

# Text Mining and Search Project

## Amazon Reviews Dataset Text Classification

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# DATASET

## AMAZON REVIEWS DATASET



**Topics:** Electronics,  
Kindle, CDs and  
Movies/TV  
**Format:** JSON  
**Size:** > 5 000 000  
reviews

```
{  
  "asin": "B00000JBLH",  
  "helpful": [3, 4],  
  "overall": 5.0,
```

Classe

```
  "reviewText": "I bought my first HP12C in about 1984 or so,  
                  and it served me faithfully until 2002 when I lost it while travelling.  
                  I searched for another one to replace it, but found one difficult  
                  to come by in my area.  
  
                  ...  
                  I didn't even have to replace the batteries in well over a decade of use!  
                  HP 12C, I'm coming home!",  
  "reviewTime": "09 3, 2004",  
  "reviewerID": "A32T2H8150OJLU",  
  "reviewerName": "ARH",  
  "summary": "A solid performer, and long time friend",  
  "unixReviewTime": 1094169600  
}
```

Testo

# SUBSET CONSIDERATO

1

***Recensioni valutate più di 5 volte***

Righe: da 5466932 a 874364

"helpful": [3, 4]

> 5

2

***Recensioni con almeno 5 parole***

Righe: da 874354 a 873718

```
cond = df['word_count'] > 4  
df = df[cond]
```

3

***Recensioni solo inglesi***

Righe: da 873718 a 872307

Rilevate con "langdetect"

# PREPROCESSING

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# PREPROCESSING

STEP  
**1**

**TOKENIZATION**

STEP  
**2**

**NORMALIZATION**

STEP  
**3**

**REMOVE NUMBERS AND PUNCTUATION**

STEP  
**4**

**STOP WORDS REMOVAL**

STEP  
**5**

**LEMMATIZATION**

STEP  
**6**

**STEMMING**

# PREPROCESSING

## TOKENIZATION

So, I decided to buy up and purchased an HP 49G.  
What a mistake! I know that many people view the  
HP 49G (now 49G+)



[ So,, i, decided, to, buy, up, and, purchased, an, hp,  
49g., what, a, mistake!, i, know, that, many, people,  
view, the, hp, 49g, (, now, 49g+, ), ]

# PREPROCESSING

## NORMALIZATION

So, I decided to buy up and purchased an HP 49G.  
What a mistake! I know that many people view the  
HP 49G (now 49G+)



so, i decided to buy up and purchased an hp 49g.  
what a mistake! i know that many people view the  
hp 49g (now 49g+)



# PREPROCESSING

## REMOVE NUMBERS AND PUNCTUATION

so / i decided to buy up and purchased an hp 49g /  
what a mistake! i know that many people view the  
hp 49g (now 49g+)



so i decided to buy up and purchased an hp g what  
a mistake i know that many people view the hp g  
now g

# PREPROCESSING

## STOP WORDS REMOVAL

so / decided to / buy up / and purchased an hp g what  
a / mistake i / know th / at many people view the hp g  
now g



decided buy purchased hp g mistake know  
many people view hp g g

# PREPROCESSING

## LEMMATIZATION/STEMMING

decided buy purchased hp g mistake know  
many people view hp g g



decid buy purchas hp g mistak know mani  
peopl view hp g g

# PREPROCESSING

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# TEXT REPRESENTATION

## COUNT VECTORIZER

- Considerati Uni-gram e Bi-gram → «decided» «decided buy» «purchased»
- Feature selection → *select Kbest*: 10000 feature con *chi2*
- 5 Classification models

Terms

Documents →

	are	call	from	hello	home	how	me	money	now	tomorrow	win	you
0	1	0	0	1	0	1	0	0	0	0	0	1
1	0	0	1	0	1	0	0	1	0	0	2	0
2	0	1	0	0	0	0	1	0	1	0	0	0
3	0	1	0	1	0	0	0	0	0	1	0	1

# TEXT REPRESENTATION

## TF-IDF

- Considerati Uni-gram e Bi-gram → «decided» «decided buy» «purchased»
- Feature selection → *select Kbest: 10000 feature con chi2*
- 5 Classification models

Documents →

	Terms	Tf-Idf value
	↓	↓
(0,	10255)	0.0727336331113147
(0,	44451)	0.35101474415480843
(0,	88374)	0.13282060609760257
(0,	32885)	0.16335768847169346
(0,	54975)	0.09907168946999645
(0,	104657)	0.16505658094946435
	...	
(0,	87359)	0.13961763759124002
(0,	82798)	0.08586926237109241
(0,	22729)	0.09028466248740455
(0,	12007)	0.06171032370388875
(0,	79021)	0.08285922464177037
(0,	44437)	0.2327801290815596

# CLASSIFICATION MODELS

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# CLASSIFICATION MODELS

Model		CountVec	Tf-Idf
LinearSVC	Time:	24min 57s	8min 33s
	Accuracy:	0.6056 %	0.6158 %
Log. Regression	Time:	30min 13s	1min 36s
	Accuracy:	0.6090 %	0.6094 %
Random Forest	Time:	39.5 s	48.7 s
	Accuracy:	0.4489 % (*)	0.4489 % (*)
Multinomial NB	Time:	745 ms	622 ms
	Accuracy:	0.4996 %	0.5131 %
Gradient Boost	Time:	2min	2min 33s
	Accuracy:	0.4957 %	0.4992 %

(\*) tutte le recensioni classificate come 5 stelle: F-Score nulla



# EVALUATION AND CONCLUSION

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# EVALUATION AND CONCLUSION

1

## RISULTATI OTTENUTI

Efficiency



TF-IDF

Effectiveness



LinearSVC  
Logistic Regression

**Accuracy: 60 %**

2

## POSSIBILI MIGLIORAMENTI

Altri metodi di Feature Extraction - Selection/Synthesis – Weighting  
(es. Word2vec, mutual dependence, Matrix decomposition...)