

# Subreddit Classification:

r/Physics and r/chemistry

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# Primary Goal

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1. Predict which subreddit a post belongs to based on its content

# Overview

1. Get subreddit data
2. Data cleaning and preprocessing
3. Baseline Accuracy?
4. Variable selection
  - a. Which features to use?
  - b. How to simplify data (lemmatization, stemming)
5. Explore:
  - a. Vectorizers
  - b. Classifiers
6. Compare results

# r/Physics

# r/Chemistry

- **r/Physics** and **r/chemistry**: two subreddits where redditors can ask questions pertaining to, or discuss topics of, physics or chemistry topics.
- Both of these fields have specific jargon, potentially allowing for easier discrete classification
- This allows us to build an appropriate model

The image shows a screenshot of the Reddit website, specifically the subreddits r/Chemistry and r/Physics. The top section displays the r/Chemistry subreddit, which has a green header. It features a pinned post titled "Chemical Literature Day—What are you reading?" and another post titled "The compounds behind the beauty" with a thumbnail image of gemstones. The right sidebar for r/Chemistry includes an "About Community" section stating it is for chemists and chemistry lovers, with 1.2m members and 648 online. It also has a "Filter by flair" section with options like Educational, Image, Video, Question, and Comic. Below this is an advertisement for "Chegg Study Pack".

The bottom section displays the r/Physics subreddit, which has a blue header. It features a pinned post titled "Physics Questions Thread - Week 40, 2020" and another post titled "Careers/Education Questions Thread - Week 40, 2020". The right sidebar for r/Physics includes an "About Community" section stating its aim is to build a subreddit for physicists, scientists, and physics enthusiasts, with 1.5m members and 546 online. It also has a "Filter by flair" section with options like Feature, Video, Question, Image, News, and Academic.

# Factors to prioritize

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1. Post text content
2. Post title

# Data Cleaning and Preprocessing

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Methodology

# Build a Set of Reddit Comments

- **Reddit API?**

- Strict limitations on amount of data we can access (can only pull 100 of the most recent comments)

- **Pushshift API**

- Open-source alternative to Reddit's API
- Can specify date and time we start our pull from
- Data returned as list of dictionaries

