



# Chapel Hill Police District:

## Is a racial biased justice observable between 2010 and 2019?

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# #Introduction - Interest






The **cliché rift between black/white people** is said observable through a difference in judicial treatment between races. But what are the characteristics explaining this denounced racial bias in the American judicial system?



## Data from Chapel Hill Open Data



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	Observations	Variables
 Arrestees	82'525	12
 Victims	16'040	18
 Police Employees	174	7

## Glimpse on the arrestees dataset




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Date of Arrest	Arrestee's Age	Arrestee's Race	Arrestee's Gender	Type of Arrest	Drugs or Alcohol Present	Weapon Present
2010-01-01	18	White	Male	SUMMONED/CITED	Yes	UNARMED
2010-01-01	18	White	Male	TAKEN INTO CUSTODY	No	UNARMED
2010-01-01	27	White	Male	ON VIEW	Unknown	UNARMED
2010-01-05	55	White	Male	SUMMONED/CITED	Yes	UNARMED
2010-01-06	23	White	Male	ON VIEW	Yes	UNARMED

## Glimpse on the victims dataset

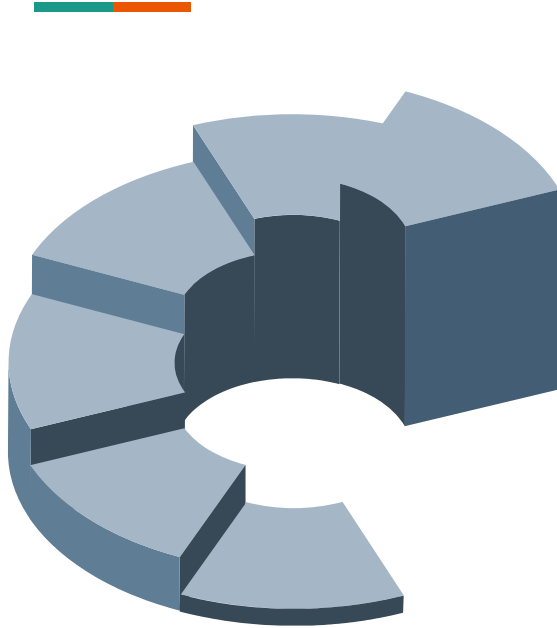
Date of Report	Date of Occurrence	Reported As	Victim Age	Victim Race	Victim Gender
2019-06-29	2019-06-29	THEFT/LARCENY	21	White	Female
2019-06-29	2019-06-29	ADMINISTRATIVE	73	White	Male
2019-06-29	2019-06-29	DISTURBANCE/NUI	29	Black	Female
2019-06-29	2019-06-29	DAMAGE/VANDALIS	25	White	Male
2019-06-28	2019-06-28	TRAFFIC/TRANSP	78	White	Male

## Glimpse on the police employees dataset



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Job Class Description	Age	Service Years	Gender	Race
POLICE OFFICER IV	50	10	Male	White
POLICE OFFICER IV	46	13	Male	White
POLICE OFFICER IV	41	15	Male	White
RELIEF PARKING LOT ATT	22	2	Female	Black
RELIEF PARKING LOT ATT	38	1	Female	Black



Analyse the distribution of victims and arrestees profiles to check the representativeness of our samples

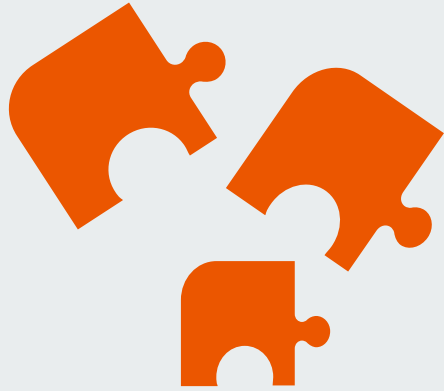
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Sketch the typical profile of a victim and of an arrestee to check if race has an impact

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# #Exploration analysis

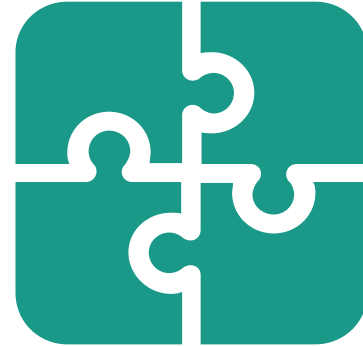


## Samples population

70-30 of white / black victims  
50-50 of white / black arrestees

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Our samples are  
not representative  
but why...?

