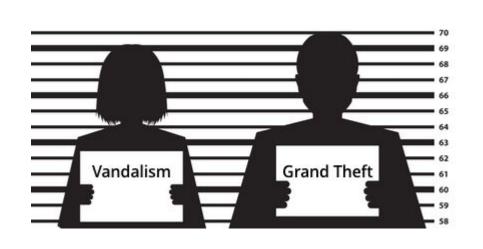
Projeto LabRedes de Conhecimento

San Francisco Crime Classification







Sobre o tema

De 1934 a 1963, San Francisco foi famoso por abrigar alguns dos criminosos mais notórios do mundo na incontornável ilha de Alcatraz.

O conjunto de dados da competição fornece quase 12 anos de relatórios sobre crimes em todos os bairros de São Francisco. Com o tempo e o local, devemos prever a categoria do crime que ocorreu.

Outro desafio proposto foi exibir estes dados de forma visual, já que temos coordenadas.

Features

							C 7 888
Descript	DayOfWeek	PdDistrict	Resolution	Address	х	Υ	Category
WARRANT ARREST	Wednesday	NORTHERN	ARREST, BOOKED	OAK ST / LAGUNA ST	-122.425892	37.774599	WARRANTS
TRAFFIC VIOLATION ARREST	Wednesday	NORTHERN	ARREST, BOOKED	OAK ST / LAGUNA ST	-122.425892	37.774599	OTHER OFFENSES
TRAFFIC VIOLATION ARREST	Wednesday	NORTHERN	ARREST, BOOKED	VANNESS AV / GREENWICH ST	-122.424363	37.800414	OTHER OFFENSES
GRAND THEFT FROM LOCKED AUTO	Wednesday	NORTHERN	NONE	1500 Block of LOMBARD ST	-122.426995	37.800873	LARCENY/T HEFT
GRAND THEFT FROM LOCKED AUTO	Wednesday	PARK	NONE	100 Block of BRODERICK ST	-122.438738	37.77154	LARCENY/T HEFT

Describe

In [172]: df_test.describe(include='all')

Out[172]:

									A	dd	ire	SS						X						Y
										88	42	62	8	842	262	2.0	000	000	1	384	126	2.0	000)00
										2	31	84					N	laN					N	laN
cl	800 BI	00 B) BI	Blo	loc	ck	of	BR	RY/	AN	IT S	ST					N	laN					N	laN
										2	69	84					N	laN					N	laN
											Na	aN		-1	122	2.4	226	593			3	7.7	714	176
											Na	aN			(0.0	309	985	,		0	0.4	848	324
											Na	aN		-1	122	2.5	136	642			3	7.7	078	379
											Na	aN		-1	122	2.4	33(069	1		3	7.7	523	374
											Na	aN		-1	122	2.4	165	517			3	7.7	754	121
											Na	aN		-1	122	2.4	069	959			3	7.7	843	353
											Na	aN		-1	120	0.5	000	000	V .		90	0.0	000)00
											Na Na	aN aN		-1 -1	122	2.4	165	517 959	,			3	37.7	37.7754 37.7843

In [173]: df_train.describe(include='all')

Out[173]:

	Category	Descript	DayOfWeek	PdDistrict	Resolution	Address	X	Y
count	878049	878049	878049	878049	878049	878049	878049.000000	878049.000000
unique	39	879	7	10	17	23228	NaN	NaN
top	LARCENY/THEFT	GRAND THEFT FROM LOCKED AUTO	Friday	SOUTHERN	NONE	800 Block of BRYANT ST	NaN	NaN
freq	174900	60022	133734	157182	526790	26533	NaN	NaN
mean	NaN	NaN	NaN	NaN	NaN	NaN	-122.422616	37.771020
std	NaN	NaN	NaN	NaN	NaN	NaN	0.030354	0.456893
min	NaN	NaN	NaN	NaN	NaN	NaN	-122.513642	37.707879
25%	NaN	NaN	NaN	NaN	NaN	NaN	-122.432952	37.752427
50%	NaN	NaN	NaN	NaN	NaN	NaN	-122.416420	37.775421
75%	NaN	NaN	NaN	NaN	NaN	NaN	-122.406959	37.784369
max	NaN	NaN	NaN	NaN	NaN	NaN	-120.500000	90.000000

