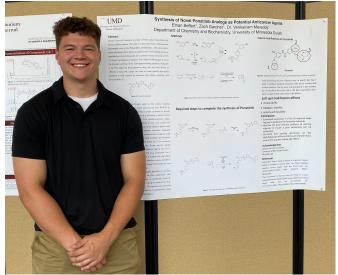
# A Statistical Review of Tommy John Surgery

By Chad Fox and Ethan Beffert



## Motivation's

- Chad
  - ▶ Plays Baseball at UMD
  - ► The issue directly affects my position

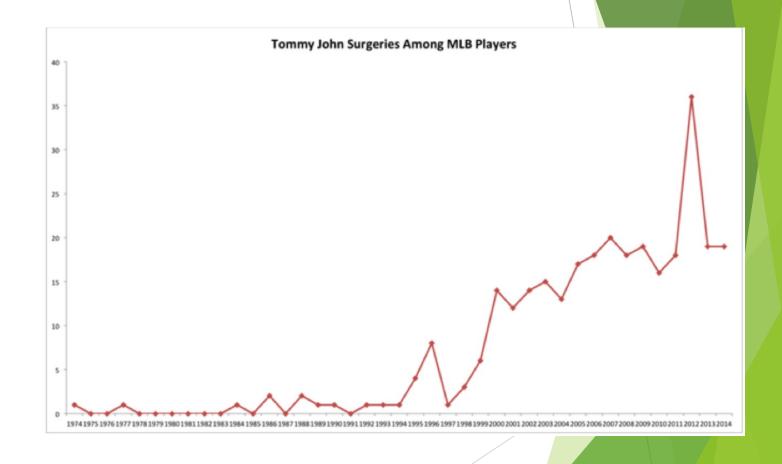


- Ethan
  - ▶ Played Football at UMD
  - I do research in the pharmacy school

## Literature review-Motivation

Tommy John's Surgery has increased in number greatly

- Possible Causes
  - Sports specialization
  - Increased Velocity
  - Youth sports coaches





## What is Tommy John Surgery

- UCL Reconstruction Surgery
- UCL is a ligament on the inner side of the elbow
- Used a harvested tendon to repair the injury (hamstring or big toe)
- Could take a year or more typically to recover

## Literature review- Combining the data

Rk.	Player	Year	BF	<b>K</b> %	BB%	Hard Hit %	New Adjusted EV	Whiff %	Swing %	4-Seam Avg MPH	4-Seam Avg Spin
1	Burnes, Corbin	2021	657	35.6	5.2	30.5	92.7	36.5	49.2	96.4	2607
2	Kershaw, Clayton	2016	544	31.6	2.0	27.3	92.4	31.0	51.7	93.6	2293
3	Bauer, Trevor	2020	278	36.0	6.1	38.0	93.7	30.2	46.4	93.5	2779
4	Kershaw, Clayton	2015	890	33.8	4.7	24.4	92.0	32.5	51.4	94.3	2215
5	Suárez, Ranger	2024	95	28.4	4.2	25.4	92.2	24.6	50.0	92.4	2072
6	Blanco, Ronel	2024	102	21.6	11.8	23.5	91.5	26.7	49.5	93.6	2196
7	Sale, Chris	2018	617	38.4	5.5	26.8	92.5	34.9	48.2	95.2	2357
8	🤵 Ryan, Joe	2024	93	32.3	3.2	32.2	93.2	27.1	53.3	93.5	2222

Baseball Savant

2	Player	TJ Surgery Date	Team	Level	Position	Throws	Country	High School	College(s)	Age	Return Date (same level)	Recovery Time (months)	mlbamid	fgid	Surgeon(s)
3	Endy Rodriguez	12/12/2023	PIT	MLB	С	R	Dominican			23			682848	25332	
4	Johan Oviedo	12/1/2023	PIT	MLB	Р	R	Cuba			25			670912	22487	
5	Jovani Moran	11/1/2023	MIN	AAA	Р	L	Puerto Rico			26			663558	20422	
6	Taylor Broadway	10/18/2023	BOS	AA	Р	R	United States	Texas	Mississippi	26			699479	sa3017205	
7	Felix Bautista	10/9/2023	BAL	MLB	Р	R	Dominican			28			642585	20666	Dr. Keith Meister
8	Sandy Alcantara	10/6/2023	MIA	MLB	Р	R	Dominican			27			645261	18684	Dr. Keith Meister
9	Angel Perdomo	10/6/2023	PIT	MLB	Р	L	Dominican			29			622780	17759	
10	Riley Greene	9/20/2023	DET	MLB	OF	L*	United States	Florida		22			682985	25976	Dr. Keith Meister
11	Jasson Dominguez	9/20/2023	NYY	MLB	OF	R	Dominican			20			691176	28080	Dr. Keith Meister
12	Cole Waites	9/13/2023	SF	AAA	Р	R	United States	Georgia	West Alabama	25			686972	25665	Dr. Keith Meister
13	Drey Jameson	9/1/2023	ARI	MLB	Р	R	United States	Indiana	Ball State	25			686753	26260	

