

8D

March 11, 2022

```
[1]: %matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import pandas as pd
import numpy as np
from sklearn.datasets import load_iris
from sklearn.linear_model import SGDClassifier
from sklearn.model_selection import GridSearchCV
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[2]: data = pd.read_csv('task_d.csv')
```

```
[3]: X = data.drop(['target'], axis=1).values
Y = data['target'].values
data.head()
```

```
[3]:
```

	x	y	z	x*x	2*y	2*z+3*x*x	w \
0	-0.581066	0.841837	-1.012978	-0.604025	0.841837	-0.665927	-0.536277
1	-0.894309	-0.207835	-1.012978	-0.883052	-0.207835	-0.917054	-0.522364
2	-1.207552	0.212034	-1.082312	-1.150918	0.212034	-1.166507	0.205738
3	-1.364174	0.002099	-0.943643	-1.280666	0.002099	-1.266540	-0.665720
4	-0.737687	1.051772	-1.012978	-0.744934	1.051772	-0.792746	-0.735054

	target
0	0
1	0
2	0
3	0
4	0

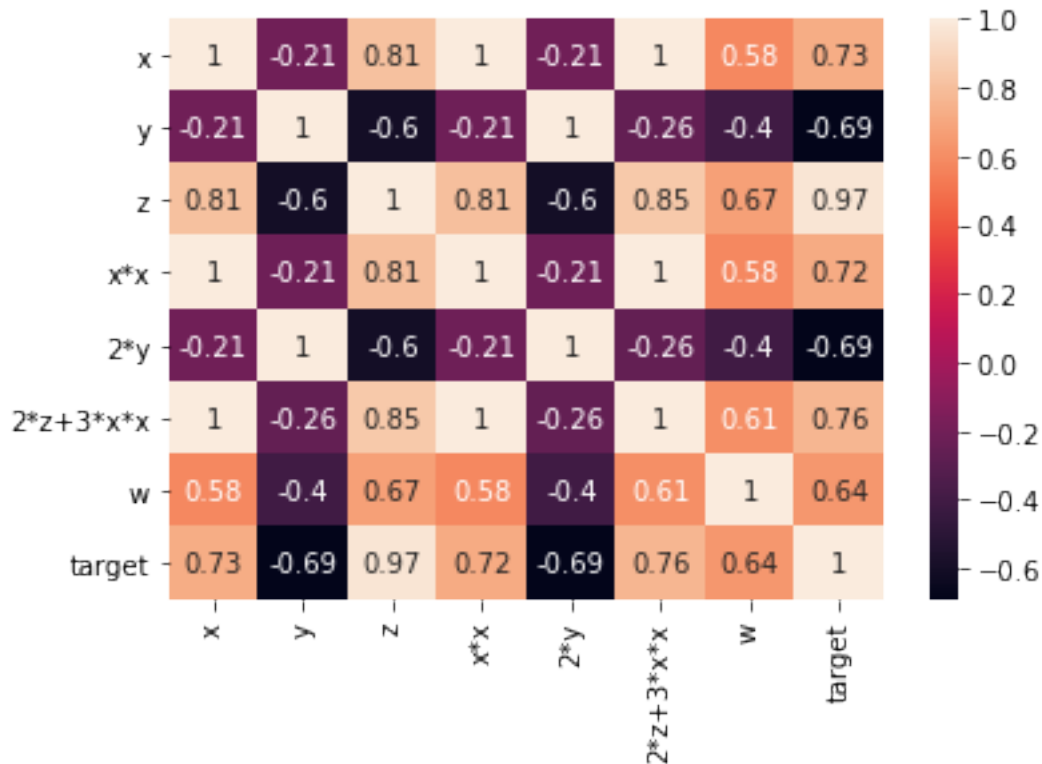
0.1 step 1 Finding the Correlation between the features

```
[4]: import seaborn as sn

corr_df=data.corr()
corr=np.array(corr_df)
label=list(data)                                     #ylabel from data heading
```

```
sn.heatmap(data=corr,annot=True,xticklabels=label,yticklabels=label)
```

