



# POLICE VIOLENCE AND BLM MOVEMENT ON TWITTER



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# 01

## BACKGROUND

The “why” question

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# BACKGROUND

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- From 2020 to 2021, we saw a large increase in activism related to the Black Lives Matter movement, both online as well as in the real world. This was following the tragic deaths of many Black Americans at the hands of the police, including George Floyd, Breonna Taylor, Ahmaud Arbery and several others.
- Many asked whether this activism was having any effect on the police system at all, and whether the police were holding themselves to a higher standard as a result of the conversation being had.
- We set our sights on examining Twitter posts (Tweets) specifically, to see if the sentiments shared by people online (representing the real conversations being had around race and policing) were having any effect on the outcome of police encounters in the real world



**HAS THERE BEEN A MEASURABLE DECREASE  
IN RATE OF THE DEATHS OF PEOPLE OF  
COLOR, AND PARTICULARLY BLACK  
AMERICANS, AT THE HANDS OF THE POLICE  
WHICH CAN BE CORRELATED WITH TWITTER  
SENTIMENT SURROUNDING THE BLM  
MOVEMENT?**



# ≡ HYPOTHESIS

## *Null Hypothesis*

- There is no significant decrease in rates of the deaths of people of color at the hands of police.
- 

## *Alternative Hypothesis*

- There is a significant decrease in the rates of deaths of people of color at the hands of police.



# ETHICS

## Potential inaccuracy :

- Data may be bias toward those willing to post their opinions on Twitter
  - Issues with data may have skewed results
- 

## Privacy:

- All data used are public
- Any personal info is removed from the report



# 02

## DATA

Mapping police violence and  
BLM Tweets

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# DATA



## Police Data

- Cases of police brutality resulting in death collected from 2013 to 2022 across the US
- Variables: cause of death, race, date, officer charged, armed
- 10K+ observations

## Twitter Data

- Collection of tweets from the beginning of the BLM movement, 2013 - 2021
- Explored date of tweets in connection to BLM hashtags
- 127+ million observations

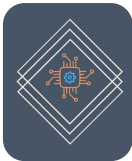




# POLICE DATA



	cause_of_death	race	date	officer_charged	armed
1602	Gunshot	White	2020-09-30	Not Charged	Allegedly Armed
1603	Gunshot	Black	2020-09-30	Not Charged	Unclear
1604	Gunshot	Unknown race	2020-09-29	Not Charged	Allegedly Armed
1605	Gunshot	Black	2020-09-28	Not Charged	Allegedly Armed
1606	Gunshot	Hispanic	2020-09-28	Not Charged	Allegedly Armed



# OVERVIEW OF UNIQUE VALUES IN POLICE DATA

Column cause\_of\_death's unique values:

```
['Gunshot' 'Gunshot,Taser' 'Taser' 'Vehicle' 'Physical Restraint'  
 'Chemical Agent' 'Beaten' 'Other' 'Bean bag' 'Pepper Spray' 'Asphyxiated'  
 'Bomb' 'Taser,Physical Restraint' nan 'Gunshot,Vehicle']
```

Column race's unique values:

```
['White' 'Black' 'Unknown race' 'Hispanic' 'Asian' 'Native American' nan  
 'Pacific Islander']
```

Column date's unique values:

```
['2020-09-30T00:00:00.000000000' '2020-09-29T00:00:00.000000000'  
 '2020-09-28T00:00:00.000000000' ... '2013-01-03T00:00:00.000000000'  
 '2013-01-02T00:00:00.000000000' '2013-01-01T00:00:00.000000000']
```

Column officer\_charged's unique values:

```
['Not Charged' 'Charged' 'Unknown']
```

Column armed's unique values:

```
['Allegedly Armed' 'Unclear' 'Unarmed/Did Not Have Actual Weapon'  
 'Vehicle' nan]
```





# TWEET LABELS



**BlackLivesMatter**



**AllLivesMatter**



**BlueLivesMatter**



# 03 PROCESS

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# TIMELINE



Cleaning Police Data

01

02

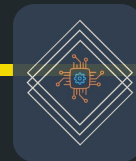
Sampling Tweet IDs

Hydrating Tweets

03

04

EDA



# ≡ POLICE

- Extract columns we need (cause of death, race, date, officer charged, arm status)
- Clean the data type in each column
  - Date into Datetime
- Remove data after December 31, 2021
  - Because we do not have corresponding tweet data at that time
- 8473 records in the end





# ≡ TWEETS

- Merge yearly CSVs into one
- Randomly select 2,500,000 tweets to hydrate using *Hydrator*
  - 1,599,186 tweets hydrated; 29% deleted
  - Took approximately 2 days to hydrate
- Hydrated tweets 13GB json file
  - Select IDs and date
- Merge IDs and date onto sample with labels
- Labels originally one-hot encoded
  - Change back to label encoding for easier graphing
- 32,524 tweets in the end





	<b>status_id</b>	<b>time</b>
0	1298955271587405800	2020-08-27T12:07:24+00:00
1	1279794731866894300	2020-07-05T15:10:16+00:00
2	1272116933740114000	2020-06-14T10:41:26+00:00
3	1269704052532789200	2020-06-07T18:53:30+00:00
4	1350089310587003000	2021-01-15T14:35:49+00:00

	<b>status_id</b>	<b>time</b>	<b>blm</b>	<b>alm</b>	<b>blulm</b>	<b>File_Name</b>	<b>label</b>
0	537684004657725440	2014-11-26	1.0	0.0	0.0	2014-11.csv	blm
1	537684004657725440	2014-11-26	1.0	0.0	0.0	2014-11.csv	blm
2	1269868146640384000	2020-06-08	1.0	0.0	0.0	2020-06.csv	blm
3	1269868146640384000	2020-06-08	1.0	0.0	0.0	2020-06.csv	blm
4	930503267662131200	2017-11-14	1.0	0.0	0.0	2017-11.csv	blm



