# FINAL PROJECT

의민정, 김윤전, 박대한, 이재상, 최은성

## CONTENTS

- · ROLES & RESPONSIBILITIES
- · PURPOSE OF THE PROJECT
- · DESCRIPTION OF THE DATA
- · EDA
- · FURTHER PLANS

### ROLES & RESPONSIBILITIES

- 김민정(조장)
  - 발표 자료 준비 및 발표
- 김윤전
  - \_ 인터랙티브 시각화
- 박대한
  - EDA
- 이재상
  - EDA
- 최은성
  - \_ 발표 자료 준비

# PURPOSE OF THE PROJECT



- 베이지안 선형회귀 모형으로 미국 범죄지도를 완성하자
- 인터랙티브 시각화: Missing value가 채워진 모드의 지역별 시각화와 없는 모드의 시각화를 그리자
- •베이지안 method를 사용하여 결측치를 채우자

### • 2215 observations and 147 variables in crimedata.csv

```
'data.frame':
               2215 obs. of 147 variables:
                       : Factor w/ 2018 levels "Aberdeencity",..: 151 1035 1781 665 141 1700 1272 41 566 1860 ...
$ communityname
                       : Factor w/ 48 levels "AK", "AL", "AR", ...: 29 36 35 32 23 24 19 15 27 41 ...
$ state
                       : Factor w/ 115 levels "?","1","101",...: 57 60 1 55 84 1 46 1 40 1 ...
$ countyCode
                       : Factor w/ 960 levels "?","100","1000",...: 511 426 1 215 473 1 468 1 177 1 ...
$ communityCode
$ fold
                        : int 1111111111...
$ population
                        : int 11980 23123 29344 16656 11245 140494 28700 59459 74111 103590 ...
$ householdsize
                        : num 3.1 2.82 2.43 2.4 2.76 2.45 2.6 2.45 2.46 2.62 ...
$ racepctblack
                        : num 1.37 0.8 0.74 1.7 0.53 ...
$ racePctWhite
                        : num 91.8 95.6 94.3 97.3 89.2 ...
$ racePctAsian
                       : num 6.5 3.44 3.43 0.5 1.17 0.9 1.47 0.4 1.25 0.92 ...
$ racePctHisp
                        : num 1.88 0.85 2.35 0.7 0.52 ...
$ agePct12t21
                        : num 12.5 11 11.4 12.6 24.5 ...
$ agePct12t29
                       : num 21.4 21.3 25.9 25.2 40.5 ...
$ agePct16t24
                        : num 10.9 10.5 11 12.2 28.7 ...
$ agePct65up
                        : num 11.3 17.2 10.3 17.6 12.6 ...
$ numb∪rban
                        : int 11980 23123 29344 0 0 140494 28700 59449 74115 103590 ...
$ pctUrban
                        : num 100 100 100 0 0 100 100 100 100 100 ...
$ medIncome
                        : int 75122 47917 35669 20580 17390 21577 42805 23221 25326 17852 ...
                        : num 89.2 79 82 68.2 69.3 ...
$ pctWWage
$ pctWFarmSelf
                        : num 1.55 1.11 1.15 0.24 0.55 1 0.39 0.67 2.93 0.86 ...
                        : num 70.2 64.1 55.7 39 42.8 ...
$ pctWInvInc
$ pctWSocSec
                        : num 23.6 35.5 22.2 39.5 32.2 ...
$ pctWPubAsst
                        : num 1.03 2.75 2.94 11.71 11.21 ...
$ pctWRetire
                        : num 18.4 22.9 14.6 18.3 14.4 ...
$ medFamInc
                        : int 79584 55323 42112 26501 24018 27705 50394 28901 34269 24058 ...
$ perCapInc
                        : int 29711 20148 16946 10810 8483 11878 18193 12161 13554 10195 ...
$ whitePerCap
                        : int 30233 20191 17103 10909 9009 12029 18276 12599 13727 12126 ...
$ blackPerCap
                        : int 13600 18137 16644 9984 887 7382 17342 9820 8852 5715 ...
$ indianPerCap
                        : int 5725 0 21606 4941 4425 10264 21482 6634 5344 11313 ...
                        : int 27101 20074 15528 3541 3352 10753 12639 8802 8011 5770 ...
$ AsianPerCap
                       : Factor w/ 1918 levels "?", "0", "10000",...: 1022 1049 1174 717 784 1418 681 1460 1068 1445 ....
$ OtherPerCap
```