Google sheet of players with Tommy John surgeries

## **Data Cleaning**

Merged data sets into one file

- Removed rows with missing data
- 35 total pitchers with TJ
  - > 7 Dr. James Andrews
  - ▶ 11 Dr. Neal ElAttrache
  - ▶ 17 other doctors

```
DELETE FROM tommy john
WHERE surgeon is null;
DELETE FROM tommy john
WHERE recovery time is null;
DELETE FROM tommy john
WHERE level != 'MLB';
DELETE FROM tommy_john
WHERE position != 'P';
```



## Baseball Statistics quick rundown

- K percentage (-)
- ▶ BB percentage (+)
- Average exit velocity (+)
- ► Hard hit percentage (+)
- Fastball Velocity (-)
- ► Fastball rpm (-)
- ► ERA (+)
- Whip (+)

## Research Questions

How does the performing surgeon in a Tommy John surgery impact the postoperative pitching statistics and outcomes of major league baseball pitchers?

Is there a significant difference between the statistics of pitchers who have had Tommy John surgery and those who haven't?



## Proposed models

Dep. Variabl	e:	kperce	ent R-squa	ared:		0.081
Model:				R-squared:		0.024
Method:		Least Squar	es F-stat			1.413
Date:	Mo	n. 15 Apr 20	24 Prob	F-statistic	):	0.258
Time:		22:35	12 Log-L:	ikelihood:		-99.716
No. Observat	ions:		35 AIC:			205.4
Df Residuals	:		32 BIC:			210.1
Df Model:			2			
Covariance T	ype:	nonrobi	ist			
	coef	std err	t	P> t	[0.025	0.975]
const	-2.8786	1.060	-2.709	0.011	-5.029	-0.712
andrews	2.8563	1.962	1.455	0.155	-1.141	6.854
neal	2.1706	1.691	1.284	0.208	-1.274	5.615
Omnibus:		0.3	84 Durbin	n-Watson:		1.826
Prob(Omnibus	):	0.9	103 Jarque	e-Bera (JB):		0.410
		-0.4	M7 Prob(	1B):		0.815
Skew:						

Dep. Variab	le:	bbpero	ent R-squ	ared:		0.03		
Model:			OLS Adj.	R-squared:		-0.030		
Method:		Least Squa	res F-sta	tistic:		0.502		
Date:	Mo	on, 15 Apr 2	024 Prob	Prob (F-statistic):				
Time:		22:35	:13 Log-L					
No. Observa	tions:		35 AIC:			189.		
Df Residual	s:		32 BIC:			194.4		
Df Model:			2					
Covariance	Type:	nonrob	oust					
	coef	std err	t	P> t	[0.025	0.975		
const	0.3412	0.847	0.403	0.690	-1.384	2.06		
andrews	-1.5697	1.568	-1.001	0.324	-4.764	1.62		
neal	-0.5230	1.351	-0.387	0.701	-3.275	2.22		
Omnibus:		0.	694 Durbi	n-Watson:		2.45		
Prob(Omnibu	s):	0.	707 Jarqu	e-Bera (JB):		0.71		
Skew:		0.	072 Prob(	JB):		0.70		
Kurtosis:		2.	317 Cond.	No.		3.3		