# 04

## EDA

Graphing to  
see trends

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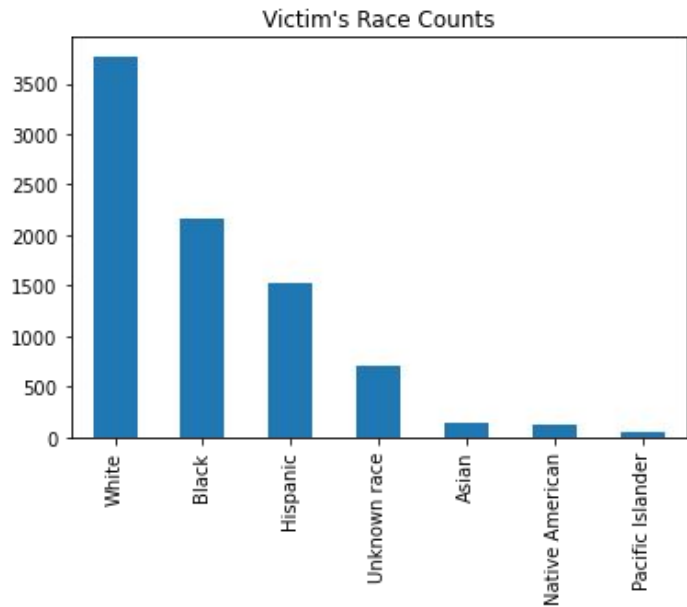
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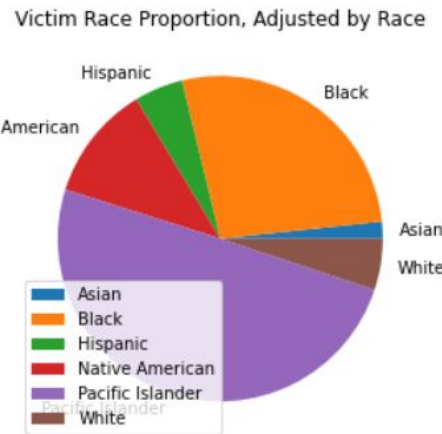
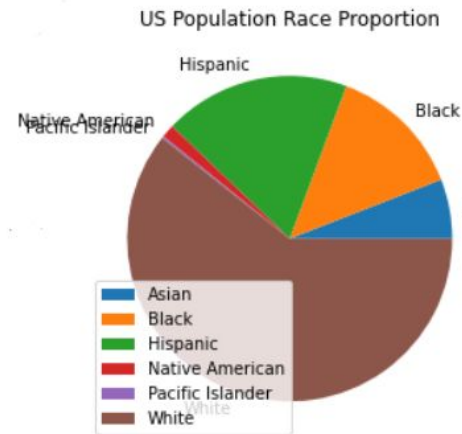




# RACE VARIABLE EDA



Victim race counts

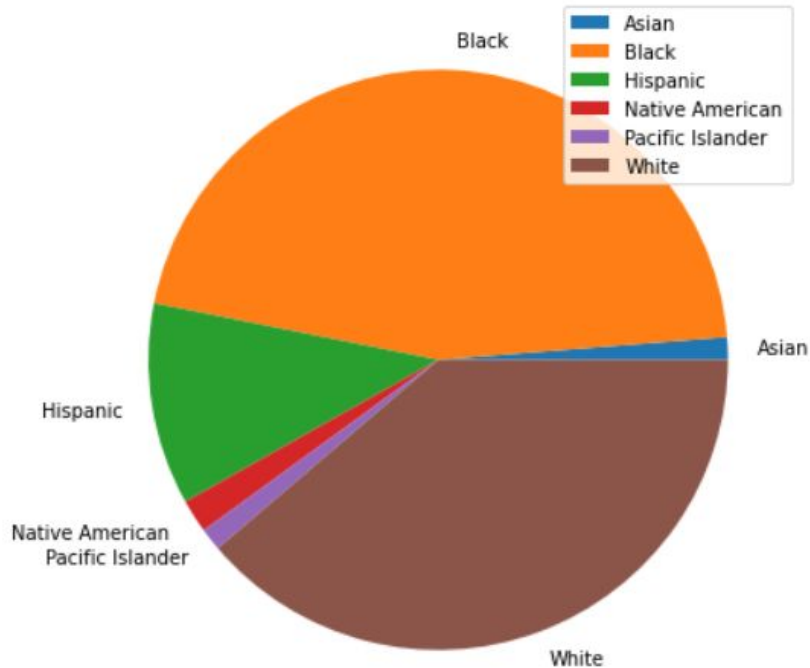


Victim race proportions compared to US Census data

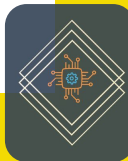
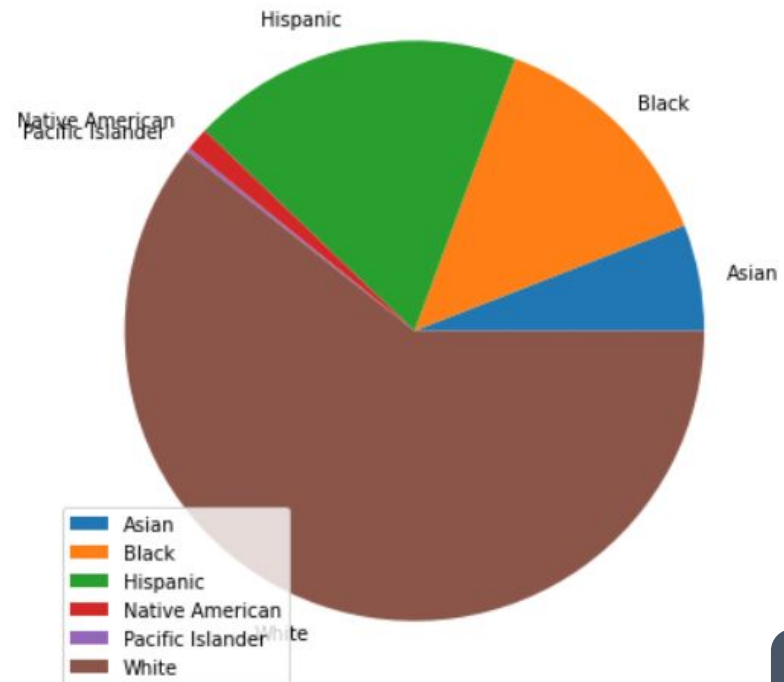


