

ASSIGNMENT 5

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Analysis Report

QUESTION 1:-

AIM:-

- 1. Build Naive Bayes Classifiers and KNN classifiers for Text Classification from scratch**
- 2. Use TF-TDF and Mutual Inforamtion for Feature Selection**

Dataset used are:-

- 20_newsgroups/sci.med/59199
- 20_newsgroups/comp.graphics/38703
- 20_newsgroups/rec.sport.hockey/53700
- 20_newsgroups/talk.politics.misc/178450
- 20_newsgroups/sci.space/60196
- 5000

Labels:- {'rec.sport.hockey': 511, 'sci.space': 508, 'sci.med': 504, 'talk.politics.misc': 493, 'comp.graphics': 484}

Total Documents:-5000

Tools Used:-

- nltk
- matplotlib
- numpy ,pandas

TRAIN-TEST SPLIT:-50:50,70:30,80:20

Graphs and Observations:-

NAIVE BAYES CLASSIFIER

Classifier's Name	Feature Selection Technique	% of features selected	Train-Test Split	Accuracy	Confusion Matrix
Naive bayes	TF-IDF	5%	50-50	97.96	[[483 11 0 2 4] [2 486 0 1 2] [2 1 491 2 1] [1 1 1 487 4] [3 11 0 2 502]]
		10%	50-50	98.04	[[483 11 0 2 4] [2 487 0 1 1] [1 1 492 2 1] [1 1 1 487 4] [2 12 0 2 502]]
		30%	50-50	98	[[483 11 0 2 4] [2 486 0 1 2] [1 1 492 2 1] [1 1 1 487 4] [2 12 0 2 502]]
		50%	50-50	97.92	[[483 11 0 2 4] [2 484 0 1 4] [1 1 492 2 1] [1 1 1 487 4] [2 12 0 2 502]]
		100%	50-50	97.92	[[483 11 0 2 4] [2 484 0 1 4] [1 1 492 2 1] [1 1 1 487 4] [2 12 0 2 502]]
	No Feature	100%	50-50	95.28	[[483 11 0 2 4] [2 484 0 1 4]

	Selection				$\begin{bmatrix} 1 & 1 & 492 & 2 & 1 \\ 1 & 1 & 1 & 487 & 4 \\ 2 & 12 & 0 & 2 & 502 \end{bmatrix}$
Naive Bayes	TF-IDF	5%	70-30	98.066	$\begin{bmatrix} 312 & 4 & 0 & 1 & 1 \\ 2 & 283 & 2 & 0 & 3 \\ 1 & 3 & 304 & 1 & 0 \\ 2 & 0 & 0 & 282 & 1 \\ 1 & 4 & 0 & 3 & 290 \end{bmatrix}$
		10%	70-30	98.133	$\begin{bmatrix} 312 & 4 & 0 & 1 & 1 \\ 2 & 283 & 2 & 0 & 3 \\ 1 & 2 & 305 & 1 & 0 \\ 2 & 0 & 0 & 282 & 1 \\ 1 & 4 & 0 & 3 & 290 \end{bmatrix}$
		30%	70-30	98.066	$\begin{bmatrix} 311 & 4 & 0 & 1 & 2 \\ 2 & 283 & 2 & 0 & 3 \\ 1 & 2 & 305 & 1 & 0 \\ 2 & 0 & 0 & 282 & 1 \\ 1 & 4 & 0 & 3 & 290 \end{bmatrix}$
		50%	70-30	98.066	$\begin{bmatrix} 311 & 4 & 0 & 1 & 2 \\ 2 & 283 & 2 & 0 & 3 \\ 1 & 2 & 305 & 1 & 0 \\ 2 & 0 & 0 & 282 & 1 \\ 1 & 4 & 0 & 3 & 290 \end{bmatrix}$
		100%	70-30	98.0	$\begin{bmatrix} 310 & 4 & 0 & 1 & 3 \\ 2 & 283 & 2 & 0 & 3 \\ 1 & 2 & 305 & 1 & 0 \\ 2 & 0 & 0 & 282 & 1 \\ 1 & 4 & 0 & 3 & 290 \end{bmatrix}$
Naive Bayes	TF-IDF	5%	80-20	98.3	$\begin{bmatrix} 187 & 3 & 0 & 1 & 2 \\ 2 & 219 & 0 & 0 & 0 \\ 2 & 1 & 193 & 1 & 0 \\ 0 & 0 & 0 & 197 & 0 \\ 2 & 2 & 0 & 1 & 187 \end{bmatrix}$
		10%	80-20	98.3	$\begin{bmatrix} 187 & 3 & 0 & 1 & 2 \\ 2 & 219 & 0 & 0 & 0 \\ 2 & 1 & 193 & 1 & 0 \\ 0 & 0 & 0 & 197 & 0 \\ 2 & 2 & 0 & 1 & 187 \end{bmatrix}$
		30%	80-20	98.3	$\begin{bmatrix} 187 & 3 & 0 & 1 & 2 \\ 2 & 219 & 0 & 0 & 0 \\ 2 & 1 & 193 & 1 & 0 \\ 0 & 0 & 0 & 197 & 0 \\ 2 & 2 & 0 & 1 & 187 \end{bmatrix}$
		50%	80-20	98.2	$\begin{bmatrix} 187 & 3 & 0 & 1 & 2 \\ 2 & 219 & 0 & 0 & 0 \end{bmatrix}$