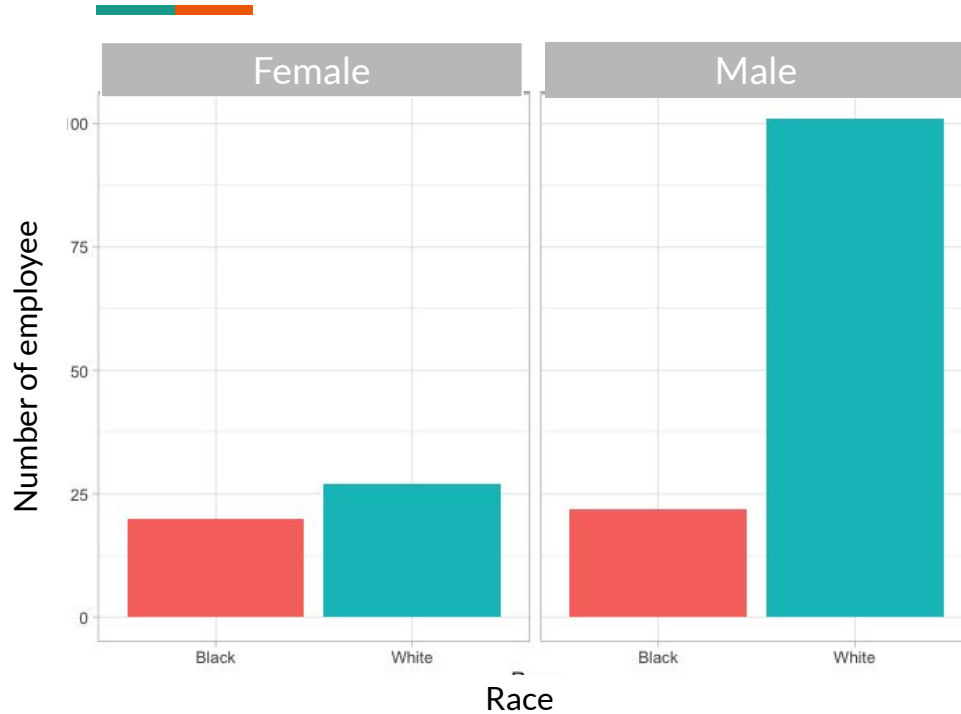


## Population of Chapel Hill

~ 90% of white citizens  
~ 10% of black citizens

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# #Exploration analysis - Employee demographics



