

# SMOTE

Data: (1) UpSamp\_Demo.csv (2) data\_final.csv

# SMOTE concept

data - DataFrame

Index	age	employ	address	default
2	40	15	14	0
3	41	15	14	0
4	24	2	0	1
5	41	5	5	0
6	39	20	9	0
7	43	12	11	0
8	24	3	4	1
9	36	0	13	0
10	27	0	1	0
11	25	4	0	0
12	52	24	14	0
13	37	6	9	0

```
# Jesus is my Saviour!
```

```
import pandas as pd
```

```
import sklearn
```

```
from sklearn.utils import resample
```

```
data = pd.read_csv("C:/Users/Dr Vinod/Desktop/DataSets1/UpSamp_ Demo.csv")
```

```
data = pd.DataFrame(data)
```

```
data.shape # 14 by 9
```

```
data.info()
```

```
data.default.value_counts()
```

```
'''
```

```
0    11
```

```
1     3
```

```
Name: default, dtype: int64
```

```
'''
```

```
3/11 # 27% are 1
```

# Separate yes and no

```
# separate minority and majority classes  
  
not_default = data[data.default==0] #11  
len(not_default)  
default = data[data.default==1] # 3  
len(default)
```

# Increase counts of 1 = counts of 0

Resampling code

Resampling will happen to this class (1)

```
# upsample minority with replacement  
from sklearn.utils import resample  
default_upsampled1 = resample(default,  
                               replace=True, # sample with replacement  
                               n_samples=len(not_default), # match number in majority class  
                               random_state=27) # reproducible results
```

Without replacement it will not be possible!

Nos of sample (1)  
= class (0)

# Combine vertically

```
# combine majority and upsampled minority

upsampled1 = pd.concat([not_default, default_upsampled1]) #22, 11, 11

# check new class counts
upsampled1.default.value_counts() #11 11


upsampled1.to_csv('C:/Users/Dr Vinod/Desktop/DataSets1/upsampled1.csv')
```

```
In [49]: upsampled1.default.value_counts() #11 11
Out[49]:
1      11
0      11
Name: default, dtype: int64
```



# Create predictors and target variable

```
data_final.to_csv("C:/Users/Dr Vinod/Desktop/data_final.csv")  
# smote  
X = data_final.loc[:, data_final.columns != 'y']  
y = data_final.loc[:, data_final.columns == 'y']
```



How you should  
apply on a large  
data set?

```
data_final DataFrame (41188, 62)
```

```
In [36]: y.value_counts()  
Out[36]:  
y  
0      36548  
1      4640  
dtype: int64
```

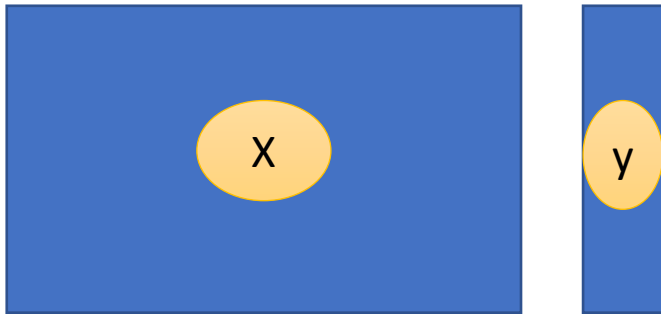
# 30 % test data

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

X_test	DataFrame	(12357, 61)
X_train	DataFrame	(28831, 61)

# Join horizontally x\_train and y\_train

```
# 1st join x_train and y_train  
train = X_train.join(y_train)  
  
train.info()
```



X_train	DataFrame	(28831, 61)
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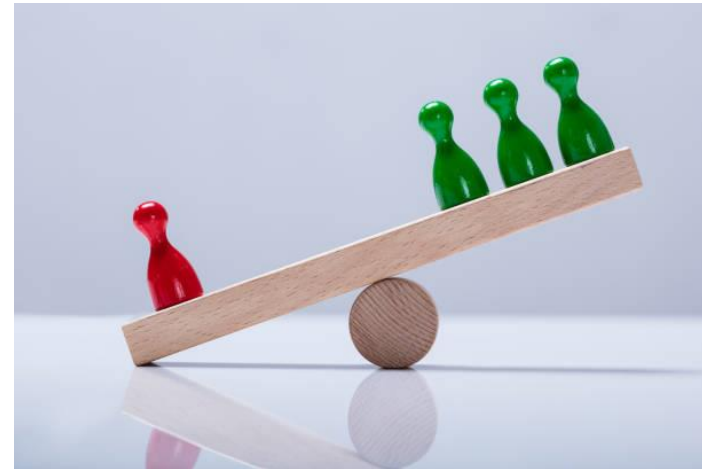
y_train	DataFrame	(28831, 1)
---------	-----------	------------

train	DataFrame	(28831, 62)
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# Count the imbalanced categories

```
not_subsc = train[train.y == 0]  
len(not_subsc) #25,567  
subsc = train[train.y == 1]  
len(subsc) # 3264
```



# Make minority class = majority class

```
# 2 upsample; minor catg 'subsc' to be increased to counts = not_subsc
from sklearn.utils import resample
subsc_os = resample(subsc,
                    replace=True, # sample with replacement
                    n_samples=len(not_subsc), # match number in majority class
                    random_state=27) # reproducible results
```

```
train_os = pd.concat([not_subsc, subsc_os])
```

```
train_os.y.value_counts()
'''
1      25567
0      25567
Name: y, dtype: int64
'''
```



not\_subsc

subsc\_os

```
not_subsc = train[train.y == 0]
len(not_subsc) #25,567
subsc = train[train.y == 1]
len(subsc) # 3264
```

train_os	DataFrame	(51134, 62)
----------	-----------	-------------

# Now make oversampled x\_train(os) & y\_train(os)

```
# 3 make x_trainos, y_trainos  
X_trainos = train_os.loc[:, train_os.columns != 'y']  
y_trainos = train_os.loc[:, train_os.columns == 'y']
```

x_trainos	DataFrame	(51134, 61)
-----------	-----------	-------------

y_trainos	DataFrame	(51134, 1)
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train_os	DataFrame	(51134, 62)
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