[4]: <AxesSubplot:>



0.2 step 2 Finding the best model for the given data

```
[19]: from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import RepeatedStratifiedKFold
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split

X_train,X_test,y_train,y_test=train_test_split(X,Y,random_state = 42,
→test_size=0.2)

parameters = {'C':(np.logspace(-5,8,10))}

cv = RepeatedStratifiedKFold(n_splits=10, n_repeats=3, random_state=1)
#https://machinelearningmastery.com/
→hyperparameter-optimization-with-random-search-and-grid-search/

log_reg=LogisticRegression()
```

```

log_reg_cv = GridSearchCV(log_reg,
    ↳parameters,n_jobs=-1,return_train_score=True,cv=cv)
log_reg_cv.fit(X, Y) # fitting on whole
    ↳data for find out best hyperparameter

print("best parameter = ",log_reg_cv.best_params_)
print("best score = ",log_reg_cv.best_score_)
print("best estimator = ",log_reg_cv.best_estimator_)

```

```

best parameter = {'C': 1e-05}
best score = 1.0
best estimator = LogisticRegression(C=1e-05)

```

0.3 step 3 Getting the weights with the original data

```

[6]: # step 3
best_model=LogisticRegression(C=1e-05)
best_model.fit(X_train,y_train) #fitting training data on
    ↳best_hyperparameter_model
y_pred=best_model.predict(X_test)
best_model_accuracy=accuracy_score(y_test, y_pred) #getting the accuracy

weight=best_model.coef_
print("coef :-",weight)
print("accuracy = ",best_model_accuracy)

```

```

coef :- [[ 0.00029657 -0.00026116  0.00037896  0.00029181 -0.00026116
 0.00030838
 0.0002199 ]]
accuracy = 0.4

```

0.4 step 4 Modifying original data

```

[7]: import random
      #print(X)
      X_=[]
      for i in range(len(X)):
          emp=[]
          for j in range(7):
              emp.append(X[i][j]+random.uniform(0.01, 0.09))
          X_.append(emp)
      X_=np.array(X_)
      for i in range(len(X_)):
          print(X_[i],X[i])

```

```

[-0.56308153  0.88729785 -0.94071359 -0.57249938  0.8626915  -0.59763304

```

-0.50720968] [-0.5810659 0.84183714 -1.01297765 -0.60402468 0.84183714
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 -0.15653725]
 [0.73973205 0.28961052 1.36498751 0.64010755 0.23943725 0.75167167
 0.5519498] [0.6719064 0.21203379 1.34439675 0.62369611 0.21203379 0.72364421
 0.5131026]
 [1.00580728 -0.61204774 0.82726455 1.00671635 -0.59174155 0.96863549
 -0.14884357] [0.98514947 -0.627704 0.78972042 0.95852905 -0.627704
 0.95802968
 -0.20345648]
 [1.34935132 -1.21337827 1.43037726 1.32882942 -1.20726855 1.40780192
 1.91324526] [1.29839254 -1.25750735 1.4137313 1.30452309 -1.25750735
 1.3447657
 1.88555113]
 [1.01549272 -0.59970929 1.29369894 0.9931277 -0.55549151 1.0620855
 2.29350979] [0.98514947 -0.627704 1.27506221 0.95852905 -0.627704
 1.01662605