Clear majority of males  
among positions in the  
Chapel Hill Police District

Employees' race

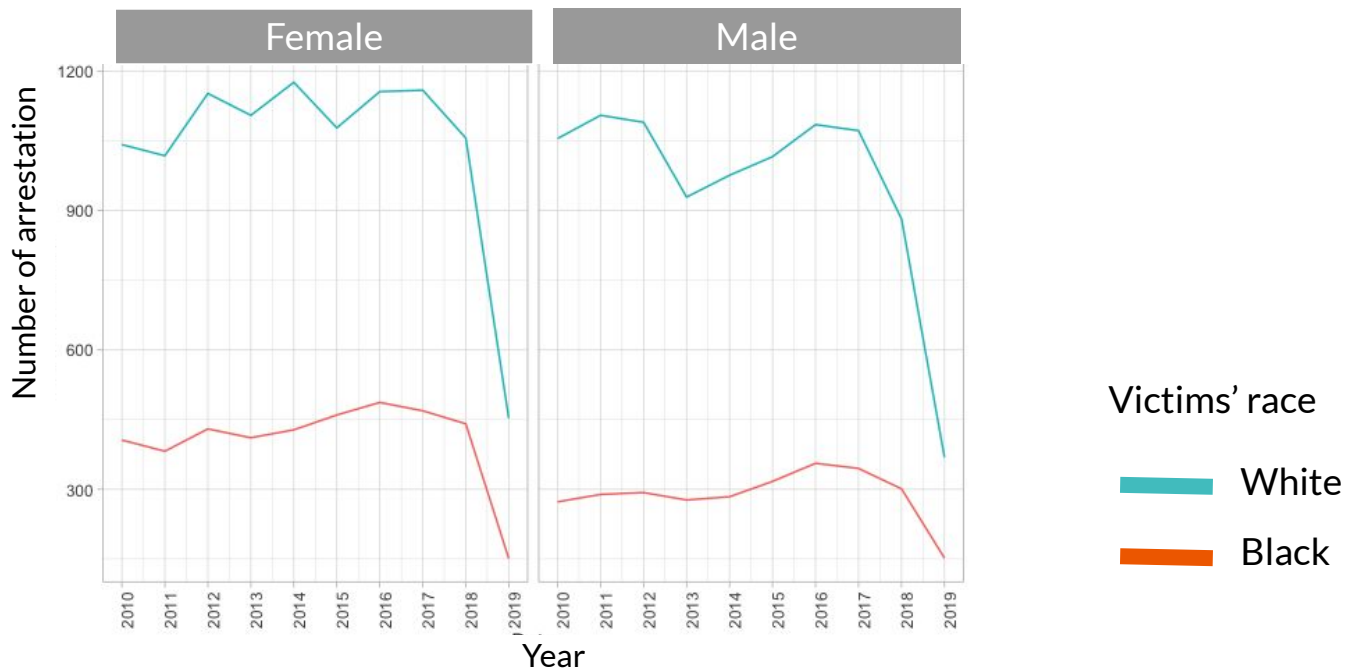
White

Black

# #Exploration analysis - Victims

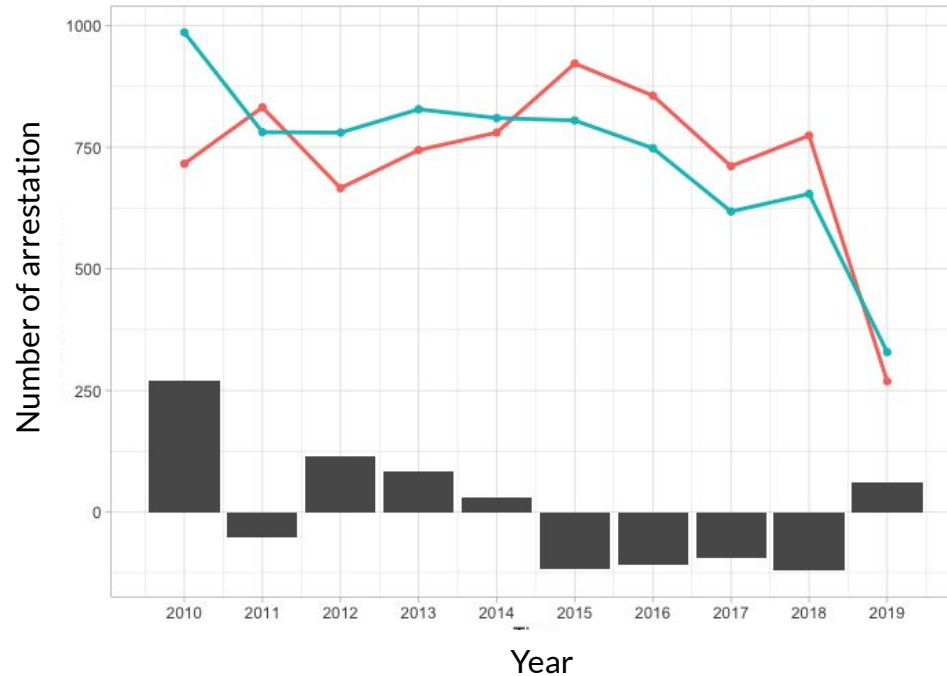
White victims outnumber black ones.