# RACE DISTRIBUTION OF VICTIMS WHEN OFFICERS WERE CHARGED

Victim Race Proportion, when Officers were Charged



US Population Race Proportion





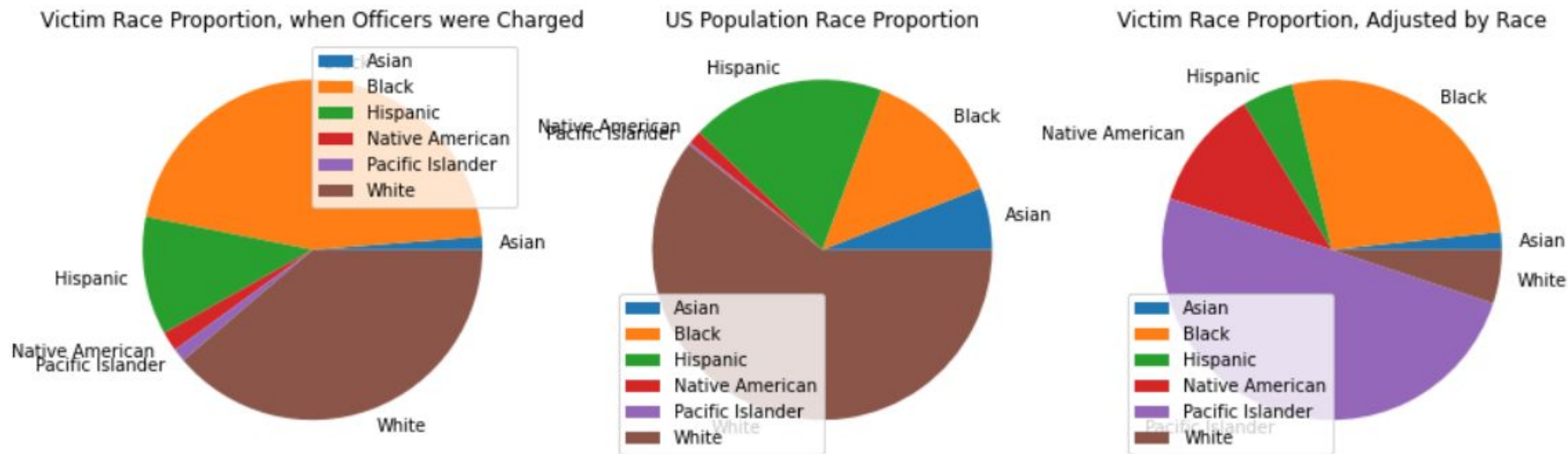
## RACE DISTRIBUTION OF VICTIMS WHEN OFFICERS WERE CHARGED



- Large disparity between the racial makeup of the US and the racial makeup of police killing victims.
- People of color, especially Black Americans, make up nearly half of all deaths despite making up under one-seventh of the US population.
- Despite White Americans making up over 60% of the population, they account for under 40% of the deaths in police encounters.



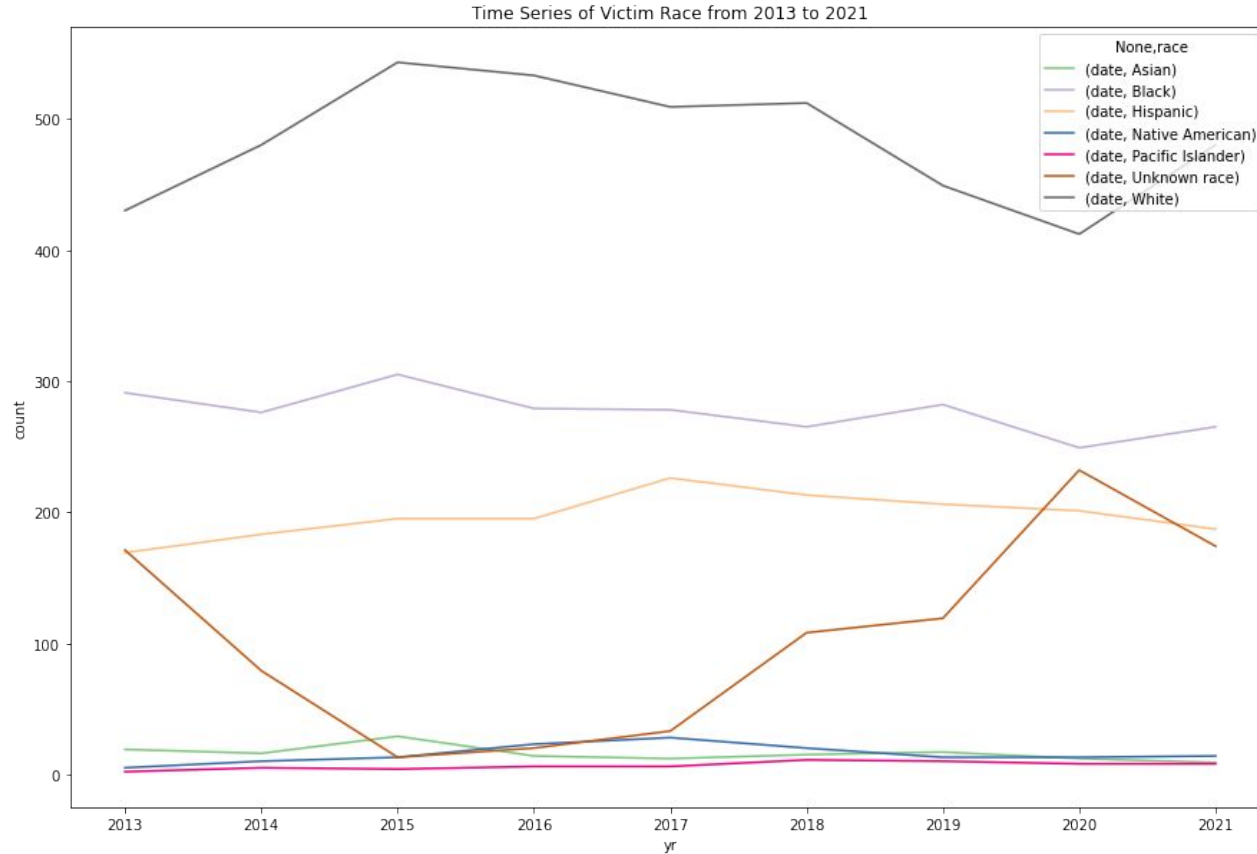
# ADJUSTING RACE DISTRIBUTION FOR US POPULATION DIFFERENCES