Dep. Variabl	e:	exitv			ared:		0.116
Model:					R-squared:		0.055
Method:		Least Squa			tistic:		1.984
Date:	M				(F-statistic):		0.154
Time:		22:35	:13	Log-L	ikelihood:		-65.797
No. Observat	ions:		35	AIC:			137.6
Df Residuals	:		32	BIC:			142.3
Df Model:			2				
Covariance T	ype:	nonrob	ust				
	coef	std err		t	P> t	[0.025	0.975]
const	0.5471	0.402	1.	360	0.183	-0.272	1.366
andrews	-1.1899	0.745	-1.5	598	0.120	-2.707	0.327
neal	-1.0652	0.642	-1.	560	0.107	-2.372	0.242
Omnibus:		0.	690	Durbi	n-Watson:		2,231
Prob(Omnibus	):	0.	708	Jargu	e-Bera (JB):		0.377
Skew:		−a.	254	Prob	JB):		0.828
Kurtosis:		2	975	Cond.	No.		3.32

Dep. Variable	91	hard	hit R-squ	ared:		0.091
Model:			OLS Adj.	R-squared:		0.034
Method:		Least Squa	res F-sta	tistic:		1.599
Date:		lon, 15 Apr 2	024 Prob	(F-statisti	c):	0.218
Time:		22:35	:13 Log-L	ikelihood:		-112.40
No. Observat:			35 AIC:			230.8
Df Residuals:			32 BIC:			235.5
Df Model:			2			
Covariance Ty	/pe:	nonrob	ust			
	coef	std err	t	P> t	[0.025	0.975]
const	0.3294	1.523	0.216	0.830	-2.773	3.432
andrews	-4.5008	2.820	-1.596	0.120	-10.245	1.244
neal	-3.1567	2.430	-1.299	0.203	-8.106	1.793
Omnibus:		4.	872 Durbi	n-Watson:		1.906
Prob(Omnibus)	:	0.	088 Jargu	e-Bera (JB)	:	3,456
Skew:		-0.	567 Prob(	JB):		0.178
Kurtosis:		4	038 Cond.	Mo		3.32

Dep. Variab	ble:	fastve	lo R-squa	red:		0.077
Model:		0	LS Adj. R	-squared:		0.026
Method:		Least Squar	es F-stat	istic:		1.344
Date:	Mo	n, 15 Apr 20	124 Prob (	F-statistic	):	0.275
Time:		22:35:	13 Log-Li	kelihood:		-73.038
No. Observa	ations:		35 AIC:			152.1
Df Residua	ls:		32 BIC:			156.7
Df Model:			2			
Covariance	Type:	nonrobu	ist			
	coef	std err	t	P> t	[0.025	0.975]
const	0.2471	0.495	0.499	0.621	-0.760	1.255
andrews	1.0815	0.916	1.181	0.246	-0.784	2.947
neal	-0.5289	0.789	-0.670	0.508	-2.136	1.079
Omnibus:		17.1	33 Durbin	-Watson:		1.344
Prob(Omnib	us):	0.0	188 Jarque	-Bera (JB):		24.807
Skew:		-1.2				4.10e-06
Kurtosis:		6.2	60 Cond.	No.		3.32

Dep. Varial Model: Method: Date: Time: No. Observ: Df Residua Df Model: Covariance	Mo ations: ls:	Least Squa	DLS Adj res F-s 024 Pro :13 Log 35 AIC 32 BIC 2		ic):	0.023 -0.039 0.3697 0.694 -220.00 446.0 450.7
	coef	std err	t	P> t	[0.025	0.9751
const andrews neal	-3.4118 43.4118 -11.2246	32.952 61.015 52.573	-0.104 0.711 -0.214	0.918 0.482	-70.532 -80.871 -118.312	63.709 167.695 95.863
Omnibus: Prob(Omnib Skew: Kurtosis:	us):	-1.	000 Jar 146 Pro	bin-Watson: que-Bera (JB) b(JB): d. No.	:	1.720 44.455 2.22e-10 3.32

- Recovery time vs K percentage
  - P value 0.258
- Recovery time vs BB percentage
  - P value 0.5024
- Recovery time vs Average Exit Velocity
  - P value 0.154
- Recovery time vs Hard Hit percentage
  - P value 0.218
- Recovery time vs Average fastball velocity
  - P value 0.275
- Recovery time vs Fastball rpm
  - P value 0.694