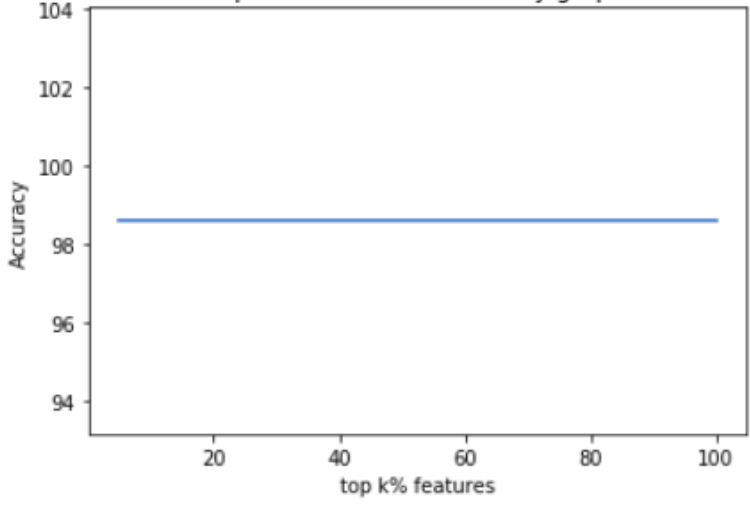
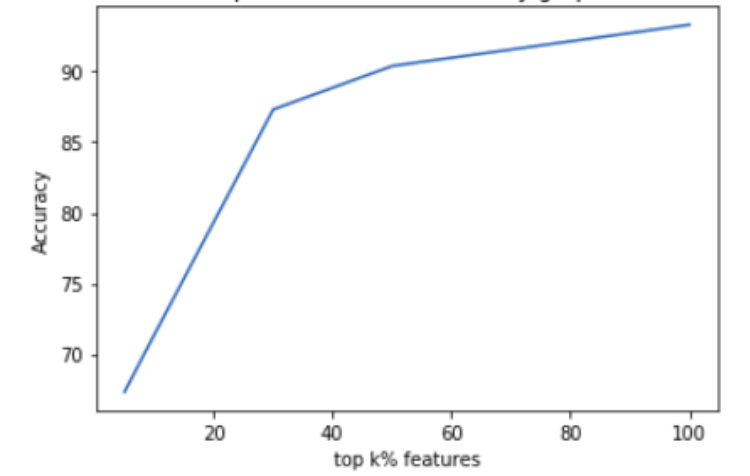
					[2 1 193 1 0] [0 0 0 197 0] [2 2 0 1 187]]
		100%	80-20	98.2	[[187 3 0 1 2] [2 219 0 0 0] [2 1 193 1 0] [0 0 0 197 0] [2 2 0 1 187]]
Naive Bayes	Mutual- Informatio n	5%	50-50	68.6	[[284 143 28 33 16] [12 425 14 3 28] [8 51 418 11 6] [92 47 18 365 9] [74 126 27 39 223]]
		10%	50-50	79.6	[[402 44 15 8 35] [18 425 13 3 23] [4 9 474 5 2] [139 13 9 308 62] [40 44 7 8 390]]
		30%	50-50	88.92	[[415 26 9 25 29] [21 417 5 10 29] [4 8 478 2 2] [11 12 9 490 9] [20 26 10 10 423]]
		50%	50-50	91.12	[[422 32 7 17 26] [15 427 12 3 25] [5 3 481 5 0] [2 5 5 505 14] [13 18 4 11 443]]
		100%	50-50	93.479	[[464 17 11 4 8] [13 446 16 1 6] [1 3 489 0 1] [1 6 16 504 4] [7 36 7 5 434]]
Naive Bayes	Mutual- Informatio n	5%	70-30	80.4	[[228 45 7 13 27] [10 254 6 4 5] [6 18 258 14 13] [28 11 6 232 21] [27 30 0 3 234]]
		10%	70-30	86.66	[[266 20 3 14 17] [15 243 3 11 7] [4 7 284 10 4] [16 4 1 266 11] [18 22 3 10 241]]
		30%	70-30	89.33	[[263 14 3 21 19] [9 245 3 8 14] [0 4 291 8 6] [5 1 1 280 11]]

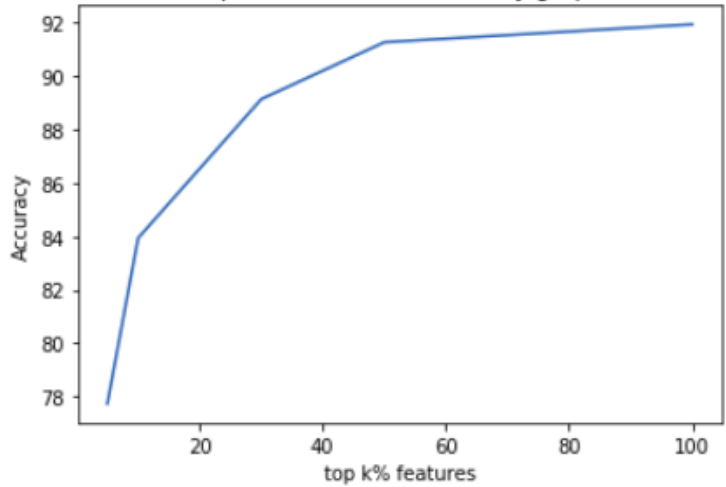
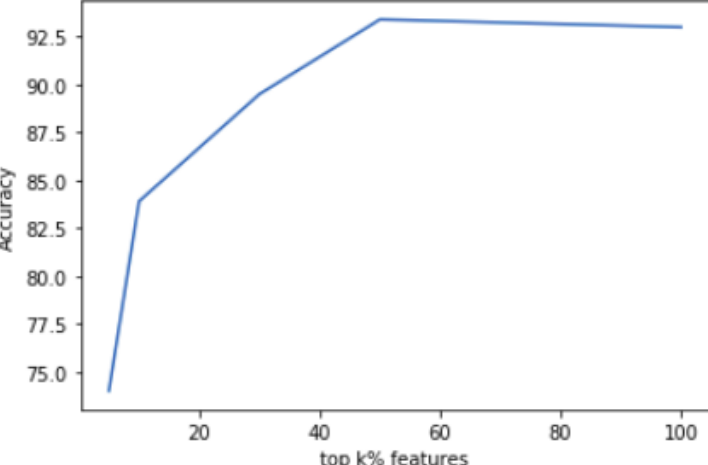
					[5 15 3 10 261]]
		50%	70-30	89.8	[[269 16 11 11 13] [11 237 10 3 18] [1 4 304 0 0] [4 3 6 277 8] [5 23 3 3 260]]
		100%	70-30	90.933	[[266 10 21 2 21] [2 244 15 1 17] [0 2 307 0 0] [0 3 24 262 9] [1 4 4 0 285]]
Naive Bayes	Mutual-Information	5%	80-20	74.0	[[112 21 4 28 40] [5 135 3 9 34] [1 3 169 22 13] [11 3 1 187 10] [2 31 1 18 137]]
		10%	80-20	83.899	[[156 11 3 18 17] [15 146 0 4 21] [7 1 192 6 2] [13 5 4 185 5] [7 12 0 10 160]]
		30%	80-20	89.5	[[178 4 3 12 8] [18 151 2 1 14] [0 0 206 1 1] [6 1 3 201 1] [5 7 4 14 159]]
		50%	80-20	93.4	[[181 5 4 6 9] [11 165 1 1 8] [0 0 205 2 1] [2 1 4 203 2] [2 2 1 4 180]]
		100%	80-20	93.0	[[182 6 10 1 6] [3 173 7 0 3] [0 1 206 1 0] [0 0 19 190 3] [1 4 5 0 179]]

GRAPHS OF ACCURACY FOR NAIVE BAYES

Train-Test Split	Feature Selection Technique	Accuracy vs top k% features
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50-50	TF-IDF	<p>top k% features vs Accuracy graph</p> <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>0</td><td>97.760</td></tr><tr><td>10</td><td>97.760</td></tr><tr><td>20</td><td>97.760</td></tr><tr><td>30</td><td>97.760</td></tr><tr><td>40</td><td>97.740</td></tr><tr><td>50</td><td>97.720</td></tr><tr><td>60</td><td>97.720</td></tr><tr><td>70</td><td>97.720</td></tr><tr><td>80</td><td>97.720</td></tr><tr><td>90</td><td>97.720</td></tr><tr><td>100</td><td>97.720</td></tr></tbody></table>	top k% features	Accuracy	0	97.760	10	97.760	20	97.760	30	97.760	40	97.740	50	97.720	60	97.720	70	97.720	80	97.720	90	97.720	100	97.720
top k% features	Accuracy																									
0	97.760																									
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100	97.720																									
70-50	TF-IDF	<p>top k% features vs Accuracy graph</p> <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>0</td><td>98.525</td></tr><tr><td>10</td><td>98.525</td></tr><tr><td>20</td><td>98.475</td></tr><tr><td>30</td><td>98.400</td></tr><tr><td>40</td><td>98.400</td></tr><tr><td>50</td><td>98.400</td></tr><tr><td>60</td><td>98.385</td></tr><tr><td>70</td><td>98.370</td></tr><tr><td>80</td><td>98.355</td></tr><tr><td>90</td><td>98.340</td></tr><tr><td>100</td><td>98.335</td></tr></tbody></table>	top k% features	Accuracy	0	98.525	10	98.525	20	98.475	30	98.400	40	98.400	50	98.400	60	98.385	70	98.370	80	98.355	90	98.340	100	98.335
top k% features	Accuracy																									
0	98.525																									
10	98.525																									
20	98.475																									
30	98.400																									
40	98.400																									
50	98.400																									
60	98.385																									
70	98.370																									
80	98.355																									
90	98.340																									
100	98.335																									

80-50	TF-IDF	<p>top k% features vs Accuracy graph</p>  <p>The graph shows a horizontal line at an accuracy of approximately 98.5% across the range of top k% features from 10 to 100. The y-axis is labeled 'Accuracy' and ranges from 94 to 104. The x-axis is labeled 'top k% features' and ranges from 0 to 100.</p> <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>10</td><td>98.5</td></tr><tr><td>20</td><td>98.5</td></tr><tr><td>30</td><td>98.5</td></tr><tr><td>40</td><td>98.5</td></tr><tr><td>50</td><td>98.5</td></tr><tr><td>60</td><td>98.5</td></tr><tr><td>70</td><td>98.5</td></tr><tr><td>80</td><td>98.5</td></tr><tr><td>90</td><td>98.5</td></tr><tr><td>100</td><td>98.5</td></tr></tbody></table>	top k% features	Accuracy	10	98.5	20	98.5	30	98.5	40	98.5	50	98.5	60	98.5	70	98.5	80	98.5	90	98.5	100	98.5
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10	98.5																							
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60	98.5																							
70	98.5																							
80	98.5																							
90	98.5																							
100	98.5																							
50-50	MI	<p>top k% features vs Accuracy graph</p>  <p>The graph shows a line that starts at an accuracy of approximately 68% for k=10, rises sharply to about 87.5% at k=30, and then continues to rise more gradually to approximately 95% at k=100. The y-axis is labeled 'Accuracy' and ranges from 70 to 90. The x-axis is labeled 'top k% features' and ranges from 0 to 100.</p> <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>10</td><td>68</td></tr><tr><td>20</td><td>78</td></tr><tr><td>30</td><td>87.5</td></tr><tr><td>40</td><td>89</td></tr><tr><td>50</td><td>90.5</td></tr><tr><td>60</td><td>91.5</td></tr><tr><td>70</td><td>92.5</td></tr><tr><td>80</td><td>93.5</td></tr><tr><td>90</td><td>94.5</td></tr><tr><td>100</td><td>95</td></tr></tbody></table>	top k% features	Accuracy	10	68	20	78	30	87.5	40	89	50	90.5	60	91.5	70	92.5	80	93.5	90	94.5	100	95
top k% features	Accuracy																							
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100	95																							

70-50	MI	<p>top k% features vs Accuracy graph</p>  <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>5</td><td>77.8</td></tr><tr><td>10</td><td>84.0</td></tr><tr><td>20</td><td>86.5</td></tr><tr><td>30</td><td>89.2</td></tr><tr><td>40</td><td>90.0</td></tr><tr><td>50</td><td>91.5</td></tr><tr><td>60</td><td>91.6</td></tr><tr><td>70</td><td>91.7</td></tr><tr><td>80</td><td>91.8</td></tr><tr><td>90</td><td>91.8</td></tr><tr><td>100</td><td>91.8</td></tr></tbody></table>	top k% features	Accuracy	5	77.8	10	84.0	20	86.5	30	89.2	40	90.0	50	91.5	60	91.6	70	91.7	80	91.8	90	91.8	100	91.8
top k% features	Accuracy																									
5	77.8																									
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80-50	MI	<p>top k% features vs Accuracy graph</p>  <table><thead><tr><th>top k% features</th><th>Accuracy</th></tr></thead><tbody><tr><td>5</td><td>74.2</td></tr><tr><td>10</td><td>84.2</td></tr><tr><td>20</td><td>86.8</td></tr><tr><td>30</td><td>89.5</td></tr><tr><td>40</td><td>91.0</td></tr><tr><td>50</td><td>93.2</td></tr><tr><td>60</td><td>93.1</td></tr><tr><td>70</td><td>93.0</td></tr><tr><td>80</td><td>92.9</td></tr><tr><td>90</td><td>92.9</td></tr><tr><td>100</td><td>92.9</td></tr></tbody></table>	top k% features	Accuracy	5	74.2	10	84.2	20	86.8	30	89.5	40	91.0	50	93.2	60	93.1	70	93.0	80	92.9	90	92.9	100	92.9
top k% features	Accuracy																									
5	74.2																									
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70	93.0																									
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90	92.9																									
100	92.9																									

KNN Classifiers

Classifiers Name	Feature Selection Technique	% of features selected	Train-Test Split	Accuracy	Confusion Matrix
K=1	TF-IDF	1	50-50	88.24	[[429 41 8 17 16] [23 454 5 1 6] [10 9 482 4 2] [12 6 2 474 4] [81 18 16 13 367]]
	TF-IDF	1	70-30	89.266	[[251 19 10 9 7] [17 282 1 2 6] [2 3 297 0 1] [10 6 7 284 1] [37 16 5 2 225]]
	TF-IDF	1	80-20	86.4	[[177 13 8 4 7] [10 185 3 1 2] [1 5 198 1 0] [9 9 1 157 3] [32 17 7 3 147]]
	TF-IDF	5	70-30	84.533	[[260 26 13 10 1] [20 270 8 2 4] [7 21 283 0 3] [12 11 17 253 3] [29 31 6 8 202]]
	MI	1	50-50	89.2	[[427 33 8 5 41] [22 452 3 3 14] [11 4 484 1 4] [10 15 4 466 3] [43 17 6 5 419]]
	MI	1	70-30	88.533	[[271 17 4 2 15] [18 274 3 3 7] [6 5 283 1 2] [14 11 2 271 8] [34 12 7 1 229]]
	MI	1	80-20	90.8	[[175 7 0 3 8] [10 188 4 3 2] [4 2 210 0 0] [3 4 2 186 2] [27 8 3 0 149]]
	MI	5	70-30	85.133	[[265 26 9 2 7] [16 276 7 1 5]