As we can see, the population-adjusted chart on the right shows the true proportion of the races of the victims, and they are far from uniform. Pacific Islanders, Black people and Native Americans are the most likely to be victims, while White and Asian Americans are the least likely.



# VICTIM'S RACE TIME SERIES



- Spans 2013-2021
- Rise in unknown race cases simultaneously suggest decreases could partially be related to a lack of victim race reporting



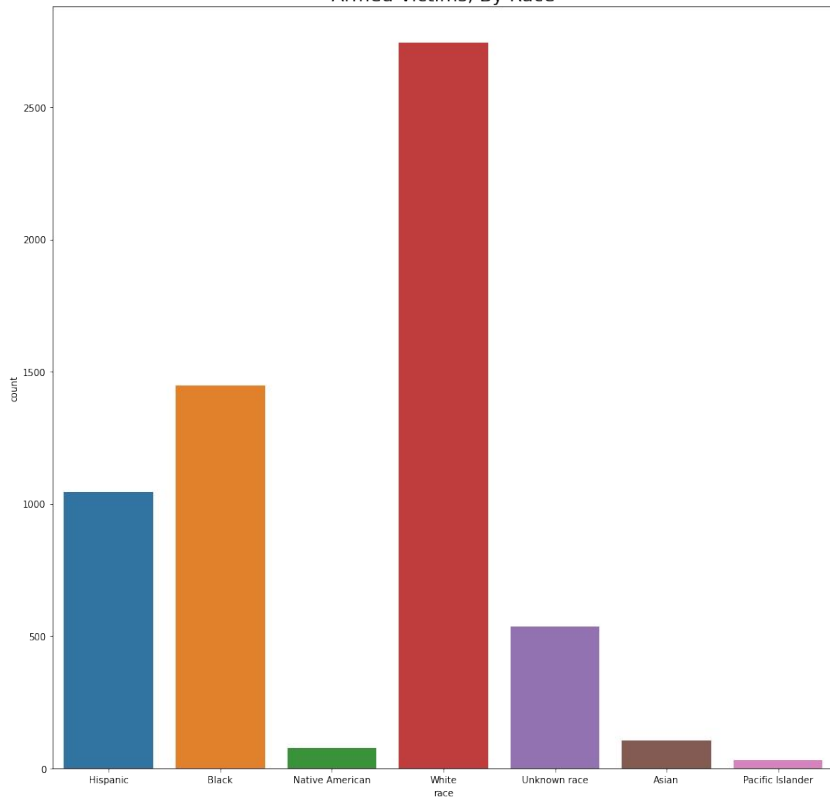




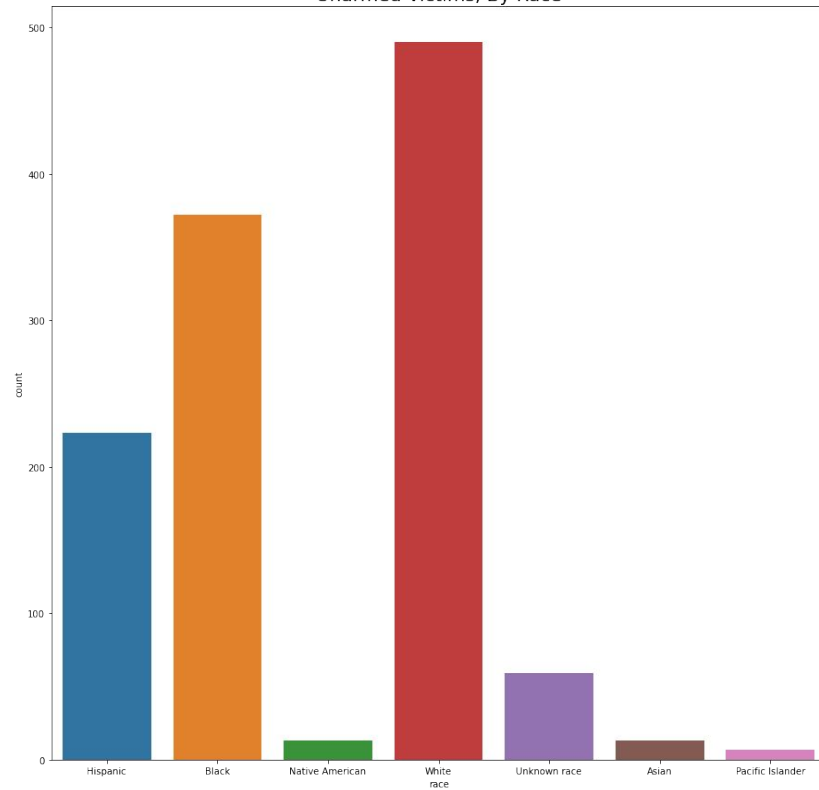
# RACE V.S. ARM STATUS



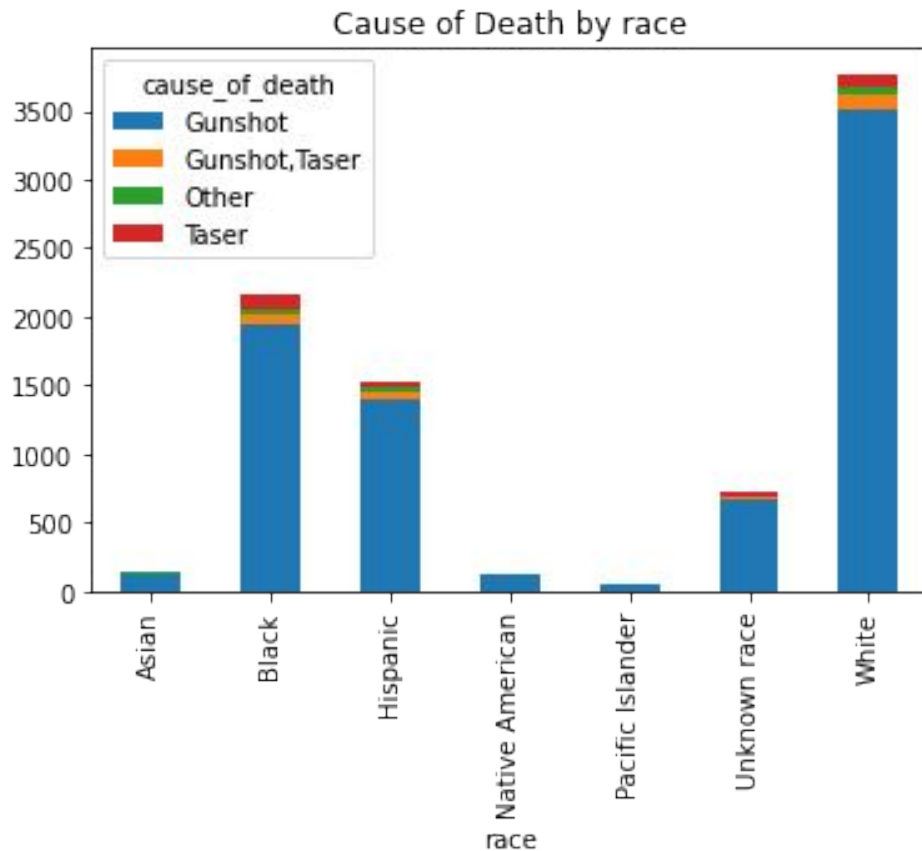
Armed Victims, By Race



Unarmed Victims, By Race



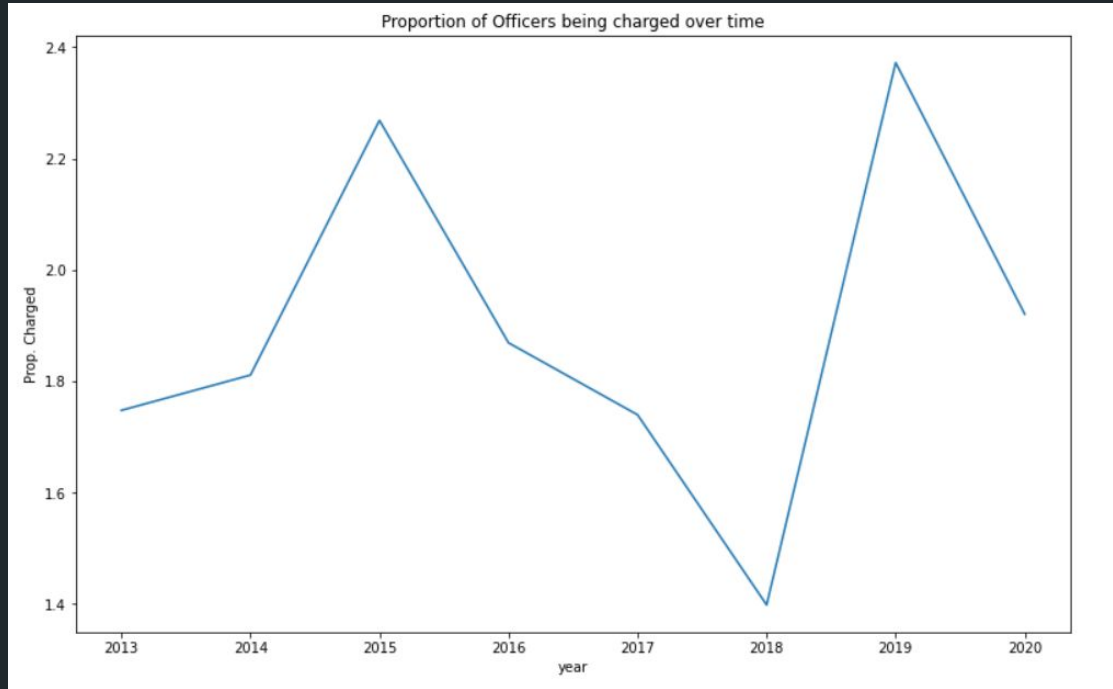
# CAUSE OF DEATH



- On average, over 92% of death were caused by Gunshot, regardless of race.