2.24724411]
 [1.53506199 -0.38699311 1.07878519 1.507425 -0.40499235 1.50459707
 0.16185049] [1.45501408 -0.41776955 0.99772405 1.48170552 -0.41776955
 1.45400607
 0.09493325]
 [1.84911951 -0.13558177 1.12955974 1.86169453 -0.18219286 1.87781741
 0.17330375] [1.76825716 -0.2078351 1.06705859 1.84444121 -0.2078351
 1.78884246
 0.11498483]
 [2.10619635 -0.59172882 1.37029428 2.26101798 -0.54083973 2.20373993
 1.97684716] [2.08150023 -0.627704 1.34439675 2.21833799 -0.627704
 2.15883666
 1.94105101]
 [1.97920742 -0.12101248 1.52255149 2.04390707 -0.13176701 2.09590556
 1.41317927] [1.92487869 -0.2078351 1.48306584 2.02999446 -0.2078351
 2.00606756
 1.36780975]
 [0.87133255 -0.38277353 1.19770899 0.81830178 -0.39584756 0.92450219
 0.8884285] [0.82852793 -0.41776955 1.13639313 0.78971744 -0.41776955
 0.84795222
 0.87421373]
 [0.38362636 -1.00196258 0.50236199 0.38316808 -0.97142085 0.3985113
 -0.50543822] [0.35866332 -1.0475729 0.44304772 0.30002427 -1.0475729
 0.32351473
 -0.5474978]
 [0.12703394 -1.45529809 0.67606752 0.04993048 -1.42556741 0.09910731
 -0.30699105] [0.04542025 -1.4674418 0.65105134 -0.01248648 -1.4674418
 0.0673649
 -0.3302076]
 [0.12092223 -1.45516826 0.62481365 0.03501553 -1.38555935 0.07048835
 0.16984455] [0.04542025 -1.4674418 0.5817168 -0.01248648 -1.4674418
 0.05899399
 0.13804358]
 [0.54711221 -0.74980026 0.76782589 0.48938316 -0.80676705 0.57075441
 1.3432073] [0.51528486 -0.83763845 0.72038588 0.46046505 -0.83763845
 0.50139656
 1.28012377]
 [0.85005126 -0.76262747 1.622674 0.85399935 -0.8134911 0.93396845
 1.98538169] [0.82852793 -0.83763845 1.55240038 0.78971744 -0.83763845
 0.89817767
 1.94454083]
 [-0.02827715 -0.17045256 1.15157978 -0.14588926 -0.12905301 0.04236614
 0.12123408] [-0.11120129 -0.2078351 1.13639313 -0.16455644 -0.2078351
 -0.01090311
 0.07381134]
 [0.87581368 0.64861579 1.1475828 0.86991184 0.66857927 0.92244965
 1.06879168] [0.82852793 0.63190269 1.13639313 0.78971744 0.63190269 0.84795222
 1.02133893]