### • Major Variable description (1/3)

#### <지역코드>

communityname state countyCode communityCode

#### ~총인구수>

population

#### <u><가구별크기></u>

householdsize

#### <u> <인종비율></u>

racepctblack racePctWhite racePctAsian racePctHisp

#### <u> <연령대></u>

agePct12t21 agePct12t29 agePct16t24 agePct65up

#### <도시확>

numbUrban pctUrban

#### <u><소득></u>

medIncome
pctWWage
pctWFarmSelf
pctWInvInc
pctWSocSec
pctWPubAsst
pctWRetire
medFamInc
perCapInc

#### <u> <인종별 소득></u>

whitePerCap blackPerCap indianPerCap AsianPerCap OtherPerCap HispPerCap

#### <u> <빈곤></u>

NumUnderPov PctPopUnderPov

### • Major Variable description (2/3)

#### <u><교육></u>

PctLess9thGrade PctNotHSGrad PctBSorMore

#### <u><고융></u>

PctUnemployed PctEmploy

PctEmplManu

PctEmplProfServ

**PctOccupManu** 

PctOccupMgmtProf

#### <u><결혼/가정></u>

MalePctDivorce

MalePctNevMarr

FemalePctDiv

TotalPctDiv

PersPerFam

PctFam2Par

PctKids2Par

PctYoungKids2Par

PctTeen2Par

PctWorkMomYoungKids

**PctWorkMom** 

NumKidsBornNeverMar

PctKidsBornNeverMar

#### <이민자>

Numlmmig

PctlmmigRecent

PctlmmigRec5

PctlmmigRec8

PctlmmigRec10

PctRecentImmig

PctRecImmig5

PctRecImmig8

PctRecImmig10

PctSpeakEnglOnly

PctNotSpeakEnglWell

#### <del><부동산관련></del>

PctLargHouseFam

PctLargHouseOccup

PersPerOccupHous

PersPerOwnOccHous

PersPerRentOccHous

PctPersOwnOccup

PctPersDenseHous

PctHousLess3BR

MedNumBR

HousVacant

PctHousOccup

PctHousOwnOcc

PctVacantBoarded

. . .