# RATE OF OFFICERS BEING CHARGED

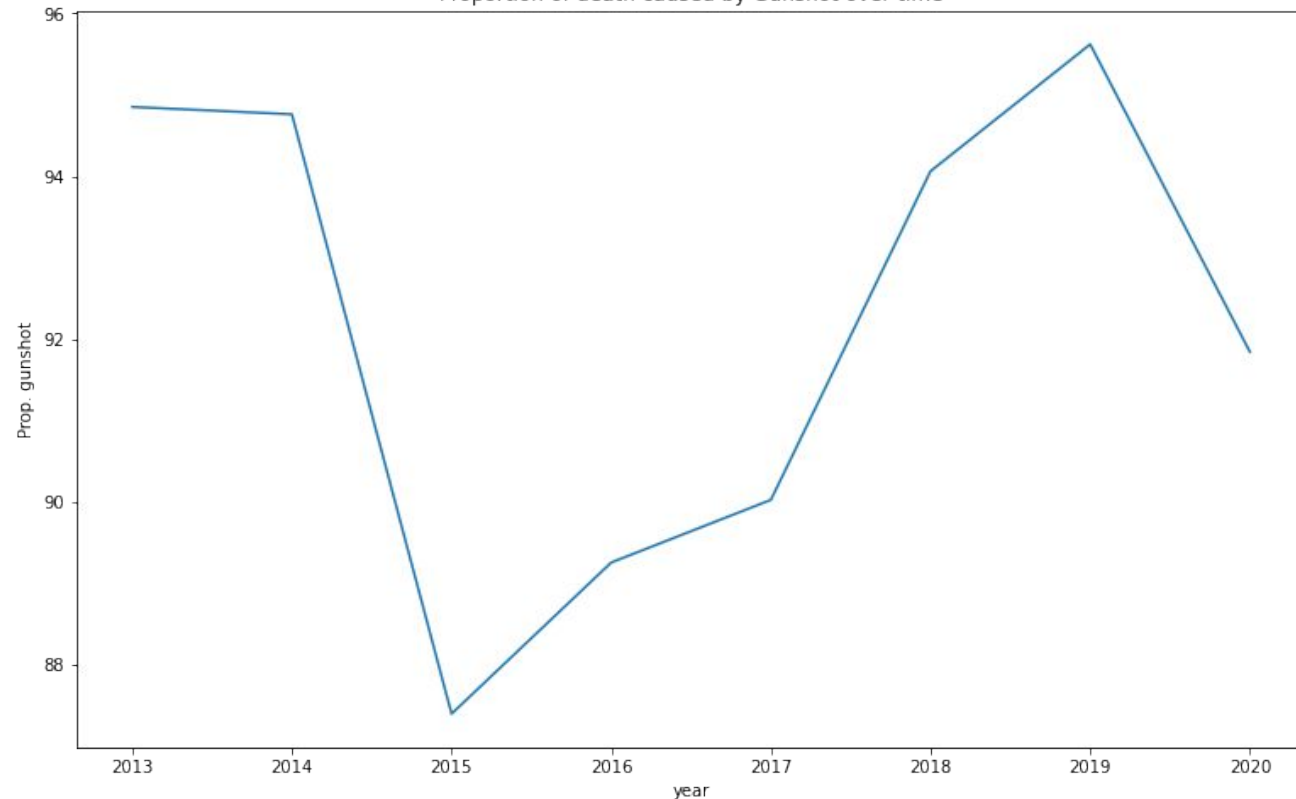


No clear trend.



# RATE OF DEATH CAUSED BY GUNSHOT

Proportion of death caused by Gunshot over time



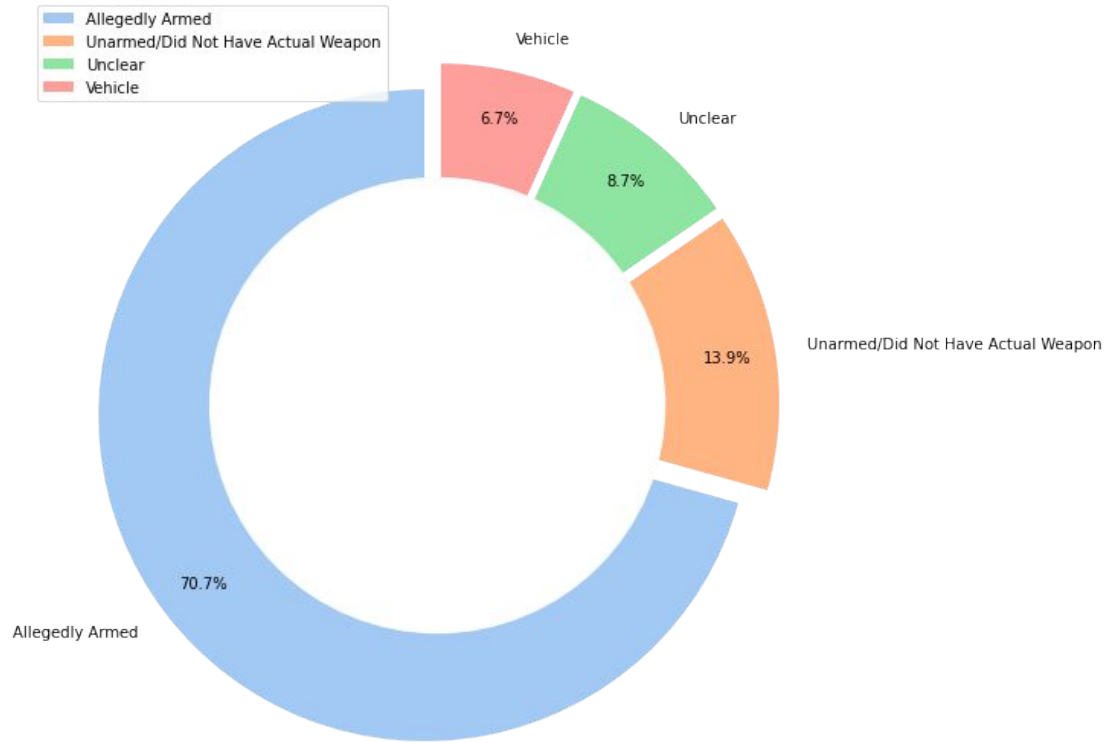
- Dropped dramatically in 2014, and slowly rise to its peak in 4 years from 2015 to 2019.
- The trend of death proportion caused by Gunshot shows negative relationship with the rate of officer being charged.





