

Text Categorization (by Topic)

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Overview

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Text Categorization

- Classify document context according to a set of pre-defined topics.
- One document might contain more than one topic.

Dataset

- Dataset Reuters-21578 news articles containing 118 unevenly distributed topics.
- Modified Apte Split: 9063 training documents and 3299 test documents.
- Using top 10 categories.

class	train	text	class	train	text
<i>earn</i>	2877	1087	<i>trade</i>	369	119
<i>acq</i>	1650	719	<i>interest</i>	347	131
<i>money-fx</i>	538	179	<i>ship</i>	197	89
<i>grain</i>	433	149	<i>wheat</i>	212	71
<i>crude</i>	389	189	<i>corn</i>	182	56

Features

- Clean document text, remove stop-words, stemming.
- tf-idf document representation of monograms and bigrams.

- Multinomial Naive Bayes, K-Nearest Neighbours and Linear Support Vector Machine
- Use grid search to look for best parameters over a set of pre-defined values.

Results

F1 top 10 categories scores per class and micro-averaged.

	NB ^[1]	NB ^[3]	kNN ^[2]	kNN ^[3]	SVM ^[1]	SVM ^[3]
<i>earn</i>	96	98	97	93	98	99
<i>acq</i>	88	97	92	82	94	97
<i>money-fx</i>	57	84	78	84	75	85
<i>grain</i>	79	90	82	91	95	95
<i>crude</i>	80	87	86	87	89	91
<i>trade</i>	64	80	77	85	76	87
<i>interest</i>	65	76	74	79	78	80
<i>ship</i>	85	58	79	75	86	78
<i>wheat</i>	70	67	77	74	92	81
<i>corn</i>	65	55	78	71	90	82
micro-avg	82	91	82	87	92	94

Dumais et al. (1998) ^[1], Joachims (1998) ^[2], Presented work ^[3]

Thank You!