### • Major Variable description (3/3)

#### <지역토박이>

PctForeignBorn

PctBornSameState

PctSameHouse85

PctSameCity85

PctSameState85

#### <7|E|>

LemasSwornFT

LemasSwFTPerPop

LemasSwFTFieldOps

LemasSwFTFieldPerPop

LemasTotalReq

LemasTotReqPerPop

#### <u><경찰></u>

PolicReqPerOffic

PolicPerPop

RacialMatchCommPol

**PctPolicWhite** 

PctPolicBlack

PctPolicHisp

**PctPolicAsian** 

PctPolicMinor

OfficAssgnDrugUnits

NumKindsDrugsSeiz

PolicAveOTWorked

#### ~범죄관련>

murders

murdPerPop

rapes

rapesPerPop

robberies

robbbPerPop

assaults

assaultPerPop

burglaries

burglPerPop

larcenies

larcPerPop

autoTheft

autoTheftPerPop

arsons

arsonsPerPop

ViolentCrimesPerPop

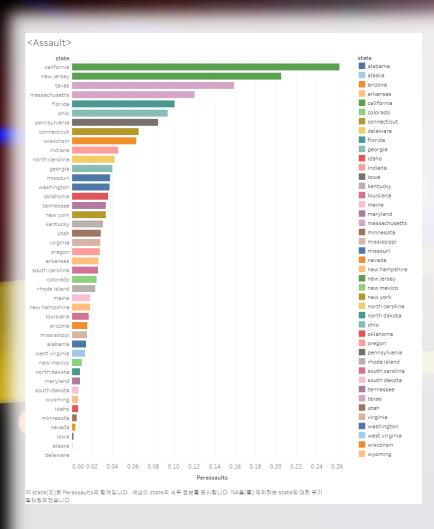
nonViolPerPop

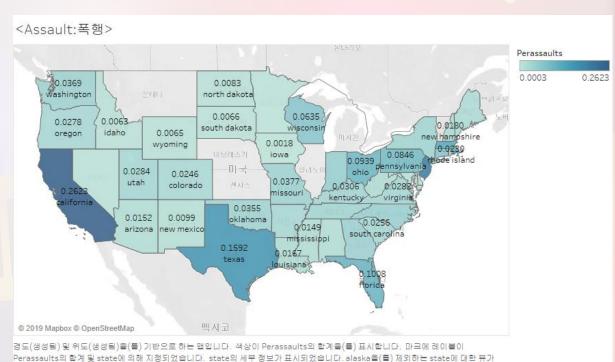
### EDA: MISSING VALUE IMPUTATION

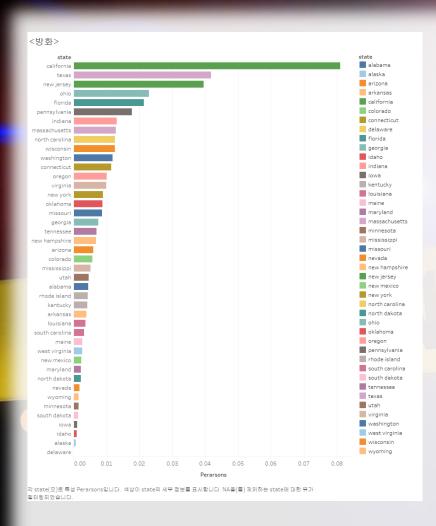
- · Total 2215 observations
- · X 변수 중 22개 변수에서 결측지 약 84%로 확인
  - → 해당 항목의 사람수가 ()이거나
  - → %의 경우는 너무 적어 0으로 나오는 경우도 발생함
  - → 따라서 22개 변수는 삭제함
- 그 외에는 Y 변수임

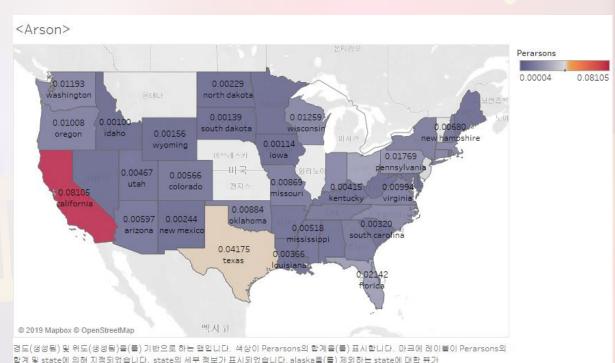
LemasSwornFT	LemasSwFTPerPop	LemasSwFTFieldOps
0.8451467	0.8451467	0.8451467
LemasSwFTFieldPerPop	LemasTotalReq	LemasTotReqPerPop
0.8451467	0.8451467	0.8451467
PolicReqPerOffic	PolicPerPop	RacialMatchCommPol
0.8451467	0.8451467	0.8451467
PctPolicWhite	PctPolicBlack	PctPolicHisp
0.8451467	0.8451467	0.8451467
PctPolicAsian	PctPolicMinor	OfficAssgnDrugUnits
0.8451467	0.8451467	0.8451467
NumKindsDrugsSeiz	PolicAveOTWorked	PolicCars
0.8451467	0.8451467	0.8451467
PolicOperBudg	LemasPctPolicOnPatr	LemasGangUnitDeploy
0.8451467	0.8451467	0.8451467
PolicBudgPerPop		
0.0454467		

