1.01)	(-0.51)	(-0.51)	(-0.10)
year	-0.00353	0.00454		-0.000630			
	(-0.19)	(0.26)		(-0.03)			
		, ,					
servicetime		0.131***	0.133***	0.133****	0.122***	0.124***	0.124***
		(11.01)	(11.13)	(11.08)	(9.91)	(10.17)	(9.70)
	40.05	L =0.0	40.00000	20.01		40 80000	10.00444
Constant	19.85	5.796	19.63***	20.91	14.35***	19.58***	18.92***
	(0.54)	(0.17)	(5.42)	(0.44)	(11.58)	(5.07)	(4.54)
Traditional Stats	No	No	No	No	Yes	No	Yes
Sabermetric Stats	No	No	Yes	Yes	No	Yes	Yes
T 1 T 1 T C +	NT.	N.T.	NT.	NT	37	37	37
Team Fixed Effects	No	No	No	No	Yes	Yes	Yes
N	286	286	286	286	251	251	251
$R^2$	0.2456	0.3319	0.3940	0.3941	0.5832	0.5738	0.5968

t statistics in parentheses

<sup>\*</sup> p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001

## Conclusion

## Findings:

- There is statistically significant evidence to support that there is pay discrimination within the MLB for hitters
- ServiceTime is a better representative for 'experience'
- Traditional stats are still great variables to use when predicting player value

#### Next Steps:

- Further examine other factors on salary differentials
  - Player position, contract length, and free agency status
- Develop more accurate measure for a player's race
- Further examine the effectiveness of diversity and inclusion in sports