

This notebook explores Chicago Crime public dataset (bigquery-public-data.chicago_crime.crime)

```
query = """
SELECT count(arrest) FROM `bigquery-public-data.chicago_crime.crime` where arrest IS true
"""
```

Call BigQuery and examine in dataframe

```
import google.datalab.bigquery as bq
df = bq.Query(query + " LIMIT 100").execute().result().to_dataframe()
```

```
print("There were " + str(df.at[0,"f0_"]) + " arrests in Chicago")
```

There were 1874936 arrests in Chicago

Chicago coordinates are: latitude 41.8781° N, longitude 87.6298° W

#example row

I create a table with ~ 1/5 of data : SELECT * FROM bigquery-public-data.chicago_crime.crime where MOD(unique_key, 5) = 0

THIS

```
query = """
SELECT * FROM `ml-sme-223918.bqml_tutorial_us.chicago_crime_subset`
"""
```

THIS

```
import google.datalab.bigquery as bq
df = bq.Query(query + " LIMIT 10000").execute().result().to_dataframe()
```

```
df.describe()
```

	unique_key	beat	district	ward	community_area	x_coordinate	y_coordi
count	1.000000e+04	10000.000000	10000.000000	9094.000000	9093.000000	9.880000e+03	9.880000e+03
mean	6.019879e+06	957.668400	8.809300	25.222784	37.324645	1.169884e+06	1.862125e+06
std	2.959119e+06	624.775001	5.129304	12.797437	17.621808	1.009025e+04	3.898737e+04
min	6.400000e+02	512.000000	5.000000	2.000000	3.000000	1.145015e+06	1.818775e+06
25%	3.354052e+06	522.000000	5.000000	9.000000	30.000000	1.162321e+06	1.828462e+06
50%	5.864905e+06	531.000000	5.000000	25.000000	49.000000	1.173140e+06	1.835508e+06
75%	8.407878e+06	1033.000000	10.000000	34.000000	53.000000	1.178061e+06	1.888487e+06
max	1.152740e+07	2323.000000	19.000000	48.000000	56.000000	1.188194e+06	1.932093e+06

I observe that latitude is between (41.658132, 41.969159) and longitude is between (-87.743523, -87.586439)

Also I see that year is between 2001 and 2018

```
df.head()
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_descript
0	3045	HL177967	2005-02-12 20:47:00	007XX E 103RD ST	0110	HOMICIDE	FIRST DEGREE MURDER	RETAIL STORE
1	3205	HL435664	2005-06-21 21:28:00	103XX S INDIANA AVE	0110	HOMICIDE	FIRST DEGREE MURDER	STREET
2	20900	HW295447	2013-05-29 15:11:00	000XX W 107TH ST	0110	HOMICIDE	FIRST DEGREE MURDER	STREET
3	1710710	G513455	2001-08-27 23:55:00	104XX S STATE ST	0265	CRIM SEXUAL ASSAULT	AGGRAVATED: OTHER	RESIDENCE
4	11363170	JB327133	2018-06-29 00:44:13	002XX W 104TH ST	0281	CRIM SEXUAL ASSAULT	NON-AGGRAVATED	RESIDENCE

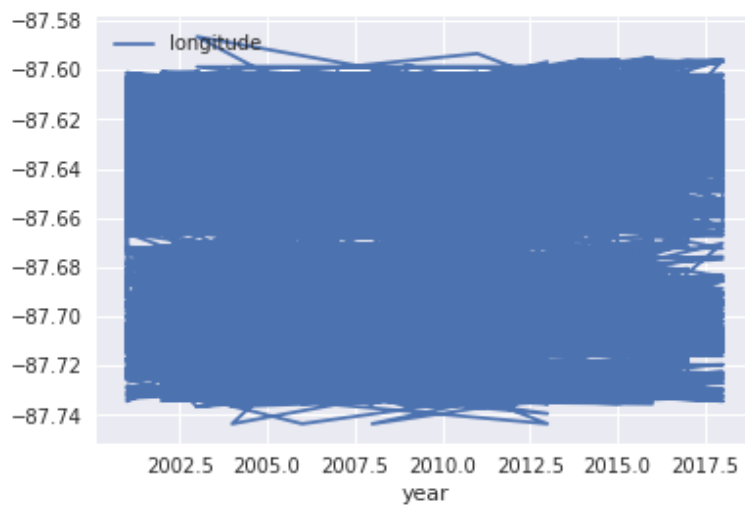
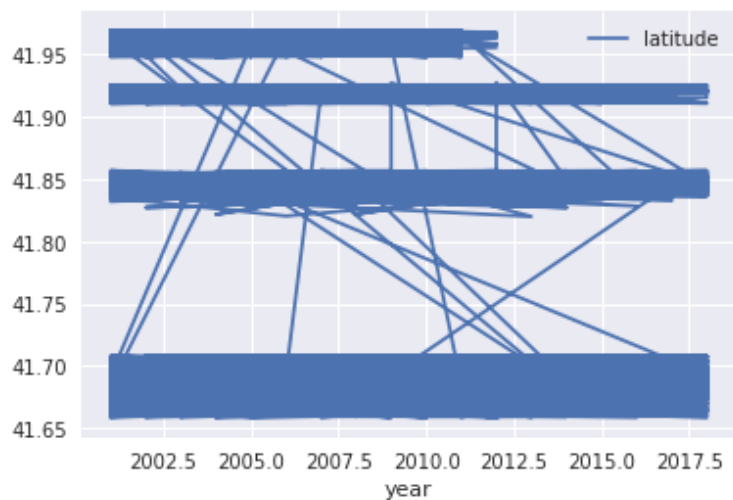
5 rows × 22 columns

I see in BigQuery: Table size 271.74 MB

Number of rows 1,353,959

```
df.plot(x='year', y='latitude')
df.plot(x='year', y='longitude')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f8bfc245250>



I see a lot of crime activity happening between 41.66 : 41.71 latitude in Chicago.

```
print(df['location'][0])
```

```
(41.707455731, -87.605637491)
```

```
print(df['location'])
```

```
for x in df['location'][0]:
    print(x)
```

```
(
4
1
.
7
0
7
4
5
5
7
3
1
,
-
8
7
.
6
0
5
6
3
7
4
9
1
)
```

```
print(df['location'][0].find(","))
```

```
13
```

so it's a string

```
##### THIS
import copy
df2=copy.deepcopy(df)
```

<https://docs.python.org/2/library/copy.html> (<https://docs.python.org/2/library/copy.html>).

Now, considering first element is (41.707455731, -87.605637491)

```
print(df2['location'][0][8:13])
```

```
55731
```

```
print(len(df2['location'][0]))
```

```
29
```

```
print(df2['location'][0][23:28])
```

```
37491
```

```
df2['location'][0][8:13]="00000"
```

```
print(df2['location'][0])
```

TypeErrorTraceback (most recent call last)

<ipython-input-25-6dbf6f697f44> in <module>()

----> 1 df2['location'][0][8:13]="00000"

2 print(df2['location'][0])

TypeError: 'newstr' object does not support item assignment

```

for index, row in df2.iterrows():
    print(row)
    print(row['location'])
    break

```

```

unique_key          3045
case_number         HL177967
date               2005-02-12 20:47:00
block              007XX E 103RD ST
iucr                0110
primary_type        HOMICIDE
description         FIRST DEGREE MURDER
location_description RETAIL STORE
arrest              True
domestic            False
beat               512
district            5
ward                9
community_area      50
fbi_code            01A
x_coordinate         1.18295e+06
y_coordinate         1.83683e+06
year                2005
updated_on          2015-08-17 15:03:40
latitude            41.7075
longitude           -87.6056
location            (41.707455731, -87.605637491)
Name: 0, dtype: object
(41.707455731, -87.605637491)

```

```

for index, row in df2.iterrows():
    print(row['location'])
    tmp = row['location'][0:8] + "00000" + row['location'][13:23] + "00000"
    print(tmp)
    print(row['location'])
    row['location'] = tmp
    print(row['location'])
    break

```

```

(41.707455731, -87.605637491)
(41.707400000, -87.605600000)
(41.707455731, -87.605637491)
(41.707400000, -87.605600000)

```

```

for index, row in df2.iterrows():
    try:
        tmp = row['location'][0:8] + "00000" + row['location'][13:23] + "00000"
        row['location'] = tmp
    except TypeError:
        print(row)
        break
print(df2.head(1))

```

So there are rows for which there is no location set. Need to clean it up.

```

(df2[df2["location"] != False]).head()
#df2.head()

```

```

NameErrorTraceback (most recent call last)
<ipython-input-1-74cf8e19ee0e> in <module>()
----> 1 (df2[df2["location"] != False]).head()
      2 #df2.head()

```

NameError: name 'df2' is not defined

```

#checking if there still are any rows with no location data set
for index, row in df2.iterrows():
    try:
        tmp = row['location'][0:8]
    except TypeError:
        print(row['location'])