[1.93670547 0.05431544 1.28556218 2.07671852 0.02503867 2.06193804
 -0.06396527] [1.92487869 0.00209934 1.27506221 2.02999446 0.00209934
 1.98095483
 -0.09909819]
 [1.38374966 -1.59354384 1.13695798 1.34019838 -1.64833912 1.32861527
 0.99953446] [1.29839254 -1.67737625 1.06705859 1.30452309 -1.67737625
 1.30291115
 0.9307629]
 [0.25482837 -0.15513792 0.90766017 0.21700519 -0.18612263 0.24366338
 0.06881416] [0.20204178 -0.2078351 0.85905497 0.14237376 -0.2078351
 0.23185327
 -0.01875666]
 [0.12275932 -1.18107654 0.86447194 0.04327025 -1.2297115 0.10692258
 -0.25258491] [0.04542025 -1.25750735 0.78972042 -0.01248648 -1.25750735
 0.08410671
 -0.31966474]
 [0.06781951 -1.00482833 1.10734951 0.02833693 -1.03276811 0.15007155
 0.66834688] [0.04542025 -1.0475729 1.06705859 -0.01248648 -1.0475729
 0.11759035
 0.63341909]
 [1.01864628 -0.14834385 1.23117853 0.97072466 -0.1929102 1.03268431
 0.4858471] [0.98514947 -0.2078351 1.20572767 0.95852905 -0.2078351
 1.00825514
 0.47544708]
 [0.5373169 -1.02401988 0.82007292 0.49628739 -1.0339105 0.5977304
 0.26944368] [0.51528486 -1.0475729 0.78972042 0.46046505 -1.0475729
 0.50976747
 0.20230709]
 [-0.6762473 -1.62447585 0.38138623 -0.66314888 -1.6245395 -0.5981955
 1.38821349] [-0.73768744 -1.67737625 0.30437864 -0.74493354 -1.67737625
 -0.63369879
 1.3635154]
 [0.25282051 -0.79031325 1.01042478 0.20206487 -0.75514288 0.32554294
 1.3262981] [0.20204178 -0.83763845 0.92838951 0.14237376 -0.83763845
 0.24022418
 1.2549275]
 [0.40583303 -0.18581259 0.98702791 0.34605366 -0.17694149 0.43545722
 1.74803285] [0.35866332 -0.2078351 0.92838951 0.30002427 -0.2078351
 0.3821111
 1.71647571]
 [0.42248867 -0.33382809 0.96236021 0.32771642 -0.39443268 0.39377218
 1.85354436] [0.35866332 -0.41776955 0.92838951 0.30002427 -0.41776955
 0.3821111
 1.84126905]
 [1.1841147 -0.35583456 1.01284878 1.18091773 -0.3385647 1.21221649
 0.09694254] [1.14177101 -0.41776955 0.99772405 1.13013093 -0.41776955
 1.13758569
 0.01985967]

```

[-0.54714892 -1.23298173  0.13256532 -0.54957579 -1.17580039 -0.49113233
 0.63383262] [-0.5810659  -1.25750735  0.09637501 -0.60402468 -1.25750735
-0.53199223
 0.5905822 ]
[ 0.40172687 -0.57828017  0.90534251  0.3670125  -0.61256391  0.42501028
-0.01692131] [ 0.35866332 -0.627704    0.85905497  0.30002427 -0.627704
0.37374019
-0.04536372]

```

```

[8]: X_train_,X_test_,y_train,y_test=train_test_split(X_,Y,random_state = 42,
    ↪test_size=0.2) #adding error

```

```

best_model.fit(X_train_,y_train)
y_pred_=best_model.predict(X_test_)
best_model_accuracy_edited=accuracy_score(y_test, y_pred_)
print("best_model_accuracy_edited = ",best_model_accuracy_edited)
weight_1=best_model.coef_
print("coef :-",weight_1)

```

```

best_model_accuracy_edited = 0.4
coef :- [[ 0.00029688 -0.00026124  0.00037905  0.00029056 -0.00026117
0.00031002
 0.00021826]]

```

```

[9]: print("different in accuracy = ",best_model_accuracy_edited-best_model_accuracy)

```

```

different in accuracy = 0.0

```

0.5 step 5 Checking deviations in metric and weights

```

[11]: change=weight-weight_1
print("Absolute change in weight = ",change)

per_chan=[]
for i in range(7):
    per_chan.append(abs(change[:,i]/weight[:,i])*100)

print(per_chan)

top=[]
idx=[]
top_fe=[]
fe= label.copy()
for i in range(4):
    idx.append(per_chan.index(max(per_chan)))
    top.append(max(per_chan))
    per_chan.pop(idx[i])
    top_fe.append(fe[idx[i]])

```

```

fe.pop(idx[i])

print("Top features are = ",top_fe)
print("with change in weight with percentage =",list(top))

Absolute change in weight = [[-3.07476576e-07  8.20232349e-08 -8.99509811e-08
1.24712284e-06
1.15782681e-08 -1.64045231e-06  1.64688780e-06]]
[array([0.10367701]), array([0.03140726]), array([0.02373631]),
array([0.42737671]), array([0.0044334]), array([0.53195253]),
array([0.74891423])]
Top features are = ['w', '2*z+3*x*x', 'x*x', 'x']
with change in weight with percentage = [array([0.74891423]),
array([0.53195253]), array([0.42737671]), array([0.10367701])]

```

0.6 According to logistic regression ‘w’, ‘ $2z+3xx$ ’, ‘xx’, ‘x’ are the most important features by weight of that features.