# VICTIMS' ARM STATUS

## Proportion of victims' arm status

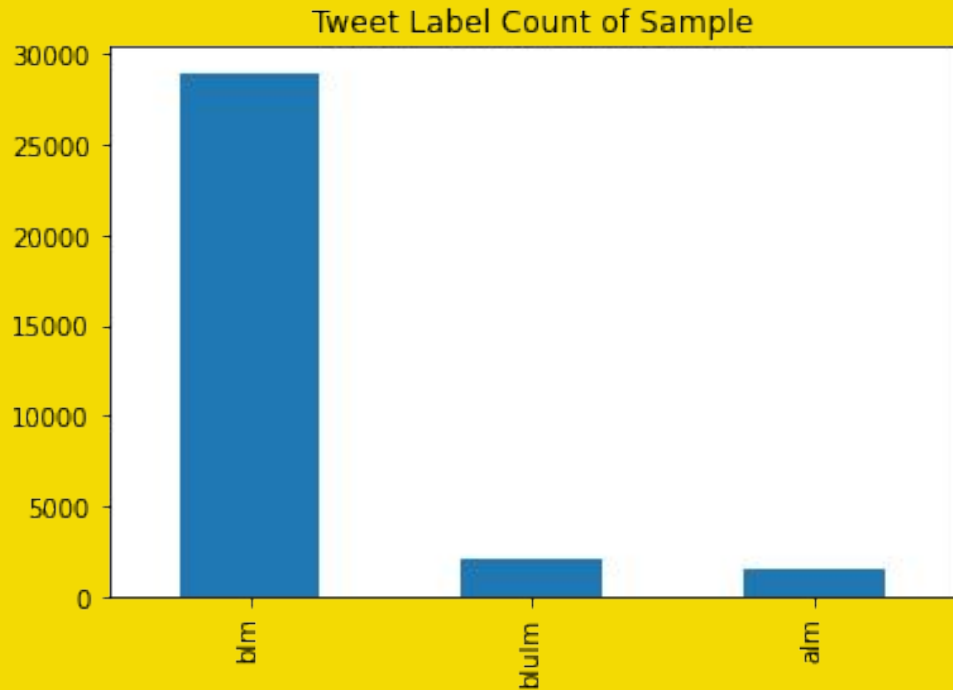
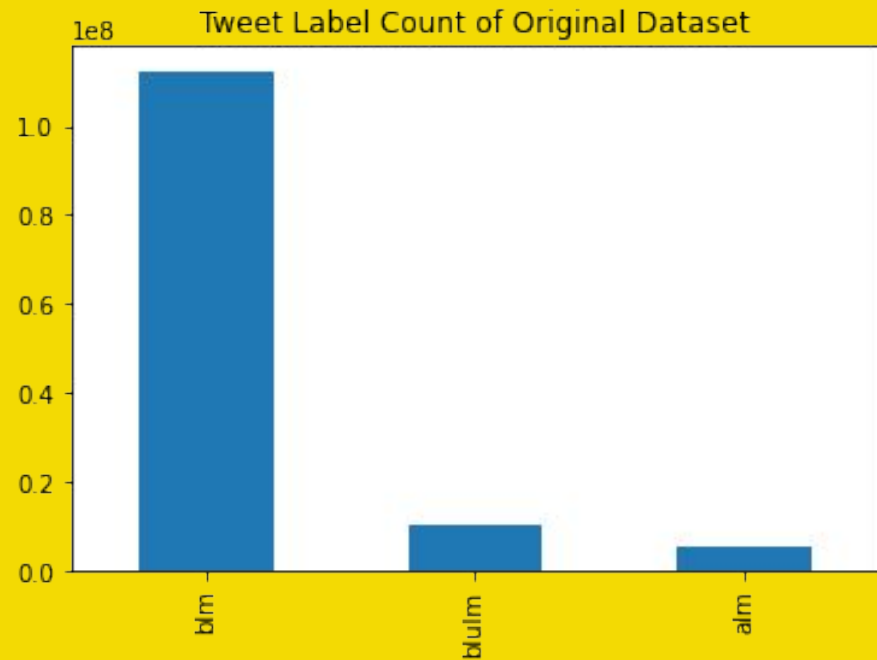


- Armed victims are dominating, which makes up of ~71% victims.