```

No rows with empty coordinates left (good) but also no change in location (bad).

```
##### THIS
#df2.head()
df3=copy.deepcopy(df)
df3.head()
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_descript
0	3045	HL177967	2005-02-12 20:47:00	007XX E 103RD ST	0110	HOMICIDE	FIRST DEGREE MURDER	RETAIL STORE
1	3205	HL435664	2005-06-21 21:28:00	103XX S INDIANA AVE	0110	HOMICIDE	FIRST DEGREE MURDER	STREET
2	20900	HW295447	2013-05-29 15:11:00	000XX W 107TH ST	0110	HOMICIDE	FIRST DEGREE MURDER	STREET
3	1710710	G513455	2001-08-27 23:55:00	104XX S STATE ST	0265	CRIM SEXUAL ASSAULT	AGGRAVATED: OTHER	RESIDENCE
4	11363170	JB327133	2018-06-29 00:44:13	002XX W 104TH ST	0281	CRIM SEXUAL ASSAULT	NON-AGGRAVATED	RESIDENCE

5 rows × 22 columns

```
print((df3[df3["location"] != False]).shape[0])
print((df3[df3["location"] == False]).shape[0])
print((df3[df3["location"].notnull()]).shape[0])
```

```
10000
0
9880
```

```
##### THIS
#this is how to filter rows with None in location
df3 = df3[df3["location"].notnull()]
print(df3.shape[0])
```

```
9880
```



```
##### THIS
#let's really change the location
for index, row in df3.iterrows():
    try:
        #print("index="+index)
        tmp = row['location'][0:7] + "000000" + row['location'][12:23] + "000000)"
        #print("tmp="+tmp)
        df3.set_value(index, 'location', tmp)
        #break
    except TypeError:
        print("TypeError in:" + row)
#print(df3.head())
```

/usr/local/envs/py2env/lib/python2.7/site-packages/ipykernel/__main__.py:8: FutureWarning: set_value is deprecated and will be removed in a future release. Please use .at[] or .iat[] accessors instead

```
print(df3.head(1))
```

```

unique_key case_number      date      block iucr \
0    3045  HL177967 2005-02-12 20:47:00 007XX E 103RD ST 0110

primary_type      description location_description arrest domestic \
0  HOMICIDE FIRST DEGREE MURDER      RETAIL STORE  True  False

...      ward community_area fbi_code \
0      ...      9.0      50.0      01A

x_coordinate y_coordinate year      updated_on latitude longitude \
0  1182951.0  1836828.0 2005 2015-08-17 15:03:40 41.707456 -87.605637

location
0 (41.707000000001, -87.6050000000)

[1 rows x 22 columns]
```

```
#let's plot the crime area
#first, sum up crime # in same location
df4 = df3.groupby('location').count()
df4.head(1)
#df4.plot(x='location', y='count', logy=True, kind='bar');
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_descrip
location								
(41.6580000000, -87.6340000000)	2	2	2	2	2	2	2	2
(41.6580000000, -87.6357000000)	1	1	1	1	1	1	1	1
(41.6580000000, -87.6380000000)	1	1	1	1	1	1	1	1
(41.6580000000, -87.6393000000)	1	1	1	1	1	1	1	1
(41.6580000000, -87.6404000000)	3	3	3	3	3	3	3	3

5 rows × 21 columns

```
df5=df4.sort_values(by='case_number', ascending=False)
df5.head(3)
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_descrip
location								
(41.7050000000, -87.6009000000)	40	40	40	40	40	40	40	40
(41.6920000000, -87.6043000000)	39	39	39	39	39	39	39	39
(41.7070000000, -87.6018000000)	39	39	39	39	39	39	39	39
(41.9640000000, -87.6547000000)	33	33	33	33	33	33	33	33
(41.8490000000, -87.7088000000)	33	33	33	33	33	33	33	33

5 rows × 21 columns

```
df5.plot(x='location', y='case_number', kind='bar')
```

The above is because index was set to location , need to be reset
<https://stackoverflow.com/questions/31167896/keyerror-in-dataframe>
(<https://stackoverflow.com/questions/31167896/keyerror-in-dataframe>).

```
df5 = df5.reset_index()
df5.head(1)
```

	index	location	unique_key	case_number	date	block	iucr	primary_type	description	location
0	0	(41.705000000000, -87.600900000000)	40	40	40	40	40	40	40	40

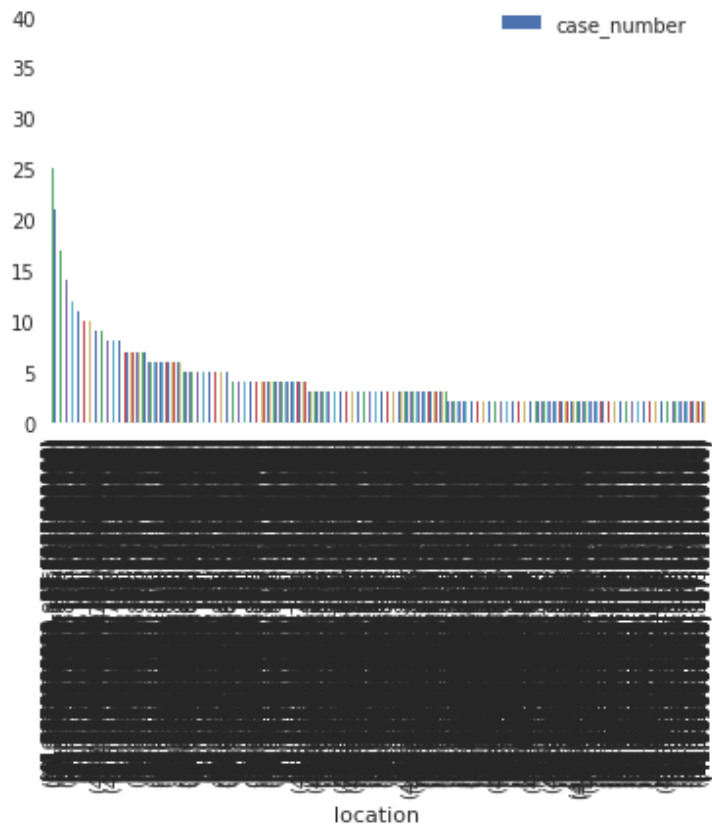
1 rows x 23 columns

```
print(df5.shape[0])
df5 = df5[df5['case_number']>1]
print(df5.shape[0])
```

4174
1784

```
df5.plot(x='location', y='case_number', kind='bar')
```

<matplotlib.axes._subplots.AxesSubplot at 0x7f9538462f90>



Cant' see much from the chart.

Anyway, this is the area with most crimes: <https://goo.gl/maps/sG6bqFV9Xcm>
(<https://goo.gl/maps/sG6bqFV9Xcm>).

in my dataframe (**not** in Chicago - since I only took 10,000 rows from the > 1 M rows)
(after removing 6xzeros from both latitude and longitude)

http://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.drop_duplicates.html
(http://pandas.pydata.org/pandas-docs/stable/generated/pandas.DataFrame.drop_duplicates.html).

```
##### THIS
#df3 has the rows that have location set
print(df3.shape[0])
df31=df3.drop_duplicates(subset="primary_type")
print(df31.shape[0])
```

9880

25

```
print(df31["primary_type"])
```

```
0          HOMICIDE
3      CRIM SEXUAL ASSAULT
7          ROBBERY
43         BATTERY
174    PUBLIC PEACE VIOLATION
289         ASSAULT
375        STALKING
376        BURGLARY
441         THEFT
576    MOTOR VEHICLE THEFT
625         ARSON
626    DECEPTIVE PRACTICE
655    CRIMINAL DAMAGE
755    CRIMINAL TRESPASS
770    WEAPONS VIOLATION
786    PROSTITUTION
803    SEX OFFENSE
805        GAMBLING
806    OFFENSE INVOLVING CHILDREN
816        KIDNAPPING
817        NARCOTICS
935    LIQUOR LAW VIOLATION
936    OTHER OFFENSE
975    INTERFERENCE WITH PUBLIC OFFICER
1614    INTIMIDATION
Name: primary_type, dtype: object
```

```
##### THIS
```

```
LIST_PRIMARY_TYPE = df31["primary_type"].tolist()
```

```
print(LIST_PRIMARY_TYPE)
```

```
['HOMICIDE', 'CRIM SEXUAL ASSAULT', 'ROBBERY', 'BATTERY', 'PUBLIC PEACE VIOLATIO
N', 'ASSAULT', 'STALKING', 'BURGLARY', 'THEFT', 'MOTOR VEHICLE THEFT', 'ARSON', 'DE
CEPTIVE PRACTICE', 'CRIMINAL DAMAGE', 'CRIMINAL TRESPASS', 'WEAPONS VIOLATIO
N', 'PROSTITUTION', 'SEX OFFENSE', 'GAMBLING', 'OFFENSE INVOLVING CHILDREN', 'KID
NAPPING', 'NARCOTICS', 'LIQUOR LAW VIOLATION', 'OTHER OFFENSE', 'INTERFERENCE
WITH PUBLIC OFFICER', 'INTIMIDATION']
```

