CS 6375.003 Fall 24 - MACHINE LEARNING PROJECT 1 - SPAM & HAM CLASSIFIER

Experiments

- 1. Multinomial Naive Bayes for Bag of Words dataset
- 2. Discrete Naive Bayes for Bernoulli dataset
- 3. Logistic Regression for Bag of Words dataset
- 4. Logistic Regression for Bernoulli dataset
- 5. Stochastic Gradient Descent for Bag of Words dataset
- 6. Stochastic Gradient Descent for Bernoulli dataset

MULTINOMIAL NAIVE BAYES

Scikit learn approach results -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	92.98%	86.57%	88.96%	91.48%
Enron 2	94.56%	87.69%	89.76%	91.93%
Enron 4	97.42%	99.48%	98.23%	97%

∫ v ei	nron1					
_		929824561403	5088			
		577181208053				
		889655172413				
- '	I Score - 0.			£1 ccana	cuppont	
		precision	recall	T1-Score	Support	
	ham	0.94				
	spam	0.91	0.87	0.89	149	
	accuracy			0.93	456	
	macro avg	0.93	0.91	0.92	456	
We	eighted avg	0.93	0.93	0.93	456	
AI	nron2					
		945606694560	6605			
		692307692307				
10	ı score - 0.	897637795275				
		precision	recall	†1-score	support	
	ham	0.95			348	
	spam	0.92	0.88	0.90	130	
	accuracy			0.95	478	
	macro avg	0.94	0.92	0.93	478	
We	eighted avg		0.95	0.95		
•		0.33	0.55	0.55	.,,	
_						
01	 nron4					
		974217311233	0000			
		488491048593				
#:	ı score - 0.	982323232323		5.		
		precision	recall	t1-score	support	
	ham	0.99	0.92	0.95	152	
	spam	0.97	0.99	0.98	391	
	accuracy			0.97	543	
	macro avg	0.98	0.96			
1.4	eighted avg		0.97	0.97	543	
DV C	ergiiceu avg	0.57	0.57	0.57	242	

Step by step approach results -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	93.20%	87.24%	89.34%	91.54%
Enron 2	94.35%	87.69%	89.41%	91.2%
Enron 4	97.23%	99.48%	98.10%	96.76%

BERNOULLI NAIVE BAYES

Scikit learn approach results -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	73.02%	20.80%	33.51%	97%
Enron 2	77.82%	20.76%	33.75%	97%
Enron 4	91.71%	100%	94.55%	97%

~	onnon1					
	enron1					
		0.730263157				
	recall - 0	0.20805369127	51678			
	f1 score -	- 0.335135135	1351351			
		precisio	n recall	f1-score	support	
	h	nam 0.7	ว ค จุด	0.83	307	
		oam 0.8				
	οh	Jaill 0.0	0.21	0.34	149	
				0.73	45.5	
	accura	асу		0.73		
	macro a	avg 0.7	9 0.60	0.58		
	weighted a	avg 0.7	7 0.73	0.67	456	
	enron2					
		- 0.778242677	8242678			
		0.20769230769				
	f1 score -		23077			
	TI Score -			-		
		precisio	n recall	†1-score	support	
	h	nam 0.7	7 0.99	0.87	348	
	sp	oam 0.9	0.21	0.34	130	
	accura	acv		0.78	478	
	macro a	avg 0.8	4 9.69			
		avg 0.8				
	weighten a	wg 0.6	0.78	0.72	4/0	
	enron4					
		0.917127071	8232044			
	recall - 1					
	f1 score -	0.945586457	9737605			
		precisio	n recall	f1-score	support	
		nam 1.0	9 0.70	0.83	152	
			1.00			
	sp	oam 0.9	1.00	0.95	391	
	accura			0.92		
		avg 0.9				
	weighted a	avg 0.9	0.92	0.91	543	

Step by step approach results -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	73.02%	20.80%	33.51%	86.11%
Enron 2	77.82%	20.76%	33.75%	90%
Enron 4	91.71%	100%	94.55%	89.67%

```
enron1
Accuracy = 73.02631578947368%
Recall = 0.2080536912751678
F1 = 0.3351351351351351

enron2
Accuracy = 77.82426778242679%
Recall = 0.2076923076923077
F1 = 0.3375

enron4
Accuracy = 91.71270718232044%
Recall = 1.0
F1 = 0.9455864570737605
```

LOGISTIC REGRESSION -

Scikit learn approach Bag of Words results -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	95.17%	95.97%	92.85%	97%
Enron 2	95.39%	90.76%	91.47%	97%
Enron 4	95.21%	99.74%	96.77%	97%

enron1 accuracy - 0.9517543859649122	
<u> </u>	
型 recall - 0.959731543624161	
f1 score - 0.9285714285714286	
precision recall f1-score support	
ham 0.98 0.95 0.96 307	
spam 0.90 0.96 0.93 149	
accuracy 0.95 456	
macro avg 0.94 0.95 0.95 456	
weighted avg 0.95 0.95 0.95 456	
enron2	
accuracy - 0.9539748953974896	
recall - 0.9076923076923077	
f1 score - 0.9147286821705427	
precision recall f1-score support	
ham 0.97 0.97 0.97 348	
spam 0.92 0.91 0.91 130	
accuracy 0.95 478	
macro avg 0.94 0.94 0.94 478	
weighted avg 0.95 0.95 0.95 478	
enron4	
accuracy - 0.9521178637200737	
recall - 0.9974424552429667	
f1 score - 0.9677419354838709	
precision recall f1-score support	
ham 0.99 0.84 0.91 152	
spam 0.94 1.00 0.97 391	
accuracy 0.95 543	
macro avg 0.97 0.92 0.94 543	
weighted avg 0.95 0.95 0.95 543	

Hyperparameter tuned BoW -

	Dataset	Parameters	Accuracy	Recall	F1 Score	Precision Score
0	enron1	$\label{eq:conditional} \begin{tabular}{ll} \$	0.942982	0.926174	0.913907	0.901961
1	enron2	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': False}	0.926778	0.792308	0.854772	0.927928
2	enron4	$\label{eq:condition} \begin{tabular}{ll} \be$	0.955801	1.000000	0.970223	0.942169
3	enron1	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.942982	0.926174	0.913907	0.901961
4	enron2	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.926778	0.792308	0.854772	0.927928
5	enron4	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.955801	1.000000	0.970223	0.942169
6	enron1	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.942982	0.926174	0.913907	0.901961
7	enron2	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.920502	0.769231	0.840336	0.925926
8	enron4	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.942910	1.000000	0.961870	0.926540
9	enron1	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.947368	0.932886	0.920530	0.908497
10	enron2	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.914226	0.746154	0.825532	0.923810
11	enron4	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.942910	1.000000	0.961870	0.926540

		Dataset	Parameters	Accuracy	Recall	F1 Score	Precision Score
	0	enron1	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': False}	0.945175	0.872483	0.912281	0.955882
	1	enron2	$\label{eq:conditional} \begin{tabular}{ll} \$	0.903766	0.715385	0.801724	0.911765
	2	enron4	$\label{eq:condition} \begin{tabular}{ll} \be$	0.950276	1.000000	0.966625	0.935407
	3	enron1	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.945175	0.872483	0.912281	0.955882
	4	enron2	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.903766	0.715385	0.801724	0.911765
	5	enron4	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': False}	0.950276	1.000000	0.966625	0.935407
	6	enron1	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.942982	0.865772	0.908451	0.955556
	7	enron2	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.887029	0.653846	0.758929	0.904255
	8	enron4	{'C': 0.1, 'solver': 'liblinear', 'penalty': 'l2', 'fit_intercept': True}	0.946593	1.000000	0.964242	0.930952
	9	enron1	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.936404	0.852349	0.897527	0.947761
	10	enron2	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.887029	0.653846	0.758929	0.904255
	11	enron4	{'C': 0.1, 'solver': 'lbfgs', 'penalty': 'l2', 'fit_intercept': True}	0.942910	1.000000	0.961870	0.926540

To achieve better results on the enron datasets using Logistic Regression, a hyperparameter tuning process was done. The emphasis was given on 4 hyperparameters - Regularizing constant C, Solvers, Penalty, fit_intercept

Regularizing constant C - Controls the strength of regularization. It has an inverse behavior, i.e. larger value weakens the effect of the regularizer and smaller value strengthens the effect of the regularizer.

Solvers - Algorithm used for optimization of the algorithm. The default value is 'lbfgs'. As the dataset was small, the experiment was carried out using 'liblinear' as it tends to work well with smaller datasets.

Penalty - Defines the type of penalty or regularization to use. For all the datasets, I2 regularizer worked the best.

Fit intercept - Decides whether to add a bias term or not.

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	93.85%	87.91%	90.03%	92.90%
Enron 2	90.37%	70.76%	80%	92%
Enron 4	94.29%	100%	96.18%	92.65%

```
λ: 0.0001, Validation Accuracy: 0.9333333333333333
    λ: 0.001, Validation Accuracy: 0.9333333333333333
    λ: 0.01, Validation Accuracy: 0.9407407407407408
    λ: 0.1, Validation Accuracy: 0.888888888888888888
    Best λ: 0.01
    enron1
    Accuracy = 93.85964912280701%
    Recall = 0.8791946308724832
    F1 = 0.9034482758620689
    λ: 1e-05, Validation Accuracy: 0.9136690647482014
    λ: 0.0001, Validation Accuracy: 0.9136690647482014
    λ: 0.001, Validation Accuracy: 0.9136690647482014
    λ: 0.01, Validation Accuracy: 0.9136690647482014
    λ: 0.1, Validation Accuracy: 0.8920863309352518
    Best λ: 1e-05
    enron2
    Accuracy = 90.3765690376569%
    Recall = 0.7076923076923077
    F1 = 0.8
    λ: 1e-05, Validation Accuracy: 0.9192546583850931
    λ: 0.0001, Validation Accuracy: 0.9192546583850931
    λ: 0.001, Validation Accuracy: 0.9192546583850931
    λ: 0.01, Validation Accuracy: 0.9192546583850931
    λ: 0.1, Validation Accuracy: 0.906832298136646
    Best λ: 1e-05
    enron4
    Accuracy = 94.29097605893186%
    Recall = 1.0
    F1 = 0.9618696186961869
```

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	95.61%	93.28%	93.28%	97%
Enron 2	94.35%	85.38%	89.15%	97%
Enron 4	95.21%	100%	96.78%	97%

0	enron1					
		9.956140350877	193			
∓÷		93288590604026				
_		3.9328859064020 3.932885906040				
	TI Score - 6			Ca		
		precision	recall	+1-score	support	
	han					
	span	n 0.93	0.93	0.93	149	
	accuracy	/		0.96		
	macro avg	g 0.95	0.95	0.95	456	
	weighted ava	0.96	0.96	0.96	456	
		•				
	enron2					
		9.943514644351	4645			
		35384615384615				
		3.891566265060				
	11 Score - 6			£1 ssans	cuppost	
		precision	recarr	11-2Core	Support	
	han		0.98			
	span	n 0.93	0.85	0.89	130	
	accuracy			0.94		
	macro avg	g 0.94	0.92	0.93	478	
	weighted ave	g 0.94	0.94	0.94	478	
	enron4					
	accuracy - 6	9.952117863720	0737			
	recall - 1.0					
		, 9.967821782178	2179			
	11 30010 0	precision		f1-score	support	
		Pi ccision	I CCGII	71 30016	заррог с	
	han	1 00	0.02	0.01	152	
					152	
	span	0.94	1.00	0.97	391	
	accuracy			0.95	543	
	macro avg				543	
	weighted ave	g 0.96	0.95	0.95	543	

Step by step approach results -

Dataset	Accuracy	Recall score	F1 score
Enron 1	91.88%	80.53%	93.75%
Enron 2	88.07%	64.61%	88.4%
Enron 4	93.73%	100%	92%

```
→ λ: 1e-05, Validation Accuracy: 0.9259259259259259
    λ: 0.0001, Validation Accuracy: 0.9259259259259259
    λ: 0.001, Validation Accuracy: 0.9259259259259259
    λ: 0.01, Validation Accuracy: 0.9185185185185185
    λ: 0.1, Validation Accuracy: 0.8592592592592593
    Best λ: 1e-05
    enron1
    Accuracy = 91.8859649122807%
    Recall = 0.8053691275167785
    F1 = 0.8664259927797834
    λ: 1e-05, Validation Accuracy: 0.9136690647482014
    λ: 0.0001, Validation Accuracy: 0.9136690647482014
    λ: 0.001, Validation Accuracy: 0.9136690647482014
    λ: 0.01, Validation Accuracy: 0.9136690647482014
    λ: 0.1, Validation Accuracy: 0.8776978417266187
    Best λ: 1e-05
    enron2
    Accuracy = 88.07531380753139%
    Recall = 0.6461538461538462
    F1 = 0.74666666666666
    λ: 1e-05, Validation Accuracy: 0.906832298136646
    λ: 0.0001, Validation Accuracy: 0.906832298136646
    λ: 0.001, Validation Accuracy: 0.906832298136646
    λ: 0.01, Validation Accuracy: 0.906832298136646
    λ: 0.1, Validation Accuracy: 0.8819875776397516
    Best λ: 1e-05
    enron4
    Accuracy = 93.73848987108656%
    Recall = 1.0
    F1 = 0.9583333333333333333
```

STOCHASTIC GRADIENT DESCENT

Scikit learn Bag of Words approach -

For enron 1 - Best Parameters: {'alpha': 0.001, 'early_stopping': False, 'learning_rate': 'optimal', 'loss': 'modified_huber', 'max_iter': 1000, 'penalty': None, 'validation fraction': 0.2, 'warm start': False}

For enron 2 - Best Parameters: {'alpha': 0.01, 'early_stopping': False, 'learning_rate': 'optimal', 'loss': 'hinge', 'max_iter': 800, 'penalty': 'l2', 'validation fraction': 0.3, 'warm start': True}

For enron 4 - Best Parameters: {'alpha': 0.0001, 'early_stopping': False, 'learning_rate': 'optimal', 'loss': 'perceptron', 'max_iter': 1000, 'penalty': None, 'validation_fraction': 0.3, 'warm_start': True}

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	91.22%	80.91%	86.48%	93.75%
Enron 2	94.97%	93.07%	90.97%	88.42%
Enron 4	95.76%	96.93%	97.05%	92%

Scikit learn Bernoulli approach -

Dataset	Accuracy	Recall score	F1 score	Precision score
Enron 1	97.07%	96.92%	94.93%	93.75%
Enron 2	94.97%	93.07%	90.97%	88.42%
Enron 4	96.66%	98.20%	97.77%	92%

POST FEATURE ENGINEERING -

To improve the performance of the models, the following features were added -

- Word count
- Average word length
- Number of nouns
- Number of adjectives
- Number of verbs
- Number of special characters
- Number of numeric characters

These continuous variables were first discretized using binning. Bins were created based on quantiles and divided into 4 groups - Low, Medium, High and Very High. These categorical variables were then encoded using Scikit Learn's Label encoder.

RESULTS

1. Multinomial Naive Bayes (BoW)

Dataset	Accuracy	Recall score	F1 score
Enron 1	92.10%	83.89%	87.41%
Enron 2	93.10%	82.30%	86.63%
Enron 4	97.23%	99.48%	98.10%

```
enron1
     accuracy - 0.9210526315789473
     recall - 0.8389261744966443
     f1 score - 0.8741258741258742
                    precision recall f1-score support
              ham 0.92 0.96 0.94
spam 0.91 0.84 0.87
                                                            307
                                                              149
    accuracy 0.92 456
macro avg 0.92 0.90 0.91 456
weighted avg 0.92 0.92 456
     (463, 8469) (478, 8469)
     (463, 8476) (478, 8476)
     enron2
     accuracy - 0.9309623430962343
     recall - 0.823076923076923
     f1 score - 0.8663967611336032
                    precision recall f1-score support
               ham 0.94 0.97 0.95 348
spam 0.91 0.82 0.87 130
              spam
    accuracy 0.93 478
macro avg 0.93 0.90 0.91 478
weighted avg 0.93 0.93 0.93 478
     (535, 15535) (543, 15535)
     (535, 15542) (543, 15542)
     enron4
     accuracy - 0.9723756906077348
     recall - 0.9948849104859335
     f1 score - 0.9810844892812105
                    precision recall f1-score support

        ham
        0.99
        0.91
        0.95
        152

        spam
        0.97
        0.99
        0.98
        391

        accuracy 0.97 543
macro avg 0.98 0.95 0.96 543
ighted avg 0.97 0.97 543
     weighted avg
```

Observation - Slight decrease in performance

2. Multinomial Naive Bayes (Bernoulli)

Dataset	Accuracy	Recall score	F1 score
Enron 1	73.02%	20.80%	33.51%
Enron 2	77.82%	20.76%	33.75%
Enron 4	91.71%	100%	94.55%

```
[43] enron1
      accuracy - 0.7302631578947368
 ₹ recall - 0.2080536912751678
      f1 score - 0.3351351351351351
                      precision recall f1-score support
                ham 0.72 0.98 0.83
spam 0.86 0.21 0.34
                spam
                                                                  149
      accuracy 0.73 456
macro avg 0.79 0.60 0.58 456
weighted avg 0.77 0.73 0.67 456
      (463, 8469) (478, 8469)
      (463, 8476) (478, 8476)
      enron2
      accuracy - 0.7782426778242678
      recall - 0.2076923076923077
      f1 score - 0.3375
                     precision recall f1-score support
                ham 0.77 0.99 0.87 348
spam 0.90 0.21 0.34 130

      accuracy
      0.78
      478

      macro avg
      0.84
      0.60
      0.60
      478

      weighted avg
      0.81
      0.78
      0.72
      478

      (535, 15535) (543, 15535)
(535, 15542) (543, 15542)
      enron4
      accuracy - 0.9171270718232044
      recall - 1.0
      f1 score - 0.9455864570737605
                      precision recall f1-score support
                      1.00 0.70 0.83
0.90 1.00 0.95
                spam
           accuracy
                                                    0.92
      macro avg 0.95 0.85 0.89 543 weighted avg 0.93 0.92 0.91 543
```

Observation - No change in the results

3. LR Bag of Words

Dataset	Accuracy	Recall score	F1 score
Enron 1	94.95%	95.97%	92.55%
Enron 2	95.39%	91.53%	91.53%
Enron 4	95.02%	99.74%	96.65%

accuracy - 0.9495614935087719 recall - 0.959731543624161 f1 score - 0.9255663430420711	[44]	enron1					
f1 score - 0.9255663430420711	, [44]	accuracy - 0.9					
f1 score - 0.9255663430420711	∑ ₹						
ham 0.98 0.94 0.96 307 spam 0.89 0.96 0.93 149 accuracy 0.95 0.95 456 macro avg 0.94 0.95 0.95 456 weighted avg 0.95 0.95 0.95 456 (463, 8469) (478, 8469) (463, 8476) (478, 8476) enron2 accuracy - 0.9539748953974896 recall - 0.9153846153846154 f1 score - 0.9153846153846154 precision recall f1-score support ham 0.97 0.97 0.97 348 spam 0.92 0.92 0.92 130 accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679 precision recall f1-score support ham 0.99 0.83 0.90 152 spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543		f1 score - 0.9					
spam 0.89 0.96 0.93 149 accuracy 0.95 456 macro avg 0.94 0.95 0.94 456 weighted avg 0.95 0.95 0.95 456 (463, 8469) (478, 8469) (463, 8476) (478, 8476) enron2 accuracy - 0.9539748953974896 recall - 0.9153846153846154 f1 score - 0.9153846153846154 precision recall f1-score support ham 0.97 0.97 0.97 348 spam 0.92 0.92 0.92 130 accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9562762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679 precision recall f1-score support ham 0.99 0.83 0.90 152 spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543			precision	recall	f1-score	support	
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accuracy							
macro avg		spam	0.89	0.96	0.93	149	
macro avg		accupacy			9 05	456	
weighted avg 0.95 0.95 0.95 456 (463, 8469) (478, 8469) (463, 8476) (478, 8476) enron2 accuracy - 0.9539748953974896 recall - 0.9153846153846154 fl score - 0.9153846153846154			a 94	a 05			
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(463, 8476) (478, 8476) enron2 accuracy - 0.9539748953974896 recall - 0.9153846153846154 f1 score - 0.9153846153846154 precision recall f1-score support ham 0.97 0.97 0.97 348 spam 0.92 0.92 130 accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679 precision recall f1-score support ham 0.99 0.83 0.90 152 spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543							
enron2 accuracy - 0.9539748953974896 recall - 0.9153846153846154 f1 score - 0.9153846153846154		(463, 8469) (4	178, 8469)				
accuracy - 0.9539748953974896 recall - 0.9153846153846154 f1 score - 0.9153846153846154		(463, 8476) (4	178, 8476)				
recall - 0.9153846153846154 f1 score - 0.9153846153846154							
f1 score - 0.9153846153846154							
precision recall f1-score support ham 0.97 0.97 0.97 348 spam 0.92 0.92 0.92 130 accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679							
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accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478		hom	0.07	0.07	0.07	240	
accuracy 0.95 478 macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478							
macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679		Shaiii	0.92	0.92	0.92	130	
macro avg 0.94 0.94 0.94 478 weighted avg 0.95 0.95 0.95 478 (535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679		accuracy			0.95	478	
weighted avg 0.95 0.95 0.95 478			0.94	0.94			
(535, 15535) (543, 15535) (535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679		_					
(535, 15542) (543, 15542) enron4 accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679							
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accuracy - 0.9502762430939227 recall - 0.9974424552429667 f1 score - 0.9665427509293679			(543, 15542)				
recall - 0.9974424552429667 f1 score - 0.9665427509293679			NEO 27 C 2 4 2 0 0 2 C	1227			
f1 score - 0.9665427509293679							
precision recall f1-score support ham 0.99 0.83 0.90 152 spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543							
ham 0.99 0.83 0.90 152 spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543		II Score - 0.9			f1-scope	support	
spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543			pi ecision	recarr	11-30016	Suppor C	
spam 0.94 1.00 0.97 391 accuracy 0.95 543 macro avg 0.96 0.91 0.93 543		ham	0.99	0.83	0.90	152	
accuracy 0.95 543 macro avg 0.96 0.91 0.93 543							
macro avg 0.96 0.91 0.93 543							
_		accuracy			0.95	543	
weighted avg 0.95 0.95 0.95 543		macro avg	0.96	0.91	0.93	543	
		weighted avg	0.95	0.95	0.95	543	

Observation - Slight decrease in performance

4. LR using Bernoulli

Dataset	Accuracy	Recall score	F1 score
Enron 1	95.83%	93.28%	93.60%
Enron 2	94.76%	87.69%	90.11%
Enron 4	95.40%	100%	96.90%

	enron1					
[45]	accuracy - 0	.958333333333	33334			
<u>.</u> ∑ •	recall - 0.9	3288590604020	584			
<u> </u>	f1 score - 0					
		precision	recall	f1-score	support	
	ham	0.97	0.97	0.97	307	
	spam				149	
	accuracy			0.96	456	
	macro avg				456	
	weighted avg	0.96	0.96	0.96	456	
	(463, 8469)	(478, 8469)				
	(463, 8476) enron2	(478, 8476)				
	accuracy - 0					
	recall - 0.8					
	f1 score - 0			£1 scana	cuppont	
		precision	recall	T1-Score	Support	
	ham	0.95	0.97	0.96	348	
	spam	0.93	0.88	0.90	130	
				2.25		
	accuracy		0.93	0.95 0.93	478 478	
	macro avg weighted avg				478 478	
	(535, 15535)					
	(535, 15542) enron4	(543, 15542)			
	accuracy - 0	.953959484346	52247			
	recall - 1.0					
	f1 score - 0	.96902106567	53407			
		precision	recall	f1-score	support	
	ham	1.00	0.84	0.91	152	
	spam		1.00		391	
	accuracy			0.95	543	
	macro avg				543	
	weighted avg	0.96	0.95	0.95	543	

Observation - Slight increase in the performance

5. SGDClassifier BoW

Dataset	Accuracy	Recall score	F1 score
Enron 1	90.57%	88.59%	85.99%
Enron 2	95.18%	94.61%	91.44%
Enron 4	95.58%	97.18%	96.93%

```
[54] enron1
      accuracy - 0.9057017543859649
 ⊋ recall - 0.8859060402684564
      f1 score - 0.8599348534201955
                    precision recall f1-score support

        ham
        0.94
        0.92
        0.93
        307

        spam
        0.84
        0.89
        0.86
        149

     accuracy 0.91 456
macro avg 0.89 0.90 0.89 456
weighted avg 0.91 0.91 0.91 456
      (463, 8469) (478, 8469)
      (463, 8476) (478, 8476)
      accuracy - 0.9518828451882845
      recall - 0.9461538461538461
      f1 score - 0.9144981412639406
                     precision recall f1-score support
               ham0.980.950.97348spam0.880.950.91130
               spam
     accuracy 0.95 478
macro avg 0.93 0.95 0.94 478
weighted avg 0.95 0.95 0.95 478
      (535, 15535) (543, 15535)
      (535, 15542) (543, 15542)
      accuracy - 0.9558011049723757
      recall - 0.9718670076726342
      f1 score - 0.9693877551020408
                    precision recall f1-score support
               ham 0.93 0.91 0.92 152 spam 0.97 0.97 0.97 391
                                                 0.96
          accuracy
         accuracy 0.95
macro avg 0.95 0.94 0.94
ighted avg 0.96 0.96 0.96
                                      0.94 0.94
0.96 0.96
                                                               543
      weighted avg
```

Observation - For enron 1 and 4 recall score slightly increased. For enron2 there was slight improvement in all the metrics.

6. SGDClassifier Bernoulli -

Dataset	Accuracy	Recall score	F1 score
Enron 1	92.54%	89.26%	88.66%
Enron 2	93.72%	82.30%	87.70%
Enron 4	97.05%	99.23%	97.97%

2 ₅ [55]	enron1 accuracy - 0.9254385964912281 recall - 0.8926174496644296						
₹	recall - 0.89261 f1 score - 0.886						
				f1-score	support		
	ham	0.95	0.94		307		
	spam	0.88	0.89	0.89	149		
	accuracy			0.93	456		
	macro avg	0.91	0.92				
	weighted avg	0.93	0.93				
	66						
	(463, 8469) (478 (463, 8476) (478 enron2						
	accuracy - 0.9372384937238494 recall - 0.823076923076923						
	f1 score - 0.877			C4			
	pr	ecision	recall	f1-score	support		
	ham	0.94	0.98	0.96	348		
	spam	0.94	0.82	0.88	130		
				0.04	470		
	accuracy	0.04	0.00	0.94			
	macro avg weighted avg	0.94 0.94	0.90 0.94				
	weighten avg	0.94	0.94	0.94	4/0		
	(535, 15535) (54 (535, 15542) (54						
	enron4 accuracy - 0.9705340699815838 recall - 0.9923273657289002 f1 score - 0.9797979797979						
	pr	ecision	recall	f1-score	support		
	ham	0.98	0.91	0.95	152		
	spam	0.97	0.99	0.98	391		
				0.07	F 4.3		
	accuracy macro avg	0.97	0.95	0.97 0.96	543 543		
	weighted avg	0.97	0.95		543 543		
	mergineed diff	0137	0137	0.57	3.5		

Observation - Slight increase in performance for enron4 but significant decrease in performance for enron 1 and 2.

Answer the following questions:

1. Which data representation and algorithm combination yields the best performance (measured in terms of the accuracy, precision, recall and F1 score) and why?

For accuracy, precision and f1 score, the Bag of Words (BoW) representation with Multinomial Naive Bayes yields the best performance. However, in terms of Recall score the Bernoulli representation with the Discrete Naive Bayes gives the best results.

BoW tends to work better than the Bernoulli representation because it captures the frequency of the token in the document, providing more information about the importance of the word. The idea here is that the higher the frequency of the token the more is the importance. Bernoulli representation, on the other hand, only indicates the presence of a token which is not enough to capture the context of data to classify a document.

The dataset provided is small in size due to which Multinomial Naive Bayes works better because of its generative nature. It takes into consideration the prior probabilities which is useful in case of smaller datasets. In contrast, Logistic Regression generally requires larger datasets to perform optimally, as it is a discriminative model that needs sufficient data to learn complex decision boundaries.

2. Does Multinomial Naive Bayes perform better (again performance is measured in terms of the accuracy, precision, recall and F1 score) than LR and SGDClassifier on the Bag of words representation? Explain your yes/no answer.

Multinomial Naive Bayes (MNB) works generally better than LR and SGDClassifier while working with Bag of Words. MNB tends to work well with discrete values which fits with the BoW representation. While MNB assumes independence between features (a strong assumption) it is suitable for higher

dimensional data. LR and SGDClassifier assume a linear relationship to draw a decision boundary between the classes which may not be ideal for the BoW representation. MNB also takes into consideration the prior probabilities which is useful when there is not a lot of data, hence it works well with smaller datasets as compared to Logistic Regression which needs large amounts of data.

3. Does Discrete Naive Bayes perform better (again performance is measured in terms of the accuracy, precision, recall and F1 score) than LR and SGDClassifier on the Bernoulli representation? Explain your yes/no answer.

Yes, Discrete Naive Bayes (DNB) performs better than the LR and SGDClassifier on the Bernoulli Representation. DNB assumes independence between the data points which fits well with the Bernoulli representation of the data. The Bernoulli representation indicates the presence of a word in the document. While the assumption of independence is generally strong (and rarely holds true in real-world scenarios), it tends to work effectively in high-dimensional data, such as text classification, where feature dependencies may not be as critical. DNB can handle sparse datasets better than LR and SGD. The Bernoulli representation provides sparse data as most of the words are absent in a given document. LR and SGDClassifier assume a linear relationship between features and class labels which may not be suitable for the Bernoulli representation of the data.

4. Does your LR implementation outperform the SGDClassifier (again performance is measured in terms of the accuracy, precision, recall and F1 score) or is the difference in performance minor? Explain your yes/no answer.

Yes LR outperforms SGDClassfier. The Logistic Regression algorithm uses Gradient Ascent for updating the weights where the entire dataset is used for calculating the gradients. On the other hand, the SGD Classifier uses Stochastic Gradient Descent which arbitrarily selects a point from the dataset to calculate the gradient. Due to this behavior, SGD is computationally efficient and converges faster compared to LR but introduces a lot of noise in the calculation of gradients. LR is smooth with its convergence and does not introduce a lot of errors while updating the gradients.