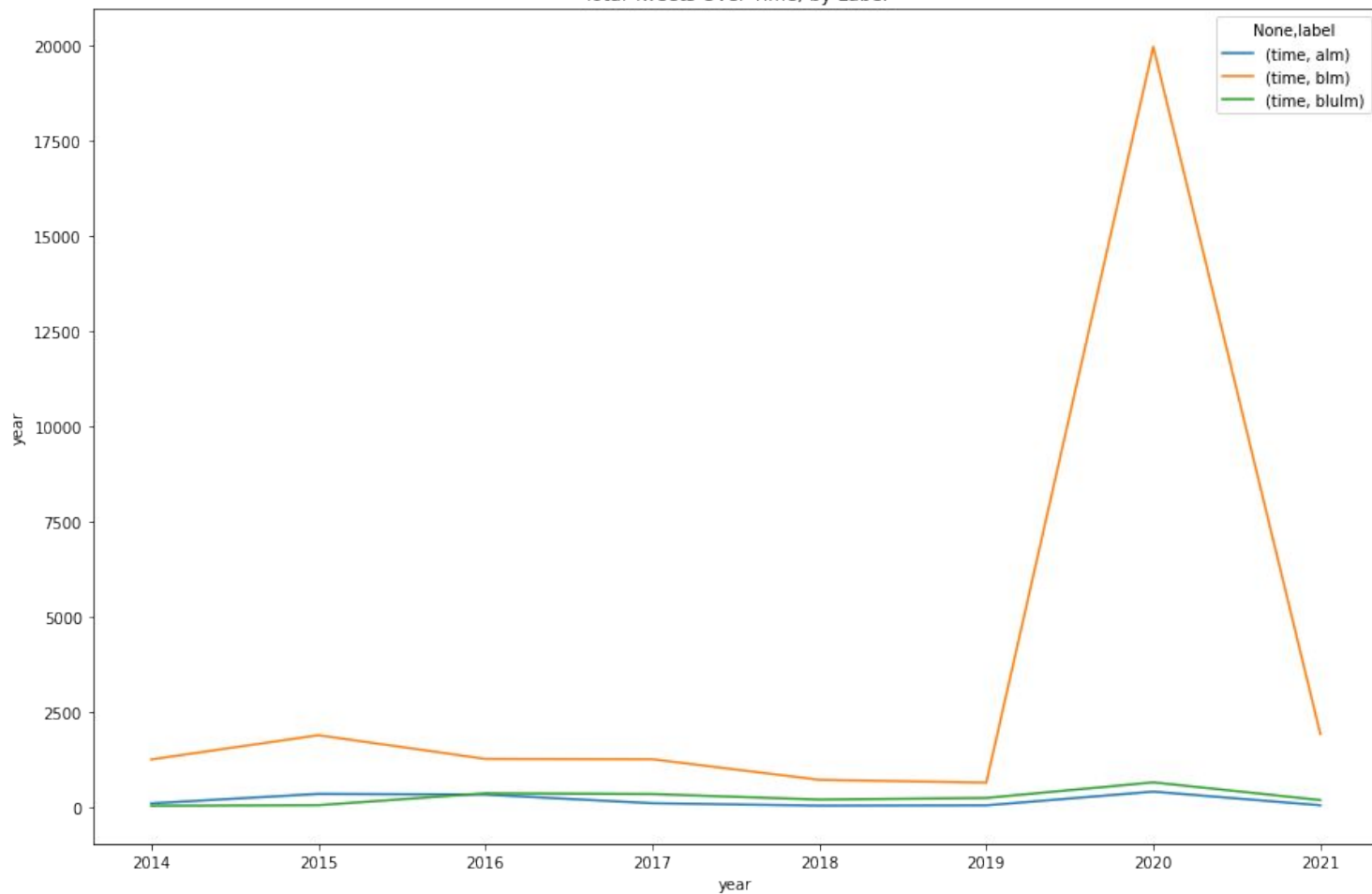




# TWEET LABEL COUNTS

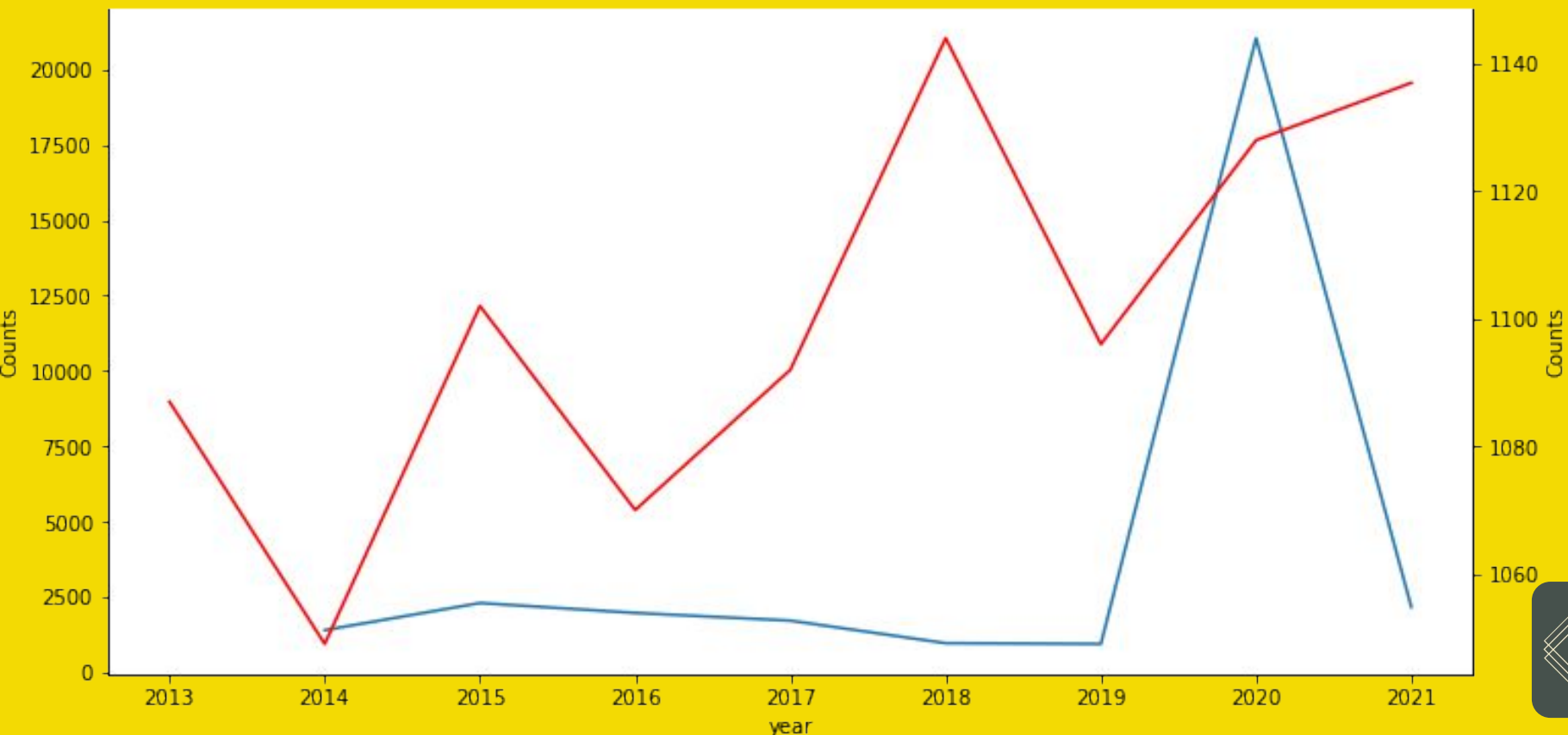


Total Tweets Over Time, by Label





Counts of Death vs Tweets over Time





# CORRELATION

0.378

Pearson's

0.277

Kendall's

0.399

Spearman's





**CONCLUSION**

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# MAIN TAKEAWAYS



**Gunshot lead to most death**

**Black Americans more likely to be killed than white Americans by police**

**Tweets are difficult to work with**

**Fail to reject full; no decrease in deaths by police**



# FUTURE IMPROVEMENTS & ADDITIONS

**Look into more variables that could be potential confounds**

Police funding, etc.

**Use all of tweets instead of sample**

**Label tweets ourselves**



# GitHub



# THANKS

Do you have any questions?

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