```
print(df31.head(1))
```

```
unique_key case_number      date      block iucr \
0    3045  HL177967 2005-02-12 20:47:00 007XX E 103RD ST 0110

primary_type      description location_description arrest domestic \
0  HOMICIDE  FIRST DEGREE MURDER      RETAIL STORE  True  False

...      ward community_area fbi_code \
0      ...      9.0      50.0      01A

x_coordinate y_coordinate year      updated_on latitude longitude \
0  1182951.0  1836828.0 2005 2015-08-17 15:03:40 41.707456 -87.605637

location
0 (41.7070000000, -87.6056000000)

[1 rows x 22 columns]
```

So things of interest: primary_type ; location_description ; arrest ; domestic ; year ; location

```
##### THIS
df32=df3.drop_duplicates(subset="location_description")
print(df32.shape[0])
print(df32["location_description"])
```

81	
0	RETAIL STORE
1	STREET
3	RESIDENCE
5	VEHICLE NON-COMMERCIAL
7	HOTEL/MOTEL
8	SIDEWALK
10	GAS STATION
13	PARKING LOT/GARAGE(NON.RESID.)
15	RESIDENCE-GARAGE
19	TAXICAB
32	SMALL RETAIL STORE
37	SCHOOL, PUBLIC, BUILDING
44	SCHOOL, PUBLIC, GROUNDS
45	RESIDENCE PORCH/HALLWAY
53	APARTMENT
70	OTHER
84	VEHICLE-COMMERCIAL
88	CTA BUS
120	ALLEY
136	RESTAURANT
216	RESIDENTIAL YARD (FRONT/BACK)
297	POLICE FACILITY/VEH PARKING LOT
323	GROCERY FOOD STORE
327	TAVERN/LIQUOR STORE
352	CHA PARKING LOT/GROUNDS
388	CONSTRUCTION SITE
392	VACANT LOT/LAND
418	CHA APARTMENT
455	DRUG STORE
460	ABANDONED BUILDING
	...
2517	HOUSE
2826	FACTORY/MANUFACTURING BUILDING
2915	CAR WASH
2994	OTHER RAILROAD PROP / TRAIN DEPOT
2996	SCHOOL, PRIVATE, BUILDING
3116	AUTO
3148	COLLEGE/UNIVERSITY GROUNDS
3161	NURSING HOME/RETIREMENT HOME
3194	OTHER COMMERCIAL TRANSPORTATION
3427	CTA GARAGE / OTHER PROPERTY
3665	FEDERAL BUILDING
3907	HOSPITAL BUILDING/GROUNDS
3955	MEDICAL/DENTAL OFFICE
3979	CLEANING STORE
4178	JAIL / LOCK-UP FACILITY
4330	FIRE STATION
4759	APPLIANCE STORE

```

4824    CHA HALLWAY/STAIRWELL/ELEVATOR
4899          VACANT LOT
5556          DAY CARE CENTER
5880          LAUNDRY ROOM
6101          BOAT/WATERCRAFT
6161          ATHLETIC CLUB
6349    SCHOOL, PRIVATE, GROUNDS
7253          BOWLING ALLEY
7389          ANIMAL HOSPITAL
8223          YARD
8948          MOVIE HOUSE/THEATER
9027          None
9151    COLLEGE/UNIVERSITY RESIDENCE HALL
Name: location_description, Length: 81, dtype: object

```

```
##### THIS
```

```

LIST_LOCATION_DESCRIPTION = df32["location_description"].tolist()
print(LIST_LOCATION_DESCRIPTION)

```

```

['RETAIL STORE', 'STREET', 'RESIDENCE', 'VEHICLE NON-COMMERCIAL', 'HOTEL/MOTEL',
'SIDEWALK', 'GAS STATION', 'PARKING LOT/GARAGE(NON.RESID.)', 'RESIDENCE-GARAG
E', 'TAXICAB', 'SMALL RETAIL STORE', 'SCHOOL, PUBLIC, BUILDING', 'SCHOOL, PUBLIC,
GROUNDS', 'RESIDENCE PORCH/HALLWAY', 'APARTMENT', 'OTHER', 'VEHICLE-COMMER
CIAL', 'CTA BUS', 'ALLEY', 'RESTAURANT', 'RESIDENTIAL YARD (FRONT/BACK)', 'POLICE
FACILITY/VEH PARKING LOT', 'GROCERY FOOD STORE', 'TAVERN/LIQUOR STORE', 'CHA P
ARKING LOT/GROUNDS', 'CONSTRUCTION SITE', 'VACANT LOT/LAND', 'CHA APARTMEN
T', 'DRUG STORE', 'ABANDONED BUILDING', 'DEPARTMENT STORE', 'CHURCH/SYNAGOG
UE/PLACE OF WORSHIP', 'BARBERSHOP', 'POOL ROOM', 'DRIVEWAY - RESIDENTIAL', 'BA
NK', 'ATM (AUTOMATIC TELLER MACHINE)', 'CONVENIENCE STORE', 'SPORTS ARENA/ST
ADIUM', 'COMMERCIAL / BUSINESS OFFICE', 'PARK PROPERTY', 'CTA TRAIN', 'BAR OR TA
VERN', 'CURRENCY EXCHANGE', 'GOVERNMENT BUILDING/PROPERTY', 'CTA PLATFOR
M', 'LIBRARY', 'CTA BUS STOP', 'PAWN SHOP', 'WAREHOUSE', 'HIGHWAY/EXPRESSWAY', 'H
OUSE', 'FACTORY/MANUFACTURING BUILDING', 'CAR WASH', 'OTHER RAILROAD PROP /
TRAIN DEPOT', 'SCHOOL, PRIVATE, BUILDING', 'AUTO', 'COLLEGE/UNIVERSITY GROUND
S', 'NURSING HOME/RETIREMENT HOME', 'OTHER COMMERCIAL TRANSPORTATION', 'CT
A GARAGE / OTHER PROPERTY', 'FEDERAL BUILDING', 'HOSPITAL BUILDING/GROUNDS',
'MEDICAL/DENTAL OFFICE', 'CLEANING STORE', 'JAIL / LOCK-UP FACILITY', 'FIRE STATIO
N', 'APPLIANCE STORE', 'CHA HALLWAY/STAIRWELL/ELEVATOR', 'VACANT LOT', 'DAY CA
RE CENTER', 'LAUNDRY ROOM', 'BOAT/WATERCRAFT', 'ATHLETIC CLUB', 'SCHOOL, PRIVA
TE, GROUNDS', 'BOWLING ALLEY', 'ANIMAL HOSPITAL', 'YARD', 'MOVIE HOUSE/THEATE
R', None, 'COLLEGE/UNIVERSITY RESIDENCE HALL']

```

We could first test a simple ML model: given primary_type, location_description => predict arrest (Y/N).

BUT it's NOT ENOUGH - these will be memorized! Need to add another feature.
even so, I can do it like this first, just to see the result.

First - need to create 3 datasets: train, eval, test

https://en.wikipedia.org/wiki/Random_seed (https://en.wikipedia.org/wiki/Random_seed) If the same random seed is deliberately shared, it becomes a secret key, so two or more systems using matching pseudorandom number algorithms and matching seeds can generate matching sequences of non-repeating numbers which can be used to synchronize remote systems

[https://docs.scipy.org/doc/numpy-](https://docs.scipy.org/doc/numpy-1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand)

[1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand](https://docs.scipy.org/doc/numpy-1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand)

([https://docs.scipy.org/doc/numpy-](https://docs.scipy.org/doc/numpy-1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand)

[1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand](https://docs.scipy.org/doc/numpy-1.15.1/reference/generated/numpy.random.RandomState.rand.html#numpy.random.RandomState.rand))

```
print(df3.shape[0])
```

```
9880
```

```
##### THIS
```

```
import numpy as np
```

```
np.random.seed(seed=1) #makes result reproducible
```

```
msk = np.random.rand(10) < 0.5
```

```
print(msk)
```

```
print(~msk)
```

```
[ True False True True True True True True True False]
```

```
[False True False False False False False False True]
```

```
##### THIS
```

```
#keeping the last ones for
```

```
testdf=df3[9000:]
```

```
print(testdf.shape[0])
```

```
df3new=df3[0:9000]
```

```
print(df3new.shape[0])
```

```
880
```

```
9000
```

```
##### THIS
```

```
msk = np.random.rand(len(df3new)) < 0.8
```

```
traindf = df3new[msk]
```

```
evaldf = df3new[~msk]
```

```
##### THIS
```

```
import pandas as pd
```

```
import tensorflow as tf
```

```
traindf.head(1)
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_description	arr
0	3045	HL177967	2005-02-12 20:47:00	007XX E 103RD ST	0110	HOMICIDE	FIRST DEGREE MURDER	RETAIL STORE	Tru