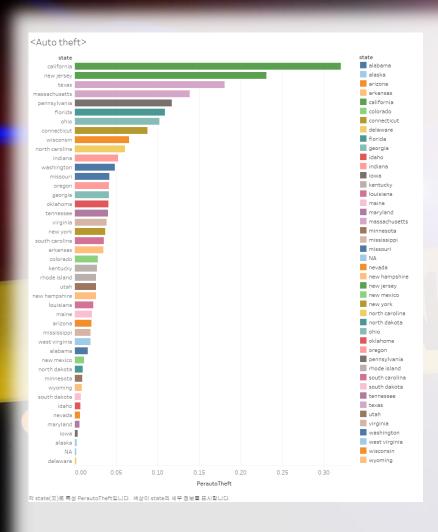
0.8451467

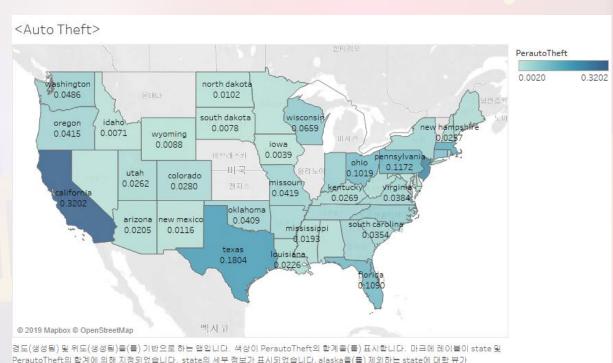


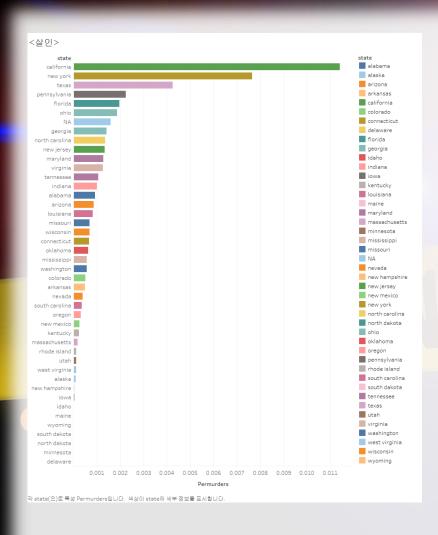


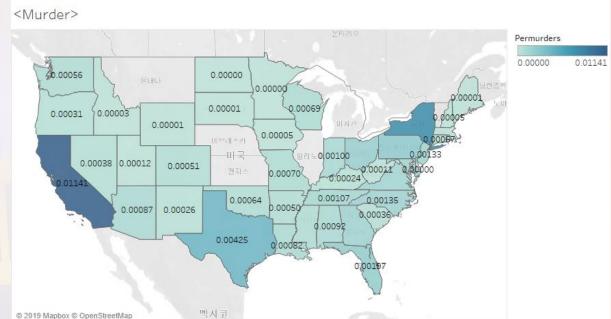






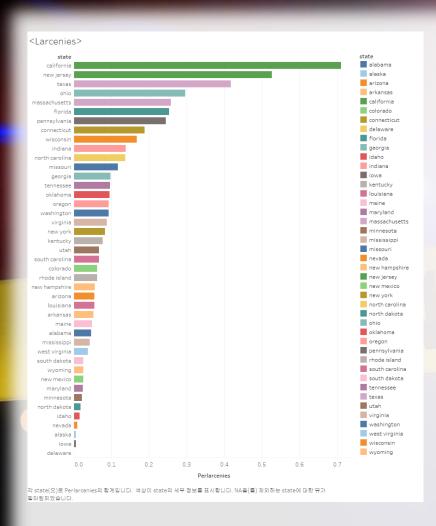


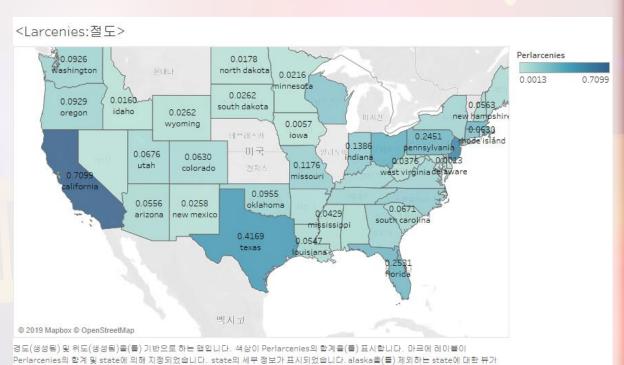


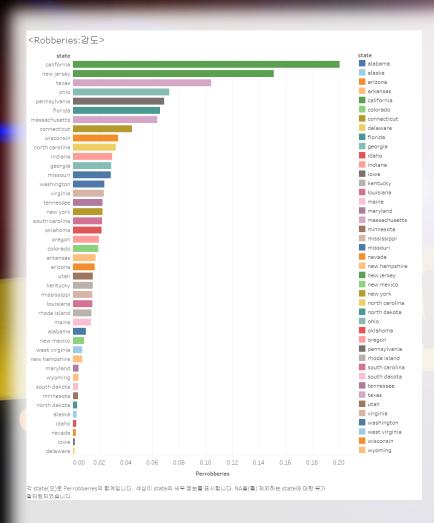


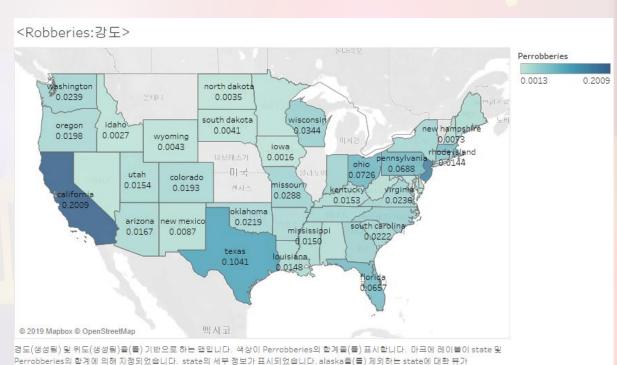
경도(생성됨) 및 위도(생성됨)을(를) 기반으로 하는 맵입니다. 색상이 Permurders의 합계을(를) 표시합니다. 마크에 레이블이

Permurders의 합계에 의해 지정되었습니다. state의 세부 정보가 표시되었습니다. alaska을(를) 제외하는 state에 대한 뷰가

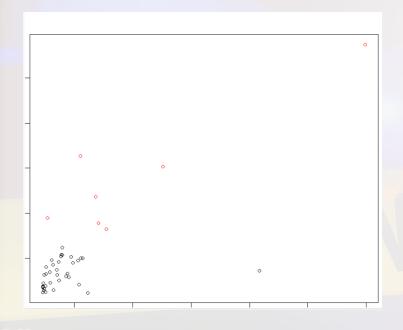








### EDA



- · K-means clustering을 통해 오른쪽과 같이 결과가 도출됨
- · 시각확률 통해 범죄율이 높게 나온 도시들의 결과가 비슷함

```
# A tibble: 2 x 9
 kmeans Permurders Perrapes Perrobberies Perassaults Perburglaries Perlarcenies PerautoTheft Perarsons
                       <db1>
   <db1>
              <db1>
                                    <db1>
                                                <db1>
                                                               <db1>
                                                                            <db1>
                                                                                         <db1>
                                                                                                   <db1>
           0.000710 0.00577
                                   0.0152
                                               0.0221
                                                             0.0358
                                                                           0.0625
                                                                                        0.0270
                                                                                                 0.00531
           0.00332
                     0.0344
                                   0.104
                                               0.147
                                                              0.243
                                                                           0.386
                                                                                        0.171
                                                                                                 0.0339
```

# FURTHER PLANS

- •다른 추가 정보를 이용한 변수 생성
- 파생변수 추가 고려:
  - State별 뿐 아니라 county별 분석
- •사용할 모델
  - Linear regression
  - Generalized linear model
  - Hierarchical model 5

# THANK YOU POLICE INE DO