1 Here we start LinearSVC

1.1 step 2

```

[32]: from sklearn.svm import SVC

parameters = {'C':(np.logspace(-5,8,10))}

cv = RepeatedStratifiedKFold(n_splits=10, n_repeats=3, random_state=42)

lr_svc=SVC(kernel='linear')
lr_svc_cv = GridSearchCV(lr_svc, parameters,n_jobs=-1,cv=cv)
lr_svc_cv.fit(X, Y) # fitting
    ↪ on whole data for find out best hyperparameter

print("best parameter = ",lr_svc_cv.best_params_)
print("best score = ",lr_svc_cv.best_score_)
print("best estimator = ",lr_svc_cv.best_estimator_)

best parameter = {'C': 0.007742636826811269}
best score = 1.0
best estimator = SVC(C=0.007742636826811269, kernel='linear')

```

1.2 step 3

```
[37]: best_model_1=SVC(kernel='linear',C=0.007742636826811269)

best_model_1.fit(X_train,y_train)           #fitting training data on
↳best_hyperparameter_model
y_pred_0=best_model_1.predict(X_test)
best_model_accuracy_1=accuracy_score(y_test, y_pred_0)

w_01=best_model_1.coef_
print("coef :-",w_01)
print("accuracy = ",best_model_accuracy_1)

coef :- [[ 0.16128427 -0.20643167  0.30599434  0.15069928 -0.20643167  0.1725741
 0.12110967]]
accuracy = 1.0
```

```
[36]: best_model_1.fit(X_train_,y_train)
y_pred_1=best_model_1.predict(X_test_)
best_model_accuracy_edited_1=accuracy_score(y_test, y_pred_1)
print("best_model_accuracy_edited = ",best_model_accuracy_edited_1)
w_12=best_model_1.coef_
print("coef :-",w_12)

best_model_accuracy_edited = 1.0
coef :- [[ 0.16009975 -0.20690403  0.3050217   0.15015462 -0.20592693
 0.17426045
 0.12377358]]
```

```
[38]: print("different in accuracy =
↳",best_model_accuracy_edited_1-best_model_accuracy_1)

different in accuracy = 0.0
```

1.3 step 5 Checking deviations in metric and weights with updated data set.

```
[39]: ch=[]

ch=abs(w_01-w_12)

print("Absolute change in weight = ",ch)

per_chan=[]
for i in range(7):
    per_chan.append(abs(ch[:,i]/w_01[:,i])*100)

print(per_chan)
```

```

top=[]
idx=[]
top_fe=[]
fe1=label.copy()
for i in range(4):
    idx.append(per_chan.index(max(per_chan)))
    top.append(max(per_chan))
    per_chan.pop(idx[i])
    top_fe.append(fe1[idx[i]])
    fe1.pop(idx[i])

print("Top features are = ",top_fe)
print("with change in weight with percentage =",list(top))

```

```

Absolute change in weight = [[0.00118452 0.00047236 0.00097264 0.00054465
0.00050473 0.00168635
0.00266391]]
[array([0.7344298]), array([0.22882139]), array([0.3178629]),
array([0.3614178]), array([0.24450327]), array([0.97717166]),
array([2.19958674])]
Top features are = ['w', '2*z+3*x*x', 'x', 'x*x']
with change in weight with percentage = [array([2.19958674]),
array([0.97717166]), array([0.7344298]), array([0.3614178])]

```

- 1.3.1 1.According to Linear SVM 'w', '2z+3xx', 'x', 'xx' are the most important features by weights.
- 1.3.2 2.After pertubation test there is not much change in accuracy and weight of feature of the both model in linear SVM and Logistic Regression.
- 1.3.3 3.Some freature are highly correlated to each other in correlation matrix.

[]: