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Python 3 Google Compute Engine backend (GPU)  
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System RAM  
1.4 / 12.7 GB



GPU RAM



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27.0 / 78.2 GB



```
!pip install ucimlrepo
from ucimlrepo import fetch_ucirepo
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
from sklearn import datasets
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
```

```
Collecting ucimlrepo
  Downloading ucimlrepo-0.0.3-py3-none-any.whl (7.0 kB)
Installing collected packages: ucimlrepo
Successfully installed ucimlrepo-0.0.3
```

```
# fetch dataset
spambase = fetch_ucirepo(id=94)
```

```
# data (as pandas dataframes)
X = spambase.data.features
y = spambase.data.targets
```

```
# variable information
print(spambase.variables)
```

0	None	None	no
1	None	None	no
2	None	None	no
3	None	None	no
4	None	None	no
5	None	None	no
6	None	None	no
7	None	None	no
8	None	None	no
9	None	None	no
10	None	None	no
11	None	None	no
12	None	None	no
13	None	None	no
14	None	None	no
15	None	None	no
16	None	None	no
17	None	None	no
18	None	None	no
19	None	None	no
20	None	None	no
21	None	None	no
22	None	None	no
23	None	None	no
24	None	None	no
25	None	None	no
26	None	None	no
27	None	None	no
28	None	None	no
29	None	None	no
30	None	None	no
31	None	None	no
32	None	None	no
33	None	None	no
34	None	None	no
35	None	None	no
36	None	None	no
37	None	None	no
38	None	None	no
39	None	None	no
40	None	None	no
41	None	None	no
42	None	None	no
43	None	None	no
44	None	None	no
45	None	None	no
46	None	None	no
47	None	None	no
48	None	None	no
49	None	None	no
50	None	None	no
51	None	None	no
52	None	None	no
53	None	None	no
54	None	None	no
55	None	None	no
56	None	None	no
57	spam (1) or not spam (0)	None	no

```
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state
```

```
Accuracy=[]  
regularization=[0,0.001,0.1,1,10,100]
```

```
print(y_train.shape)  
y_train = y_train.to_numpy().ravel()  
y_test = y_test.to_numpy().ravel()
```

```
(3680, 1)
```

```
def metrics(y_test,y_pred):  
    # Calculate accuracy  
    accuracy = accuracy_score(y_test, y_pred)  
  
    # Calculate precision  
    precision = precision_score(y_test, y_pred)  
  
    # Calculate recall  
    recall = recall_score(y_test, y_pred)  
  
    # Calculate F1-score  
    f1 = f1_score(y_test, y_pred)  
  
    print("Accuracy:", accuracy)  
    print("Precision:", precision)  
    print("Recall:", recall)  
    print("F1-Score:", f1)  
    Accuracy.append(accuracy)
```

## ▼ Without Regularization

```
svm_classifier = SVC(kernel='linear', random_state=42)  
svm_classifier.fit(X_train, y_train)  
y_pred = svm_classifier.predict(X_test)  
metrics(y_test,y_pred)
```

```
Accuracy: 0.9457111834961998  
Precision: 0.9393139841688655  
Recall: 0.9295039164490861  
F1-Score: 0.9343832020997375
```

## Regularization 0.001

```
svm_classifier = SVC(C=0.001, kernel='linear', random_state=42)  
svm_classifier.fit(X_train, y_train)  
y_pred = svm_classifier.predict(X_test)
```

```
metrics(y_test,y_pred)
```

```
Accuracy: 0.8849077090119435  
Precision: 0.9037900874635568  
Recall: 0.8093994778067886  
F1-Score: 0.8539944903581267
```

## Regularization 0.01

```
svm_classifier = SVC(C=0.1, kernel='linear', random_state=42)  
svm_classifier.fit(X_train, y_train)  
y_pred = svm_classifier.predict(X_test)
```

```
metrics(y_test,y_pred)
```

```
Accuracy: 0.9446254071661238  
Precision: 0.9414893617021277  
Recall: 0.9242819843342036  
F1-Score: 0.932806324110672
```

## Regularization 1

```
svm_classifier = SVC(C=1, kernel='linear', random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
```

```
metrics(y_test,y_pred)
```

```
Accuracy: 0.9457111834961998
Precision: 0.9393139841688655
Recall: 0.9295039164490861
F1-Score: 0.9343832020997375
```

## regularization 10

```
svm_classifier = SVC(C=10, kernel='linear', random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
```

```
metrics(y_test,y_pred)
```

```
Accuracy: 0.9424538545059717
Precision: 0.9319371727748691
Recall: 0.9295039164490861
F1-Score: 0.930718954248366
```

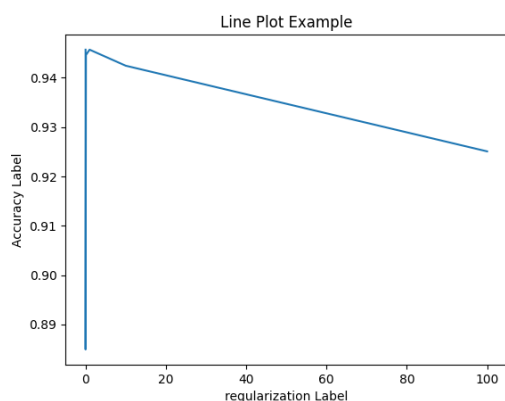
## regularization 100

```
svm_classifier = SVC(C=100, kernel='linear', random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
```

```
metrics(y_test,y_pred)
```

```
Accuracy: 0.9250814332247557
Precision: 0.9005102040816326
Recall: 0.9216710182767625
F1-Score: 0.9109677419354839
```

```
plt.plot(regularization, Accuracy)
plt.xlabel('regularization Label')
plt.ylabel('Accuracy Label')
plt.title('Line Plot Example')
plt.show()
```



## kernel Tricks

```
Accuracy=[]
x_labels=['poly degree 2','poly degree 3','sigmoid','RBF']
```

```
svm_classifier = SVC(C=1, kernel='poly',degree = 2, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
```

```
Accuracy: 0.6579804560260586
Precision: 0.8695652173913043
Recall: 0.20887728459530025
F1-Score: 0.3368421052631579
```

```
svm_classifier = SVC(C=1, kernel='poly',degree = 3, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
```

```
Accuracy: 0.6362649294245385
Precision: 0.9615384615384616
Recall: 0.13054830287206268
F1-Score: 0.2298850574712644
```

```
svm_classifier = SVC(C=1, kernel='sigmoid', gamma=0.1, coef0=0.5, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
```

```
Accuracy: 0.5331161780673181
Precision: 0.02040816326530612
Recall: 0.0026109660574412533
F1-Score: 0.00462962962962963
```

```
svm_classifier = SVC(C=1, kernel='rbf', random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
```

```
Accuracy: 0.7176981541802389
Precision: 0.7530864197530864
Recall: 0.47780678851174935
F1-Score: 0.584664536741214
```

## Part C

```
Accuracy=[]
training_accuracy=[]

x_labels=['poly degree 1 c=0.01','poly degree 1 c=100','poly degree 3 c=0.01','poly d
```

```
svm_classifier = SVC(C=0.01, kernel='poly',degree = 1, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
y_pred = svm_classifier.predict(X_train)
training_accuracy.append(accuracy_score(y_train, y_pred))
```

```
Accuracy: 0.6471226927252985
Precision: 0.8152173913043478
Recall: 0.195822454308094
F1-Score: 0.3157894736842105
```

```
svm_classifier = SVC(C=100, kernel='poly',degree = 1, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
y_pred = svm_classifier.predict(X_train)
training_accuracy.append(accuracy_score(y_train, y_pred))
```

```
Accuracy: 0.7861020629750272
Precision: 0.8690476190476191
Recall: 0.5718015665796344
F1-Score: 0.6897637795275591
```

```
svm_classifier = SVC(C=0.01, kernel='poly',degree = 3, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
y_pred = svm_classifier.predict(X_train)
training_accuracy.append(accuracy_score(y_train, y_pred))
```

```
Accuracy: 0.6232356134636265
Precision: 0.95
Recall: 0.09921671018276762
F1-Score: 0.1796690307378605
```

```
svm_classifier = SVC(C=100, kernel='poly',degree = 3, random_state=42)
svm_classifier.fit(X_train, y_train)
y_pred = svm_classifier.predict(X_test)
metrics(y_test,y_pred)
y_pred = svm_classifier.predict(X_train)
training_accuracy.append(accuracy_score(y_train, y_pred))
```

```
Accuracy: 0.6829533116178067
Precision: 0.9690721649484536
Recall: 0.2454308093994778
F1-Score: 0.39166666666666666
```

```
plt.plot(x_labels, Accuracy,'green')
plt.plot(x_labels, training_accuracy,'orange')
plt.xlabel('parameter Label')
plt.ylabel('Accuracy Label')
plt.title('Line Plot Example')
plt.show()
```