1 rows × 22 columns

https://www.tensorflow.org/api_docs/python/tf/estimator/inputs/pandas_input_fn
(https://www.tensorflow.org/api_docs/python/tf/estimator/inputs/pandas_input_fn)

```
##### THIS
# Define your feature columns
def create_feature_cols():
    return [
        tf.feature_column.categorical_column_with_vocabulary_list(key='primary_type', vocabulary_list=LIST_
PRIMARY_TYPE, num_oov_buckets=0),
        tf.feature_column.categorical_column_with_vocabulary_list(key='location_description', vocabulary_list=
LIST_LOCATION_DESCRIPTION, num_oov_buckets=0),
    ]
```

https://www.tensorflow.org/api_docs/python/tf/estimator/inputs/pandas_input_fn
(https://www.tensorflow.org/api_docs/python/tf/estimator/inputs/pandas_input_fn)

```
##### THIS
#https://stackoverflow.com/questions/3765533/python-array-with-string-indices
#def input_fn_train: # returns x, y (where y represents label's class index).
label_arrest_dict={True:0, False:1}
print(label_arrest_dict.keys())
print(label_arrest_dict[False])
```

```
[False, True]
```

```
1
```

```
import numpy as np
x = np.array([1, 2, 3, 4, 5])
f = lambda x: x ** 2
squares = f(x)
print(squares)
```

```
[ 1  4  9 16 25]
```

```

#NOT GOOD
tmptestdf=copy.deepcopy(testdf)
ox = tmptestdf["arrest"][0:5]
print(type(ox))
print(ox)
print(ox.data)
#ox=[True, False]
#print(ox)
func = lambda x: label_arrest_dict[x]
from numpy import vectorize
vfunc = vectorize(func)
#sy = f(ox)
sy = np.apply_along_axis(vfunc, 0, ox)
print(type(sy))
print(sy)
#NOT GOOD

```

```

<class 'pandas.core.series.Series'>
9109  False
9110   True
9111   True
9112   True
9113  False
Name: arrest, dtype: bool

```

```

<type 'numpy.ndarray'>
[1 0 0 0 1]
<class 'pandas.core.series.Series'>
9109  False
9110   True
9111   True
9112   True
9113  False
Name: arrest, dtype: bool

```

```

<type 'numpy.ndarray'>
[1 0 0 0 1]

```

#THIS is what I need

#[https://pandas.pydata.org/pandas-docs/version/0.23.4/generated/pandas.Series.apply.html#pandas.Series.a](https://pandas.pydata.org/pandas-docs/version/0.23.4/generated/pandas.Series.apply.html#pandas.Series.apply)

```
tmptestdf=copy.deepcopy(testdf)
```

```
ox = tmptestdf["arrest"][0:5]
```

```
print(type(ox))
```

```
print(ox)
```

```
func = lambda x: label_arrest_dict[x]
```

```
sy = ox.apply(func)
```

```
print(type(sy))
```

```
print(sy)
```

```
<class 'pandas.core.series.Series'>
```

```
9109  False
```

```
9110  True
```

```
9111  True
```

```
9112  True
```

```
9113  False
```

```
Name: arrest, dtype: bool
```

```
<class 'pandas.core.series.Series'>
```

```
9109  1
```

```
9110  0
```

```
9111  0
```

```
9112  0
```

```
9113  1
```

```
Name: arrest, dtype: int64
```

```
##### THIS
# Create pandas input function
def make_input_fn(df, num_epochs, predictMode=False):
    print("in make_input_fn")
    print("got df of length " + str(df.shape[0]))
    df=df[['primary_type','location_description','arrest']]
    df=df.dropna(how='any')#this is critical as I was getting some strange Internal Errors https://stackoverflow.com/questions/45974009/tensorflow-python-framework-errors-impl-internalerror-unable-to-get-element-fr
    o
    print("after removing null, df has length " + str(df.shape[0]))

    if (not predictMode):
        print("train/evaluate mode")
        func = lambda x: label_arrest_dict[x]
        y = df['arrest'].apply(func)
        shuffle = True
    else:
        print("predict mode")
        y = None
        shuffle = False

    return tf.estimator.inputs.pandas_input_fn(
        x = df[['primary_type','location_description']],
        y = y,
        batch_size = 128,
        num_epochs = num_epochs,
        shuffle = shuffle,
        queue_capacity = 1000,
        num_threads = 1
    )
```

<https://stackoverflow.com/questions/45974009/tensorflow-python-framework-errors-impl-internalerror-unable-to-get-element-fro> (<https://stackoverflow.com/questions/45974009/tensorflow-python-framework-errors-impl-internalerror-unable-to-get-element-fro>).

https://www.tensorflow.org/api_docs/python/tf/feature_column (https://www.tensorflow.org/api_docs/python/tf/feature_column).

```
##### THIS
# Create estimator train and evaluate function
def train_and_evaluate(output_dir, num_train_steps):
    estimator = tf.estimator.LinearClassifier(model_dir = output_dir, feature_columns = create_feature_cols())
    train_spec = tf.estimator.TrainSpec(input_fn = make_input_fn(traindf, None),
                                         max_steps = num_train_steps)
    eval_spec = tf.estimator.EvalSpec(input_fn = make_input_fn(evaldf, 1),
                                       steps = None,
                                       start_delay_secs = 1, # start evaluating after N seconds,
                                       throttle_secs = 5) # evaluate every N seconds
    tf.estimator.train_and_evaluate(estimator, train_spec, eval_spec)
```

```
##### THIS
# Launch tensorboard
from google.datalab.ml import TensorBoard

OUTDIR = './trained_model'
TensorBoard().start(OUTDIR)
```

TensorBoard was started successfully with pid 3460. Click [here \(/_proxy/45643/\)](/_proxy/45643/) to access it.

3460

```
traindf.head(1)
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_description	arr
0	3045	HL177967	2005-02-12 20:47:00	007XX E 103RD ST	0110	HOMICIDE	FIRST DEGREE MURDER	RETAIL STORE	Tru

1 rows × 22 columns

```
traindf[['primary_type','location_description']].head(2)
```

	primary_type	location_description
0	HOMICIDE	RETAIL STORE
1	HOMICIDE	STREET