					[11 6 278 1 1] [20 18 9 253 6] [44 26 6 2 205]]
K=3	TF-IDF	1	50-50	82.0	[[266 93 29 37 95] [0 424 30 15 36] [1 2 470 13 16] [0 4 4 438 28] [10 15 11 11 452]]
	TF-IDF	1	70-30	85.33	[169 54 23 20 30] [1 262 18 7 20] [0 0 293 6 4] [1 0 2 296 9] [5 10 8 2 260]]
	TF-IDF	1	80-20	84.2	[[117 38 20 10 24] [0 173 20 2 6] [0 0 202 3 0] [0 0 1 172 6] [4 12 8 4 178]]
	TF-IDF	5	70-30	79.533	[[147 77 42 28 16] [0 243 31 22 8] [0 0 295 14 5] [0 1 7 274 14] [1 16 15 10 234]]
	MI	1	50-50	84.92	[[281 73 20 17 123] [0 425 13 13 43] [0 0 484 3 17] [1 1 1 462 33] [7 6 4 2 471]]
	MI	1	70-30	83.866	[[172 41 16 11 69] [0 250 15 12 28] [0 0 280 3 14] [0 1 1 286 18] [3 6 3 1 270]]
	MI	1	80-20	87.12	[[118 25 4 8 38] [0 179 10 9 9] [0 0 211 2 3] [0 0 0 187 10] [4 3 1 2 177]]
	MI	5	70-30	82.933	[[182 57 26 7 37] [0 255 19 6 25] [0 0 283 3 11] [0 3 7 267 29] [4 11 7 4 257]]
K=5	TF-IDF	1	50-50	74.44	[[141 105 56 67 142] [1 340 57 35 56] [0 0 456 27 24] [0 0 0 449 49]

					[[1 5 6 8 475]]
	TF-IDF	1	70-30	76.2	[[89 63 52 30 62] [0 222 35 12 39] [0 0 286 9 8] [0 1 0 273 34] [0 3 5 4 273]]
	TF-IDF	1	80-20	76.6	[[69 46 38 12 44] [0 145 31 10 15] [0 0 194 5 6] [0 0 0 163 16] [0 1 8 2 195]]
	TF-IDF	5	70-30	71.733	[94 81 52 49 34] [0 196 47 47 14] [0 0 275 33 6] [0 0 3 264 29] [0 6 14 9 247]]
	MI	1	50-50	77.16	[[195 67 22 24 206] [0 368 28 17 81] [0 0 462 6 36] [1 0 0 419 78] [0 0 4 1 485]]
	MI	1	70-30	74.333	[[108 40 27 16 118] [0 203 31 19 52] [0 0 272 3 22] [0 0 0 255 51] [0 2 3 1 277]]
	MI	1	80-20	80.5	[[72 35 8 14 64] [0 160 14 13 20] [0 0 207 2 7] [0 0 0 181 16] [0 0 0 2 185]]
	MI	5	70-30	73.666	[[101 78 45 15 70] [0 224 24 12 45] [0 0 262 10 25] [0 1 2 250 53] [0 8 6 1 268]]

INFERENCES FROM RESULTS:-

NAIVE BAYES Classifiers

1. In Naive Bayes Classifiers top 5% features according to TF-IDF are giving very good results in all train test splits ie approx 98%. If we increase our no of features further there is no improvements in result ie they are same . It means that there is only 5% top features that are important based on TF-IDF
2. In Naive Bayes Classifiers when we are taking features according to Mutual-Information . We can see from graphs for all train-test splits there is improvement in accuracy results when we increase no of features . After 40-50% top features results are nearly constant or overfitted .
3. We are getting 98% accuracy in TF-IDF and 93% accuracy in Mutual Information because in MI we have considered top k% for each class where as in TF-IDF top k % same for each class . And TF-IDF are considered according to classwise and MI according to document wise
4. In TF_IDF we can clearly see overfitting as accuracy is nearly same or less

KNN Classifiers

1. With increase in value of K in KNN accuracy start decreasing . With increase in K It is including noise also . So testing accuracy start decreasing
2. IN TF-IDF and MI top 1% features are giving good results . If we increase it to top 5% accuracy start decreasing slightly . With more features accuracy start decreasing as it is overfitting and including noise also .
3. With 80-20 split we have achieved highest accuracy ie 90% . As increase in data KNN start increasing its accuracy . With more data KNN performs better

ADDITIONAL INFERENCES:-

1. Naive Bayes performs better than KNN because Naive works better on small dataset and its assumption of conditional independence
2. Naive Bayes work on principle of conditional independence so it performs better
3. Naive Bayes is very fast when compared to KNN as it doesn't calculate cosine similarity and distance . Execution time is very fast when compared to KNN