#### OLS Regression Results

e:	wh	nip R-squa	red:		0.131
	(	DLS Adj. F	R-squared:		0.077
	Least Squar	res F-stat	istic:		2.422
Mo	n, 15 Apr 20	024 Prob (	F-statistic	:):	0.105
	22:35:	:13 Log-Li	ikelihood:		-3.0036
ions:		35 AIC:			12.01
:		32 BIC:			16.67
		2			
ype:	nonrobu	ıst			
coef	std err	t	P> t	[0.025	0.975]
0.1494	0.067	2.234	0.033	0.013	0.286
-0.2351	0.124	-1.899	0.067	-0.487	0.017
-0.1803	0.107	-1.690	0.101	-0.398	0.037
	4.6	066 Durbir	 n-Watson:		2.236
):	0.1	131 Jarque	e-Bera (JB):		2.672
	0.5	595 Prob(3	IB):		0.263
	3.6	543 Cond.	No.		3.32
	Modions: : : :ype: 	Least Squa Mon, 15 Apr 20 22:35 ions: : ype: nonrobs coef std err 0.1494 0.067 -0.2351 0.124 -0.1803 0.107 4.0 0: 0.0	OLS   Adj.	OLS Adj. R-squared: Least Squares F-statistic: Mon, 15 Apr 2024 Mon, 15 Apr 2024 Prob (F-statistic) 1001: 1002: 1003: 1003: 1004: 1004: 1005: 1005: 1005: 1006: 1007: 1006: 1007: 1006: 1007: 10	OLS Adj. R-squared: Least Squares F-statistic: Mon, 15 Apr 2024 Prob (F-statistic): 22:35:13 Log-likelihood: ions: 35 ALC: 32 BLC: 22 ype: nonrobust

111 250000	I U LI I VI 3 033	ume char che	LUVOI TOIL	e marity of	the errora .	13 COLLECT
		OLS Reg	ression Re			
Dep. Varial	ble:	e	ra R-squ	ared:		0.149
Model:		0	LS Adj.	R-squared:		0.096
Method:		Least Squar	es F-sta	tistic:		2.795
Date:	Mo	n, 15 Apr 20	24 Prob	(F-statistic	):	0.0761
Time:		22:35:	13 Log-L	ikelihood:		-64.252
No. Observa	ations:		35 AIC:			134.5
Df Residua	ls:		32 BIC:			139.2
Df Model:			2			
Covariance	Type:	nonrobu	st			
	coef	std err	t	P> t	[0.025	0.975]
const	0.7553	0.385	1.963	0.058	-0.029	1.539
andrews	-0.9696	0.713	-1.361	0.183	-2.421	0.482
neal	-1.3998	0.614	-2.280	0.029	-2.650	-0.149

0.886 Jarque-Bera (JB): 0.110 Prob(JB): 2.500 Cond. No.

0.804 3.32

#### OLS Regression Results

Omnibus:

Kurtosis:

Prob(Omnibus):

Dep. Variable:	recov	R-squar	ed:		0.194
Model:	0LS	Adj. R-	-squared:		0.144
Method:	Least Squares	F-stati	istic:		3.854
Date:	Mon, 15 Apr 2024	Prob (F	-statistic	:):	0.0316
Time:	22:35:12	Log-Lik	kelihood:		-127.31
No. Observations:	35	AIC:			260.6
Df Residuals:	32	BIC:			265.3
Df Model:	2				
Covariance Type:	nonrobust				
=======================================					
coe	f std err	t	P> t	[0.025	0.975]
const 20.0588	3 2.332	8.601	0.000	15.308	24.809
andrews 9.3697	7 4.318	2.170	0.038	0.574	18.166
neal -3.3316	3.721 -	-0.895	0.377	-10.911	4.248
Omnibus:	 14.669		======= -Watson:		1.845
Prob(Omnibus):	0.001		Bera (JB):		18.040
Skew:	1.174				0.000121
Kurtosis:	5.619	Cond. N	•		3.32
=======================================	5.019				3.32

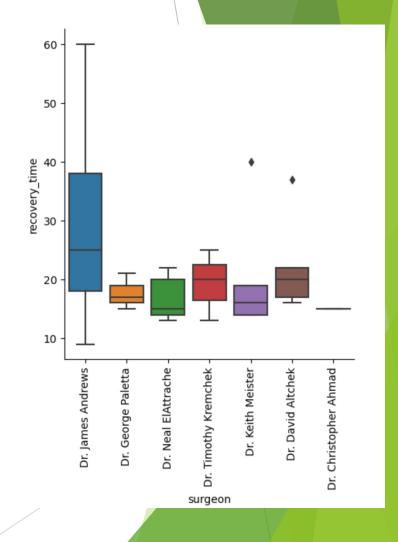
## Proposed Models

- Recovery time vs Era
  - **0.0761**
- Recovery time vs WHIP
  - **0.105**
- Recovery time vs Surgeon
  - P value 0.0316

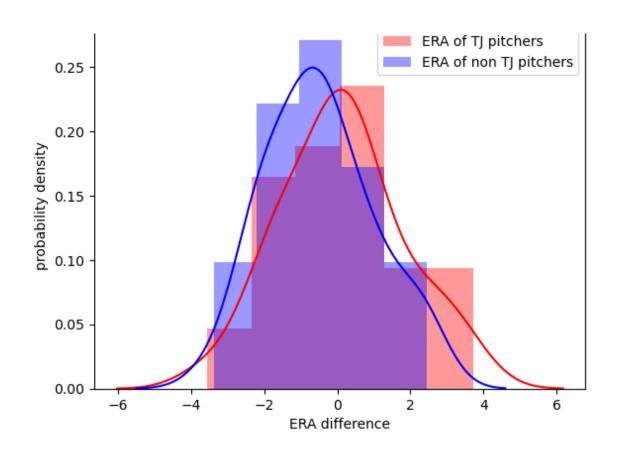
## The relationship between surgeon and recovery time

[12]:

	count	mean	std	min	25%	50%	75%	max
surgeon								
Dr. Christopher Ahmad	1.0	15.000000	NaN	15.0	15.0	15.0	15.0	15.0
Dr. David Altchek	5.0	22.400000	8.502941	16.0	17.0	20.0	22.0	37.0
Dr. George Paletta	3.0	17.666667	3.055050	15.0	16.0	17.0	19.0	21.0
Dr. James Andrews	7.0	29.428571	17.775049	9.0	18.0	25.0	38.0	60.0
Dr. Keith Meister	5.0	20.600000	11.036304	14.0	14.0	16.0	19.0	40.0
Dr. Neal ElAttrache	11.0	16.727273	3.495452	13.0	14.0	15.0	20.0	22.0
Dr. Timothy Kremchek	3.0	19.333333	6.027714	13.0	16.5	20.0	22.5	25.0

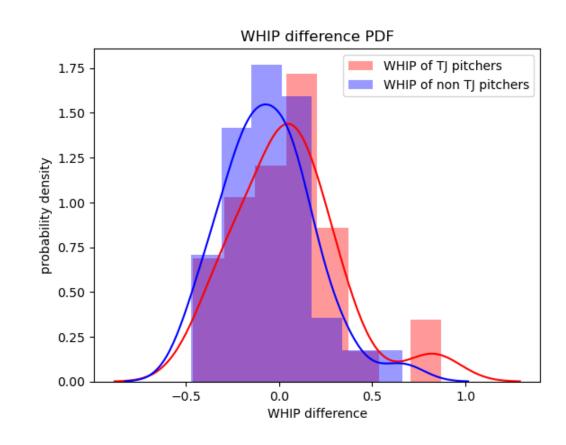


## T-Test TJ vs no TJ ERA



- ttestERA=ttest\_ind(era,earNOTJ)
- TtestResult(statistic=1.591418198901 537
- pvalue=0.11615403130184827

## T-Test TJ vs no TJ WHIP



- ttestWHIP=ttest\_ind(whip,whipNOTJ)
- TtestResult(statistic=1.6424610891642755
- pvalue=0.10511137950195451, df=68.0)

### Conclusions

- How does the performing surgeon in a Tommy John surgery impact the postoperative pitching statistics and outcomes of major league baseball pitchers?
  - No pitching statistics are impacted significantly based on the performing surgeon. However, there is a relationship between recovery time and the performing surgeon. Statistically, we would recommend Dr. Neal for the shortest recovery
- Is there a significant difference between the statistics of pitchers who have had Tommy John surgery and those who haven't?
  - It appears that the overall statistical performance of the ERA and WHIP metrics is not statistically significant between Tommy John and non-Tommy John pitchers. However, it is interesting that pitchers post-Tommy John appear to be improving in performance more than pitchers without the surgery.

### Future Ideas

- ► Look at how Tommy John surgery affects different types of pitchers (reliever, closer, starter)
- Look at if there are any predictors for someone to be more prone to UCL injury (innings pitched, velocity, months taken off after the season)
- Potentially create a model to predict statistical output after Tommy John Surgery

Questions ???

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