```
for i in LIST_PRIMARY_TYPE:
```

```
    print(str(LIST_PRIMARY_TYPE.index(i)) + " " + i)
```

```
    0 HOMICIDE
    1 CRIM SEXUAL ASSAULT
    2 ROBBERY
    3 BATTERY
    4 PUBLIC PEACE VIOLATION
    5 ASSAULT
    6 STALKING
    7 BURGLARY
    8 THEFT
    9 MOTOR VEHICLE THEFT
   10 ARSON
   11 DECEPTIVE PRACTICE
   12 CRIMINAL DAMAGE
   13 CRIMINAL TRESPASS
   14 WEAPONS VIOLATION
   15 PROSTITUTION
   16 SEX OFFENSE
   17 GAMBLING
   18 OFFENSE INVOLVING CHILDREN
   19 KIDNAPPING
   20 NARCOTICS
   21 LIQUOR LAW VIOLATION
   22 OTHER OFFENSE
   23 INTERFERENCE WITH PUBLIC OFFICER
   24 INTIMIDATION
```

Run the model

import shutil

shutil.rmtree(OUTDIR, ignore_errors = True)

train_and_evaluate(OUTDIR, 2000)

INFO:tensorflow:Saving checkpoints for 2 into ./trained_model/model.ckpt.
INFO:tensorflow:loss = 76.45322, step = 2
INFO:tensorflow:global_step/sec: 66.7835
INFO:tensorflow:loss = 47.88486, step = 102 (1.498 sec)
INFO:tensorflow:Saving checkpoints for 184 into ./trained_model/model.ckpt.
INFO:tensorflow:Loss for final step: 48.669632.
INFO:tensorflow:Calling model_fn.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Starting evaluation at 2018-12-25-11:52:50
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-184
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Finished evaluation at 2018-12-25-11:52:53
INFO:tensorflow:Saving dict for global step 184: accuracy = 0.84891677, accuracy_baseline = 0.6995439, auc = 0.8444734, auc_precision_recall = 0.90213645, average_loss = 0.393542, global_step = 184, label/mean = 0.6995439, loss = 49.305187, precision = 0.8406516, prediction/mean = 0.6844185, recall = 0.96740013
INFO:tensorflow:Calling model_fn.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Create CheckpointSaverHook.
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-184
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Saving checkpoints for 185 into ./trained_model/model.ckpt.
INFO:tensorflow:loss = 54.28233, step = 185
INFO:tensorflow:global_step/sec: 180.828
INFO:tensorflow:loss = 43.465343, step = 285 (0.557 sec)
INFO:tensorflow:global_step/sec: 271.662
INFO:tensorflow:loss = 60.615414, step = 385 (0.368 sec)
INFO:tensorflow:global_step/sec: 312.029
INFO:tensorflow:loss = 35.615685, step = 485 (0.320 sec)
INFO:tensorflow:global_step/sec: 223.729
INFO:tensorflow:loss = 58.584946, step = 585 (0.448 sec)
INFO:tensorflow:global_step/sec: 104.313
INFO:tensorflow:loss = 42.10739, step = 685 (0.959 sec)
INFO:tensorflow:global_step/sec: 104.045
INFO:tensorflow:loss = 40.585495, step = 785 (0.961 sec)
INFO:tensorflow:Saving checkpoints for 805 into ./trained_model/model.ckpt.
INFO:tensorflow:Loss for final step: 48.950504.
INFO:tensorflow:Calling model_fn.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.

eful_interpolation" instead.

INFO:tensorflow:Done calling model_fn.

INFO:tensorflow:Starting evaluation at 2018-12-25-11:53:04

INFO:tensorflow:Graph was finalized.

INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-805

INFO:tensorflow:Running local_init_op.

INFO:tensorflow:Done running local_init_op.

INFO:tensorflow:Finished evaluation at 2018-12-25-11:53:07

INFO:tensorflow:Saving dict for global step 805: accuracy = 0.8472064, accuracy_baseline = 0.6995439, auc = 0.847172, auc_precision_recall = 0.9050649, average_loss = 0.3903024, global_step = 805, label/mean = 0.6995439, loss = 48.899315, precision = 0.8388693, prediction/mean = 0.6878698, recall = 0.96740013

INFO:tensorflow:Calling model_fn.

INFO:tensorflow:Done calling model_fn.

INFO:tensorflow:Create CheckpointSaverHook.

INFO:tensorflow:Graph was finalized.

INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-805

INFO:tensorflow:Running local_init_op.

INFO:tensorflow:Done running local_init_op.

INFO:tensorflow:Saving checkpoints for 806 into ./trained_model/model.ckpt.

INFO:tensorflow:loss = 53.863503, step = 806

INFO:tensorflow:global_step/sec: 180.103

INFO:tensorflow:loss = 30.358458, step = 906 (0.560 sec)

INFO:tensorflow:global_step/sec: 262.041

INFO:tensorflow:loss = 45.376297, step = 1006 (0.381 sec)

INFO:tensorflow:global_step/sec: 255.385

INFO:tensorflow:loss = 39.570805, step = 1106 (0.392 sec)

INFO:tensorflow:global_step/sec: 256.42

INFO:tensorflow:loss = 43.170433, step = 1206 (0.390 sec)

INFO:tensorflow:global_step/sec: 284.199

INFO:tensorflow:loss = 41.762203, step = 1306 (0.352 sec)

INFO:tensorflow:global_step/sec: 252.268

INFO:tensorflow:loss = 54.657562, step = 1406 (0.396 sec)

INFO:tensorflow:global_step/sec: 280.675

INFO:tensorflow:loss = 55.234707, step = 1506 (0.355 sec)

INFO:tensorflow:global_step/sec: 263.675

INFO:tensorflow:loss = 42.038834, step = 1606 (0.380 sec)

INFO:tensorflow:global_step/sec: 277.7

INFO:tensorflow:loss = 52.49805, step = 1706 (0.360 sec)

INFO:tensorflow:global_step/sec: 277.667

INFO:tensorflow:loss = 61.475388, step = 1806 (0.360 sec)

INFO:tensorflow:Saving checkpoints for 1846 into ./trained_model/model.ckpt.

INFO:tensorflow:Loss for final step: 54.05728.

INFO:tensorflow:Calling model_fn.

WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.

WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.

INFO:tensorflow:Done calling model_fn.

INFO:tensorflow:Starting evaluation at 2018-12-25-11:53:15

INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-1846
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Finished evaluation at 2018-12-25-11:53:15
INFO:tensorflow:Saving dict for global step 1846: accuracy = 0.8449259, accuracy_baseline = 0.6995439, auc = 0.844983, auc_precision_recall = 0.9037896, average_loss = 0.3934129, global_step = 1846, label/mean = 0.6995439, loss = 49.289013, precision = 0.8379335, prediction/mean = 0.69625634, recall = 0.96495515
INFO:tensorflow:Calling model_fn.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Create CheckpointSaverHook.
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-1846
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Saving checkpoints for 1847 into ./trained_model/model.ckpt.
INFO:tensorflow:loss = 46.124084, step = 1847
INFO:tensorflow:global_step/sec: 168.554
INFO:tensorflow:loss = 48.816284, step = 1947 (0.597 sec)
INFO:tensorflow:Saving checkpoints for 2000 into ./trained_model/model.ckpt.
INFO:tensorflow:Loss for final step: 46.2726.
INFO:tensorflow:Calling model_fn.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Starting evaluation at 2018-12-25-11:53:19
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-2000
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Finished evaluation at 2018-12-25-11:53:20
INFO:tensorflow:Saving dict for global step 2000: accuracy = 0.8460661, accuracy_baseline = 0.6995439, auc = 0.8448546, auc_precision_recall = 0.90349704, average_loss = 0.39417997, global_step = 2000, label/mean = 0.6995439, loss = 49.38512, precision = 0.83912116, prediction/mean = 0.68976164, recall = 0.96495515

```
for i in LIST_LOCATION_DESCRIPTION:  
    print(str(LIST_LOCATION_DESCRIPTION.index(i)) + " " + i)  
#LIST_LOCATION_DESCRIPTION[None]
```

- 0 RETAIL STORE
- 1 STREET
- 2 RESIDENCE
- 3 VEHICLE NON-COMMERCIAL
- 4 HOTEL/MOTEL
- 5 SIDEWALK
- 6 GAS STATION
- 7 PARKING LOT/GARAGE(NON.RESID.)
- 8 RESIDENCE-GARAGE
- 9 TAXICAB
- 10 SMALL RETAIL STORE
- 11 SCHOOL, PUBLIC, BUILDING
- 12 SCHOOL, PUBLIC, GROUNDS
- 13 RESIDENCE PORCH/HALLWAY
- 14 APARTMENT
- 15 OTHER
- 16 VEHICLE-COMMERCIAL
- 17 CTA BUS
- 18 ALLEY
- 19 RESTAURANT
- 20 RESIDENTIAL YARD (FRONT/BACK)
- 21 POLICE FACILITY/VEH PARKING LOT
- 22 GROCERY FOOD STORE
- 23 TAVERN/LIQUOR STORE
- 24 CHA PARKING LOT/GROUNDS
- 25 CONSTRUCTION SITE
- 26 VACANT LOT/LAND
- 27 CHA APARTMENT
- 28 DRUG STORE
- 29 ABANDONED BUILDING
- 30 DEPARTMENT STORE
- 31 CHURCH/SYNAGOGUE/PLACE OF WORSHIP
- 32 BARBERSHOP
- 33 POOL ROOM
- 34 DRIVEWAY - RESIDENTIAL
- 35 BANK
- 36 ATM (AUTOMATIC TELLER MACHINE)
- 37 CONVENIENCE STORE
- 38 SPORTS ARENA/STADIUM
- 39 COMMERCIAL / BUSINESS OFFICE
- 40 PARK PROPERTY
- 41 CTA TRAIN
- 42 BAR OR TAVERN
- 43 CURRENCY EXCHANGE
- 44 GOVERNMENT BUILDING/PROPERTY
- 45 CTA PLATFORM
- 46 LIBRARY
- 47 CTA BUS STOP
- 48 PAWN SHOP

49 WAREHOUSE
 50 HIGHWAY/EXPRESSWAY
 51 HOUSE
 52 FACTORY/MANUFACTURING BUILDING
 53 CAR WASH
 54 OTHER RAILROAD PROP / TRAIN DEPOT
 55 SCHOOL, PRIVATE, BUILDING
 56 AUTO
 57 COLLEGE/UNIVERSITY GROUNDS
 58 NURSING HOME/RETIREMENT HOME
 59 OTHER COMMERCIAL TRANSPORTATION
 60 CTA GARAGE / OTHER PROPERTY
 61 FEDERAL BUILDING
 62 HOSPITAL BUILDING/GROUNDS
 63 MEDICAL/DENTAL OFFICE
 64 CLEANING STORE
 65 JAIL / LOCK-UP FACILITY
 66 FIRE STATION
 67 APPLIANCE STORE
 68 CHA HALLWAY/STAIRWELL/ELEVATOR
 69 VACANT LOT
 70 DAY CARE CENTER
 71 LAUNDRY ROOM
 72 BOAT/WATERCRAFT
 73 ATHLETIC CLUB
 74 SCHOOL, PRIVATE, GROUNDS
 75 BOWLING ALLEY
 76 ANIMAL HOSPITAL
 77 YARD
 78 MOVIE HOUSE/THEATER

TypeErrorTraceback (most recent call last)

```

<ipython-input-40-79e021edda86> in <module>()
      1 for i in LIST_LOCATION_DESCRIPTION:
----> 2   print(str(LIST_LOCATION_DESCRIPTION.index(i)) + " " + i)
      3 #LIST_LOCATION_DESCRIPTION[None]
  
```

TypeError: cannot concatenate 'str' and 'NoneType' objects

```

##### THIS
print(LIST_LOCATION_DESCRIPTION[79])
LIST_LOCATION_DESCRIPTION[79]='None'
print(LIST_LOCATION_DESCRIPTION[79])
  
```

None

None

```
##### THIS
# Run the model
import shutil
OUTDIR = './trained_model'
shutil.rmtree(OUTDIR, ignore_errors = True)
train_and_evaluate(OUTDIR, 2000)
```

```
INFO:tensorflow:Using default config.
INFO:tensorflow:Using config: {'_save_checkpoints_secs': 600, '_session_config': None, '_keep_checkpoint_max': 5, '_task_type': 'worker', '_train_distribute': None, '_is_chief': True, '_cluster_spec': <tensorflow.python.training.server_lib.ClusterSpec object at 0x7f203e47bfd0>, '_evaluation_master': '', '_save_checkpoints_steps': None, '_keep_checkpoint_every_n_hours': 10000, '_service': None, '_num_ps_replicas': 0, '_tf_random_seed': None, '_master': '', '_num_worker_replicas': 1, '_task_id': 0, '_log_step_count_steps': 100, '_model_dir': './trained_model', '_global_id_in_cluster': 0, '_save_summary_steps': 100}
in make_input_fn
got df of length 7246
after removing null, df has length 7244
train/evaluate mode
in make_input_fn
got df of length 1754
after removing null, df has length 1754
train/evaluate mode
INFO:tensorflow:Running training and evaluation locally (non-distributed).
INFO:tensorflow:Start train and evaluate loop. The evaluate will happen after 5 secs (eval_spec.throttle_secs) or training is finished.
INFO:tensorflow:Calling model_fn.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Create CheckpointSaverHook.
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Saving checkpoints for 1 into ./trained_model/model.ckpt.
INFO:tensorflow:loss = 88.722855, step = 1
INFO:tensorflow:global_step/sec: 199.716
INFO:tensorflow:loss = 39.79486, step = 101 (0.505 sec)
INFO:tensorflow:global_step/sec: 335.006
INFO:tensorflow:loss = 50.90493, step = 201 (0.298 sec)
INFO:tensorflow:global_step/sec: 342.503
INFO:tensorflow:loss = 45.327618, step = 301 (0.292 sec)
INFO:tensorflow:global_step/sec: 312.776
INFO:tensorflow:loss = 43.32466, step = 401 (0.320 sec)
INFO:tensorflow:global_step/sec: 352.972
INFO:tensorflow:loss = 45.720856, step = 501 (0.283 sec)
INFO:tensorflow:global_step/sec: 337.533
INFO:tensorflow:loss = 48.32393, step = 601 (0.299 sec)
INFO:tensorflow:global_step/sec: 356.561
INFO:tensorflow:loss = 45.16932, step = 701 (0.279 sec)
INFO:tensorflow:global_step/sec: 321.088
INFO:tensorflow:loss = 42.367653, step = 801 (0.311 sec)
INFO:tensorflow:global_step/sec: 363.011
INFO:tensorflow:loss = 41.64128, step = 901 (0.275 sec)
INFO:tensorflow:global_step/sec: 347.941
INFO:tensorflow:loss = 49.06259, step = 1001 (0.288 sec)
INFO:tensorflow:global_step/sec: 295.572
INFO:tensorflow:loss = 46.48625, step = 1101 (0.339 sec)
INFO:tensorflow:global_step/sec: 327.573
```


INFO:tensorflow:loss = 50.491714, step = 1201 (0.305 sec)
INFO:tensorflow:Saving checkpoints for 1251 into ./trained_model/model.ckpt.
INFO:tensorflow:Loss for final step: 29.640606.
INFO:tensorflow:Calling model_fn.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Starting evaluation at 2018-12-25-12:17:59
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-1251
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Finished evaluation at 2018-12-25-12:18:00
INFO:tensorflow:Saving dict for global step 1251: accuracy = 0.84663624, accuracy_baseline = 0.6995439, auc = 0.8465388, auc_precision_recall = 0.904173, average_loss = 0.39155462, global_step = 1251, label/mean = 0.6995439, loss = 49.056202, precision = 0.8392351, prediction/mean = 0.69019306, recall = 0.9657702
INFO:tensorflow:Calling model_fn.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Create CheckpointSaverHook.
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-1251
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Saving checkpoints for 1252 into ./trained_model/model.ckpt.
INFO:tensorflow:loss = 49.580597, step = 1252
INFO:tensorflow:global_step/sec: 220.915
INFO:tensorflow:loss = 35.716206, step = 1352 (0.456 sec)
INFO:tensorflow:global_step/sec: 330.115
INFO:tensorflow:loss = 42.335045, step = 1452 (0.303 sec)
INFO:tensorflow:global_step/sec: 300.052
INFO:tensorflow:loss = 43.848114, step = 1552 (0.333 sec)
INFO:tensorflow:global_step/sec: 339.156
INFO:tensorflow:loss = 44.315094, step = 1652 (0.295 sec)
INFO:tensorflow:global_step/sec: 326.282
INFO:tensorflow:loss = 52.973618, step = 1752 (0.308 sec)
INFO:tensorflow:global_step/sec: 303.292
INFO:tensorflow:loss = 37.343773, step = 1852 (0.328 sec)
INFO:tensorflow:global_step/sec: 316.617
INFO:tensorflow:loss = 42.67659, step = 1952 (0.316 sec)
INFO:tensorflow:Saving checkpoints for 2000 into ./trained_model/model.ckpt.
INFO:tensorflow:Loss for final step: 39.020844.
INFO:tensorflow:Calling model_fn.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
WARNING:tensorflow:Trapezoidal rule is known to produce incorrect PR-AUCs; please switch to "careful_interpolation" instead.
INFO:tensorflow:Done calling model_fn.

```
INFO:tensorflow:Starting evaluation at 2018-12-25-12:18:06
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-2000
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
INFO:tensorflow:Finished evaluation at 2018-12-25-12:18:06
INFO:tensorflow:Saving dict for global step 2000: accuracy = 0.845496, accuracy_baseline = 0.69954
39, auc = 0.84496135, auc_precision_recall = 0.9007598, average_loss = 0.39374632, global_step = 2
000, label/mean = 0.6995439, loss = 49.330788, precision = 0.8385269, prediction/mean = 0.6945196,
recall = 0.96495515
```

lets try to predict now

```
##### THIS
print(label_arrest_dict)
reverse_label_dict={}
for key, value in label_arrest_dict.iteritems():
    print(str(key)+","+str(value))
    reverse_label_dict[value]=key
print(reverse_label_dict)

{False: 1, True: 0}
False,1
True,0
{0: True, 1: False}
```

```
##### THIS
OUTDIR = './trained_model'
estimator = tf.estimator.LinearClassifier(model_dir = OUTDIR, feature_columns = create_feature_cols())
# set steps to None to run evaluation until all data consumed.
results = estimator.predict(
    input_fn = make_input_fn(testdf, 1, predictMode=True))
print("model directory = %s" % OUTDIR)
```

```
INFO:tensorflow:Using default config.
INFO:tensorflow:Using config: {'_save_checkpoints_secs': 600, '_session_config': None, '_keep_check
point_max': 5, '_task_type': 'worker', '_train_distribute': None, '_is_chief': True, '_cluster_spec': <tensor
flow.python.training.server_lib.ClusterSpec object at 0x7f20453aa1d0>, '_evaluation_master': '', '_save
_checkpoints_steps': None, '_keep_checkpoint_every_n_hours': 10000, '_service': None, '_num_ps_rep
licas': 0, '_tf_random_seed': None, '_master': '', '_num_worker_replicas': 1, '_task_id': 0, '_log_step_cou
nt_steps': 100, '_model_dir': './trained_model', '_global_id_in_cluster': 0, '_save_summary_steps': 100}
in make_input_fn
got df of length 880
after removing null, df has length 880
predict mode
model directory = ./trained_model
```

```
print(type(results))
```

```
<type 'generator'>
```

```
print(testdf.loc[[0]])
```

```
print(testdf.iloc[[0]])
```

```
print(testdf.loc[0:0])
```

```
print(testdf.loc[0])
```

```
print(testdf.iloc[0])
```

```

unique_key      8457875
case_number     HV134540
date            2012-01-27 16:20:00
block           010XX W NORTH AVE
iucr            0860
primary_type    THEFT
description     RETAIL THEFT
location_description OTHER
arrest          False
domestic        False
beat            1811
district        18
ward            32
community_area  7
fbi_code        06
x_coordinate    1.16934e+06
y_coordinate    1.91083e+06
year            2012
updated_on      2018-02-10 15:50:01
latitude        41.9108
longitude       -87.6534
location        (41.9100000005, -87.6533000000)

```

Name: 9109, dtype: object

```

unique_key      8457875
case_number     HV134540
date            2012-01-27 16:20:00
block           010XX W NORTH AVE
iucr            0860
primary_type    THEFT
description     RETAIL THEFT
location_description OTHER
arrest          False
domestic        False
beat            1811
district        18
ward            32
community_area  7
fbi_code        06
x_coordinate    1.16934e+06
y_coordinate    1.91083e+06
year            2012
updated_on      2018-02-10 15:50:01
latitude        41.9108
longitude       -87.6534
location        (41.9100000005, -87.6533000000)

```

Name: 9109, dtype: object

```

unique_key      8457875
case_number     HV134540
date            2012-01-27 16:20:00

```

```

block          010XX W NORTH AVE
iucr           0860
primary_type   THEFT
description    RETAIL THEFT
location_description OTHER
arrest         False
domestic       False
beat          1811
district       18
ward           32
community_area 7
fbi_code       06
x_coordinate   1.16934e+06
y_coordinate   1.91083e+06
year           2012
updated_on     2018-02-10 15:50:01
latitude       41.9108
longitude      -87.6534
location       (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
unique_key     8457875
case_number    HV134540
date           2012-01-27 16:20:00
block          010XX W NORTH AVE
iucr           0860
primary_type   THEFT
description    RETAIL THEFT
location_description OTHER
arrest         False
domestic       False
beat          1811
district       18
ward           32
community_area 7
fbi_code       06
x_coordinate   1.16934e+06
y_coordinate   1.91083e+06
year           2012
updated_on     2018-02-10 15:50:01
latitude       41.9108
longitude      -87.6534
location       (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
unique_key     8457875
case_number    HV134540
date           2012-01-27 16:20:00
block          010XX W NORTH AVE
iucr           0860
primary_type   THEFT
description    RETAIL THEFT

```

```

location_description      OTHER
arrest                   False
domestic                 False
beat                    1811
district                 18
ward                     32
community_area           7
fbi_code                 06
x_coordinate             1.16934e+06
y_coordinate             1.91083e+06
year                     2012
updated_on               2018-02-10 15:50:01
latitude                 41.9108
longitude                -87.6534
location                 (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
unique_key               8457875
case_number              HV134540
date                    2012-01-27 16:20:00
block                   010XX W NORTH AVE
iucr                     0860
primary_type             THEFT
description              RETAIL THEFT
location_description      OTHER
arrest                   False
domestic                 False
beat                    1811
district                 18
ward                     32
community_area           7
fbi_code                 06
x_coordinate             1.16934e+06
y_coordinate             1.91083e+06
year                     2012
updated_on               2018-02-10 15:50:01
latitude                 41.9108
longitude                -87.6534
location                 (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
unique_key               8457875
case_number              HV134540
date                    2012-01-27 16:20:00
block                   010XX W NORTH AVE
iucr                     0860
primary_type             THEFT
description              RETAIL THEFT
location_description      OTHER
arrest                   False
domestic                 False
beat                    1811

```

```

district      18
ward          32
community_area      7
fbi_code       06
x_coordinate    1.16934e+06
y_coordinate    1.91083e+06
year          2012
updated_on     2018-02-10 15:50:01
latitude      41.9108
longitude     -87.6534
location      (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
unique_key    8457875
case_number   HV134540
date          2012-01-27 16:20:00
block         010XX W NORTH AVE
iucr          0860
primary_type  THEFT
description   RETAIL THEFT
location_description  OTHER
arrest        False
domestic      False
beat         1811
district      18
ward          32
community_area      7
fbi_code       06
x_coordinate    1.16934e+06
y_coordinate    1.91083e+06
year          2012
updated_on     2018-02-10 15:50:01
latitude      41.9108
longitude     -87.6534
location      (41.9100000005, -87.6533000000)
Name: 9109, dtype: object

```

NO IDEA WHY it has 8 times the same element

```
print(testdf.iloc[0]["arrest"])
```

```

False
False
False
False
False
False
False
False

```



```
print(type(testdf.iloc[0]))
```

```
<class 'pandas.core.series.Series'>
```

```
print(testdf.iloc[0].iloc[0])
```

```
8457875
```

```
jsn_str=testdf.iloc[0].to_json()
```

```
print((testdf.iloc[0])["arrest"])
```

```
print(jsn_str)
```

```
print(type(jsn_str))
```

```
import json
```

```
jsn = json.loads(jsn_str)
```

```
print(jsn["arrest"])
```

```
False
```

```
{"unique_key":8457875,"case_number":"HV134540","date":1327681200000,"block":"010XX W NO  
RTH AVE","iucr":"0860","primary_type":"THEFT","description":"RETAIL THEFT","location_descri  
ption":"OTHER","arrest":false,"domestic":false,"beat":1811,"district":18,"ward":32.0,"community_are  
a":7.0,"fbi_code":"06","x_coordinate":1169336.0,"y_coordinate":1910832.0,"year":2012,"updated_o  
n":1518277801000,"latitude":41.910835515,"longitude":-87.653351515,"location":"(41.9100000005, -  
87.65330000000)"}

```

```
<type 'str'>
```

```
False
```

```
print(testdf.iloc[0])
```

```
unique_key      8457875
case_number      HV134540
date      2012-01-27 16:20:00
block      010XX W NORTH AVE
iucr      0860
primary_type      THEFT
description      RETAIL THEFT
location_description      OTHER
arrest      False
domestic      False
beat      1811
district      18
ward      32
community_area      7
fbi_code      06
x_coordinate      1.16934e+06
y_coordinate      1.91083e+06
year      2012
updated_on      2018-02-10 15:50:01
latitude      41.9108
longitude      -87.6534
location      (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
```

```
testdf_tmp=copy.deepcopy(testdf)
testdf_tmp.head(1)
testdf_tmp.reset_index()
testdf_tmp.head(1)
```

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

	unique_key	case_number	date	block	iucr	primary_type	description	location_description
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER

1 rows × 22 columns

<https://wiki.python.org/moin/Generators> (<https://wiki.python.org/moin/Generators>).

```

numberofprintedtimes=0
def printmax5times(stringg):
    global numberofprintedtimes
    if (numberofprintedtimes<40):
        print(stringg)
        numberofprintedtimes = numberofprintedtimes + 1

```

Generators are iterators, a kind of iterable **you can only iterate over once**. Generators do not store all the values in memory, they generate the values on the fly.

<https://stackoverflow.com/questions/231767/what-does-the-yield-keyword-do>
(<https://stackoverflow.com/questions/231767/what-does-the-yield-keyword-do>).

<https://stackoverflow.com/questions/1663807/how-to-iterate-through-two-lists-in-parallel>
(<https://stackoverflow.com/questions/1663807/how-to-iterate-through-two-lists-in-parallel>).
<https://docs.python.org/2/library/functions.html#zip>
(<https://docs.python.org/2/library/functions.html#zip>).