This dataset better represents the real Chapel Hill's race distribution



# #Exploration analysis - Arrestees

Higher number of arrestation for black people since 2015



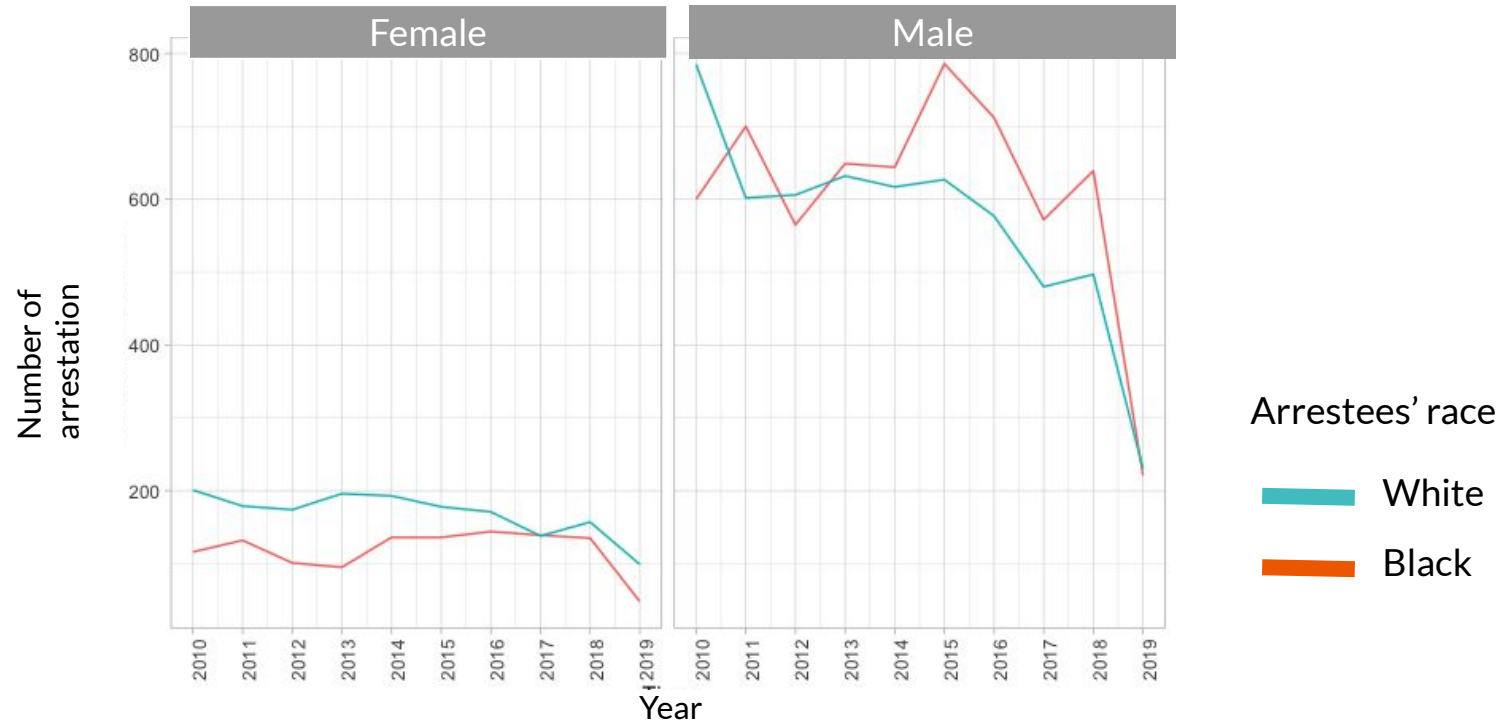
Arrestees' race

White

Black

# #Exploration analysis - Arrestees

Men outnumber women in the arrestees sample



# #Exploration analysis - Arrestees



Black people are systematically more taken into custody than white people.

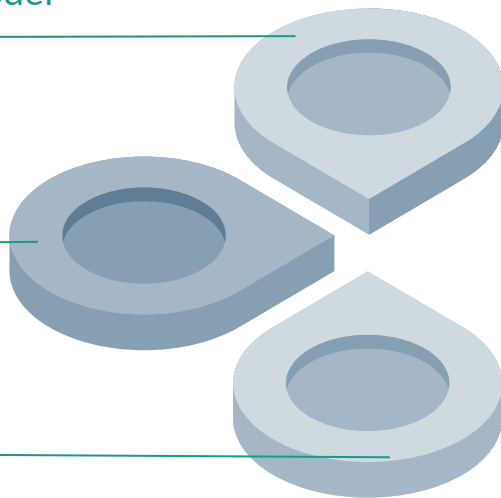
# #Modeling - The modeling process

— To quantitatively assess our exploratory analysis...

Gaussian linear model

Poisson  
linear model

Random Forest



**Linear models** to check the representativeness of our samples

**Classification method** to check the racial bias in arrestee profile

# #Conclusion

## Results



Non significance for race in linear models due to the biased dataset (black overrepresented)

A satisfying classification, even if fixing the race distribution bias would have steer better generalizing results

## Learning points



Always contextualize datasets to confirm its representativeness

Manage the different time of data collection when working with many datasets

Let the figures do the talking, not extrapolate on unverified assumptions