Mapeamento de crimes

```
In [3]: categories = df_train["Category"].unique()
    category_map = {}
    for key, crime in enumerate(categories):
        category_map[key] = crime

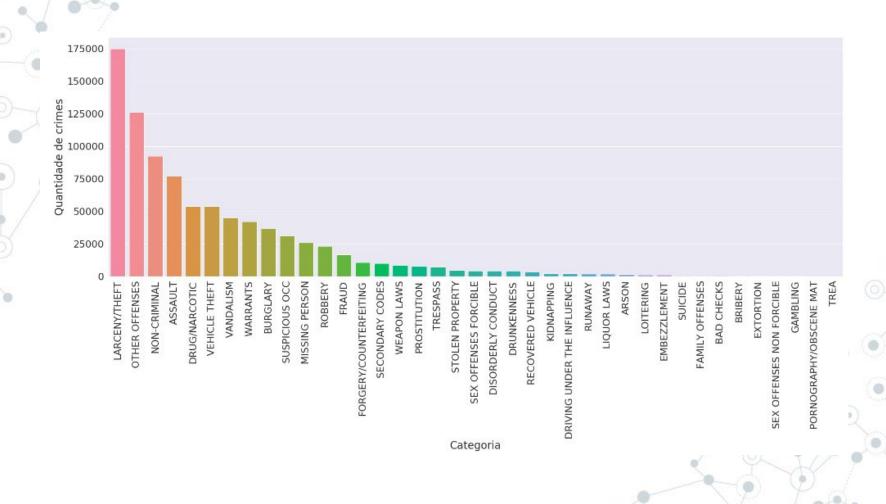
print(category_map)
    print("\nQuantidade de crimes:",len(categories))

{0: 'WARRANTS', 1: 'OTHER OFFENSES', 2: 'LARCENY/THEFT', 3: 'VEHICLE THEFT', 4: 'VANDALISM', 5: 'NON-CRIMINAL', 6:
        'ROBBERY', 7: 'ASSAULT', 8: 'WEAPON LAWS', 9: 'BURGLARY', 10: 'SUSPICIOUS OCC', 11: 'DRUNKENNESS', 12: 'FORGERY/COU
        NTERFEITING', 13: 'DRUG/NARCOTIC', 14: 'STOLEN PROPERTY', 15: 'SECONDARY CODES', 16: 'TRESPASS', 17: 'MISSING PERSO
        N', 18: 'FRAUD', 19: 'KIDNAPPING', 20: 'RUNAWAY', 21: 'DRIVING UNDER THE INFLUENCE', 22: 'SEX OFFENSES FORCIBLE', 2
        3: 'PROSTITUTION', 24: 'DISORDERLY CONDUCT', 25: 'ARSON', 26: 'FAMILY OFFENSES', 27: 'LIQUOR LAWS', 28: 'BRIBERY',
        29: 'EMBEZZLEMENT', 30: 'SUICIDE', 31: 'LOITERING', 32: 'SEX OFFENSES NON FORCIBLE', 33: 'EXTORTION', 34: 'GAMBLIN
        G', 35: 'BAD CHECKS', 36: 'TREA', 37: 'RECOVERED VEHICLE', 38: 'PORNOGRAPHY/OBSCENE MAT'}

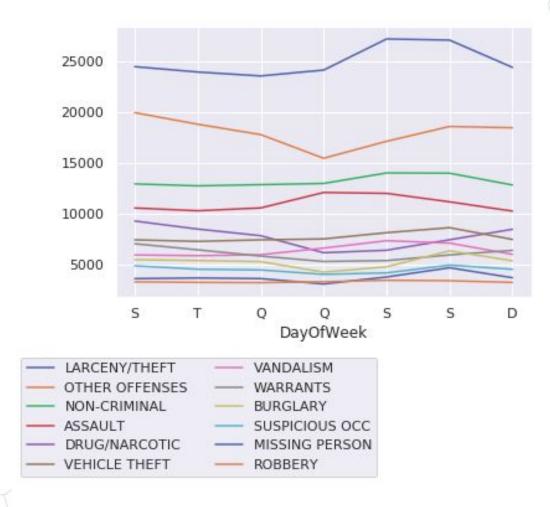
Ouantidade de crimes: 39
```



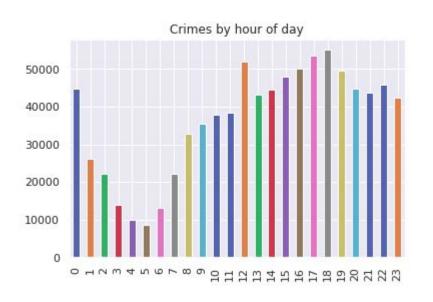
Quantidade de crimes por categoria

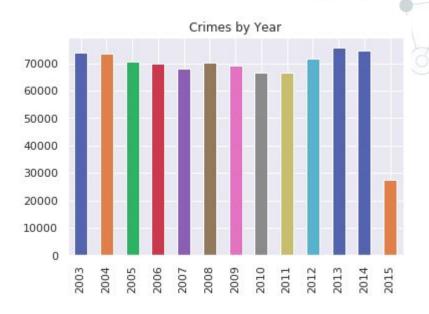


Crimes por dia da semana



Quantidade por tempo







0.2

0.0

0.0

0.2

0.4

0.6

0.8

0.2

0.0

0.0

0.2

1.0

1.0

8.0