```
##### THIS
#this can only be ran ONCE bcs results is a generator
correct_results=0
for idx,result in enumerate(results):
    printmax5times("\n")
    printmax5times("#####")
    printmax5times('result: '+str(result))
    printmax5times(result['classes'][0])

    prediction_label = str(reverse_label_dict[int(result['classes'][0])])
    actual_label = str(testdf.iloc[idx]["arrest"])

    printmax5times("prediction was: arrest? " + prediction_label)
    printmax5times("index is " + str(idx))
    printmax5times("and in reality arrest was " + actual_label)
    printmax5times("data was " + str(testdf.iloc[idx]))

    if (prediction_label == actual_label):
        correct_results = correct_results + 1

print("\n")
print("#####")
print("#####")
print("Number of correct results: " + str(correct_results) + " out of a total of " + str(testdf.shape[0]))
```

```
INFO:tensorflow:Calling model_fn.
INFO:tensorflow:Done calling model_fn.
INFO:tensorflow:Graph was finalized.
INFO:tensorflow:Restoring parameters from ./trained_model/model.ckpt-2000
INFO:tensorflow:Running local_init_op.
INFO:tensorflow:Done running local_init_op.
```

```
#####
```

```
result: {'probabilities': array([0.08003888, 0.9199611 ], dtype=float32), 'logits': array([2.441819], dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic': array([0.9199611], dtype=float32)}
```

```
1
```

```
prediction was: arrest? False
```

```
index is 0
```

```
and in reality arrest was False
```

```
data was unique_key          8457875
case_number                   HV134540
date                          2012-01-27 16:20:00
block                          010XX W NORTH AVE
iucr                           0860
primary_type                   THEFT
description                    RETAIL THEFT
location_description           OTHER
arrest                         False
domestic                       False
beat                           1811
district                       18
ward                           32
community_area                 7
fbi_code                       06
x_coordinate                   1.16934e+06
y_coordinate                   1.91083e+06
year                           2012
updated_on                    2018-02-10 15:50:01
latitude                       41.9108
longitude                      -87.6534
location                      (41.9100000005, -87.6533000000)
Name: 9109, dtype: object
```

```
#####
```

```
result: {'probabilities': array([0.30248347, 0.69751656], dtype=float32), 'logits': array([0.8354996], dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic': array([0.69751656], dtype=float32)}
```

```
1
```

```
prediction was: arrest? False
```

```
index is 1
```

```
and in reality arrest was True
```

```

data was unique_key      2659165
case_number              HJ270211
date                    2003-03-30 21:15:00
block                   020XX N CLYBOURN AVE
iucr                    0860
primary_type            THEFT
description              RETAIL THEFT
location_description     SMALL RETAIL STORE
arrest                  True
domestic                False
beat                    1811
district                18
ward                    43
community_area          7
fbi_code                06
x_coordinate             1.16745e+06
y_coordinate             1.91364e+06
year                    2003
updated_on              2018-02-28 15:56:25
latitude                41.9186
longitude                -87.6602
location                (41.9180000008, -87.6601000000)
Name: 9110, dtype: object

```

```
#####
```

```

result: {'probabilities': array([0.30248347, 0.69751656], dtype=float32), 'logits': array([0.8354996], dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic': array([0.69751656], dtype=float32)}

```

```
1
```

```
prediction was: arrest? False
```

```
index is 2
```

```
and in reality arrest was True
```

```

data was unique_key      6041120
case_number              HP142752
date                    2008-01-25 15:00:00
block                   017XX W FULLERTON AVE
iucr                    0860
primary_type            THEFT
description              RETAIL THEFT
location_description     SMALL RETAIL STORE
arrest                  True
domestic                False
beat                    1811
district                18
ward                    32
community_area          7
fbi_code                06
x_coordinate             1.16414e+06
y_coordinate             1.91599e+06

```



```

year                2008
updated_on          2018-02-28 15:56:25
latitude            41.9251
longitude           -87.6723
location            (41.9250000006, -87.6722000000)
Name: 9111, dtype: object

```

```
#####
```

```

result: {'probabilities': array([0.30248347, 0.69751656], dtype=float32), 'logits': array([0.8354996], dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic': array([0.69751656], dtype=float32)}

```

```
1
```

```
prediction was: arrest? False
```

```
index is 3
```

```
and in reality arrest was True
```

```
data was unique_key          3422835
```

```
case_number                  HK483188
```

```
date                        2004-07-08 21:15:00
```

```
block                       010XX W NORTH AVE
```

```
iucr                        0860
```

```
primary_type                THEFT
```

```
description                  RETAIL THEFT
```

```
location_description         SMALL RETAIL STORE
```

```
arrest                      True
```

```
domestic                    False
```

```
beat                        1811
```

```
district                    18
```

```
ward                        32
```

```
community_area              7
```

```
fbi_code                    06
```

```
x_coordinate                 1.16934e+06
```

```
y_coordinate                 1.91083e+06
```

```
year                        2004
```

```
updated_on                  2018-02-28 15:56:25
```

```
latitude                    41.9108
```

```
longitude                   -87.6534
```

```
location                    (41.9100000005, -87.6533000000)
```

```
Name: 9112, dtype: object
```

```
#####
```

```

result: {'probabilities': array([0.09117065, 0.9088294 ], dtype=float32), 'logits': array([2.2994244], dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic': array([0.9088294], dtype=float32)}

```

```
1
```

```
prediction was: arrest? False
```

```
index is 4
```

```
and in reality arrest was False
```

```
data was unique_key          6545005
```

```

case_number      HP617629
date             2008-10-07 11:45:00
block            023XX N SHEFFIELD AVE
iucr             0870
primary_type     THEFT
description      POCKET-PICKING
location_description SIDEWALK
arrest           False
domestic         False
beat            1811
district         18
ward             32
community_area   7
fbi_code         06
x_coordinate     1.16924e+06
y_coordinate     1.9158e+06
year             2008
updated_on       2018-02-28 15:56:25
latitude         41.9245
longitude        -87.6536
location         (41.924000000, -87.65357000000)
Name: 9113, dtype: object

```

```

#####
#####

```

Number of correct results: 766 out of a total of 880

Number of correct results: 766 out of a total of 880

```
print(str((766.0/880)*100) + "% accuracy")
```

87.0454545455% accuracy

```

#####
#####
#####
#####
#####
#####
#####
#####
#####
#####

```

** From <https://stackoverflow.com/questions/46948172/predict-in-tensorflow-estimator-using-input-fn> (<https://stackoverflow.com/questions/46948172/predict-in-tensorflow-estimator-using-input-fn>).

The prediction result for one sample is below:

```
{
'probabilities': array([0.78595656, 0.21404342], dtype = float32),
'logits': array([-1.3007226], dtype = float32),
'classes': array(['0'], dtype = object),
'class_ids': array([0]),
'logistic': array([0.21404341], dtype = float32)
}
```

What each field means are

'probabilities': array([0.78595656, 0.21404342], dtype = float32).

It predicts the output label is class-0 (in this case <=50K) with confidence 0.78595656

'logits': array([-1.3007226], dtype = float32)

The value of z in equation $1/(1+e^{(-z)})$ is -1.3.

'classes': array(['0'], dtype = object)

The class label is 0

```
result: {'probabilities': array([0.31800354, 0.68199646], dtype=float32), 'logits': array([0.762962],
dtype=float32), 'classes': array(['1'], dtype=object), 'class_ids': array([1]), 'logistic':
array([0.68199646], dtype=float32)} result: {'probabilities': array([0.83636373, 0.1636363 ],
dtype=float32), 'logits': array([-1.6314174], dtype=float32), 'classes': array(['0'], dtype=object),
'class_ids': array([0]), 'logistic': array([0.16363628], dtype=float32)}
```

```
##### THIS  
df321=df3.drop_duplicates(subset="arrest")  
print(df32.shape[0])  
print(df32["arrest"])
```

```
81
0    True
1    True
3    False
5    True
7    False
8    False
10   True
13   False
15   False
19   False
32   False
37   True
44   False
45   False
53   False
70   True
84   False
88   False
120  False
136  True
216  False
297  False
323  True
327  False
352  False
388  False
392  False
418  False
455  True
460  False
...
2517 False
2826 False
2915 True
2994 True
2996 False
3116 True
3148 True
3161 False
3194 False
3427 True
3665 False
3907 True
3955 False
3979 False
4178 True
4330 False
4759 False
```

4824 True
 4899 False
 5556 False
 5880 True
 6101 False
 6161 False
 6349 True
 7253 False
 7389 False
 8223 False
 8948 False
 9027 False
 9151 False

Name: arrest, Length: 81, dtype: bool

testdf.head(3)

	unique_key	case_number	date	block	iucr	primary_type	description	location_descri
9109	8457875	HV134540	2012-01-27 16:20:00	010XX W NORTH AVE	0860	THEFT	RETAIL THEFT	OTHER
9110	2659165	HJ270211	2003-03-30 21:15:00	020XX N CLYBOURN AVE	0860	THEFT	RETAIL THEFT	SMALL RETAI STORE
9111	6041120	HP142752	2008-01-25 15:00:00	017XX W FULLERTON AVE	0860	THEFT	RETAIL THEFT	SMALL RETAI STORE

3 rows x 22 columns

result: {'probabilities': array([0.91876584, 0.0812341], dtype=float32), 'logits': array([-2.4256961], dtype=float32), 'classes': array(['0'], dtype=object), 'class_ids': array([0]), 'logistic': array([0.0812341], dtype=float32)} result: {'probabilities': array([0.69484776, 0.30515227], dtype=float32), 'logits': array([-0.82288194], dtype=float32), 'classes': array(['0'], dtype=object), 'class_ids': array([0]), 'logistic': array([0.30515227], dtype=float32)} result: {'probabilities': array([0.69484776, 0.30515227], dtype=float32), 'logits': array([-0.82288194], dtype=float32), 'classes': array(['0'], dtype=object), 'class_ids': array([0]), 'logistic': array([0.30515227], dtype=float32)}