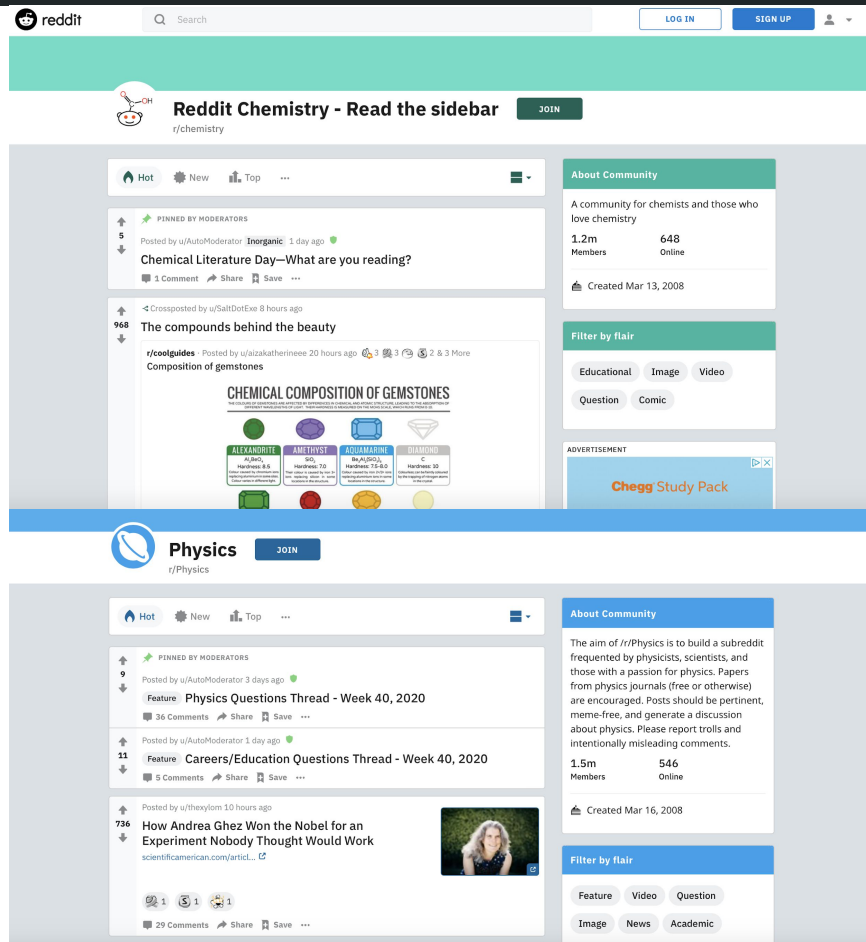
- **My data**

- All comments for each subreddit in the past two years

```
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      "all_awardings": [],
      "allow_live_comments": false,
      "author": "Acanthocephala0k5166",
      "author_flair_css_class": null,
      "author_flair_richtext": [],
      "author_flair_text": null,
      "author_flair_type": "text",
      "author_fullname": "t2_81pyq1n0",
      "author_patreon_flair": false,
      "author_premium": false,
      "awards": [],
      "can_mod_post": false,
      "contest_mode": false,
      "created_utc": 1602250070,
      "domain": "youtube.com",
      "full_link": "https://www.reddit.com/r/audiophile/cc",
      "gildings": {},
      "id": "j7ywrs",
      "is_crosspostable": true,
      "is_meta": false,
      "is_original_content": false,
      "is_reddit_media_domain": false,
      "is_robot_indexable": true,
      "is_self": false,
      "is_video": false,
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      "link_flair_css_class": "red",
      "link_flair_richtext": [],
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      "link_flair_text_color": "dark",
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          "author_url": "https://www.youtube.com/chanr",
          "height": 338,
          "html": "<iframe width=\"600\" height=\"338\" src=\"https://www.youtube.com/embed/feature=oembed&enablejsapi=1&frameborder=0&allow=autoplay&allowfullscreen&allowscriptaccess=always\"></iframe>",
          "provider_name": "YouTube",
          "provider_url": "https://www.youtube.com/",
          "thumbnail_height": 360,
          "thumbnail_url": "https://i.ytimg.com/vi/3f:",
          "thumbnail_width": 480
        }
      }
    }
  ]
}
```

# Data Cleaning

- **Only keep rows:**
  - Where 'selftext' and 'title' are present, and have at least 4 words
- **Remove:**
  - Special characters
  - Any word that is an overly common word, or 'stop word' in the english language
    - Reduces noise
  - Non-letter characters
- **Expand contractions**
  - Didn't end up working
- **Tokenizing**
  - Easier for processing
- **Lemmatizing**
  - Keep roots of words
- **Stemming**
  - Crude version of lemmatizing





# Data Preprocessing

- **Created CSV files:**
  - Lemmatized data (removed stop words)
  - Stemmed data (removed stop words)
- **Kept columns:**
  - Subreddit
  - Selftext
  - Title

## LEMMATIZATION

subreddit	selftext	title
1	D simulation using CUDA	Choice GPU running MD simulation NAMD GROMACS
1	logize difficult understand	Balance Bird v Spin top
1	etic field Thoughts please	Do electron convective core
1	ne would unravel problem	Is quantum computing dangerous impossible

## STEMMING

subreddit	selftext	title
1	1 t come MD simul use cuda	choic gpu run MD simul namd gromac
2	1 apolog difficult understand	balanc bird v spin top
3	1 magnet field thought pleas	Do electron convect core
4	1 on would unravel problem	Is quantum comput danger imppos

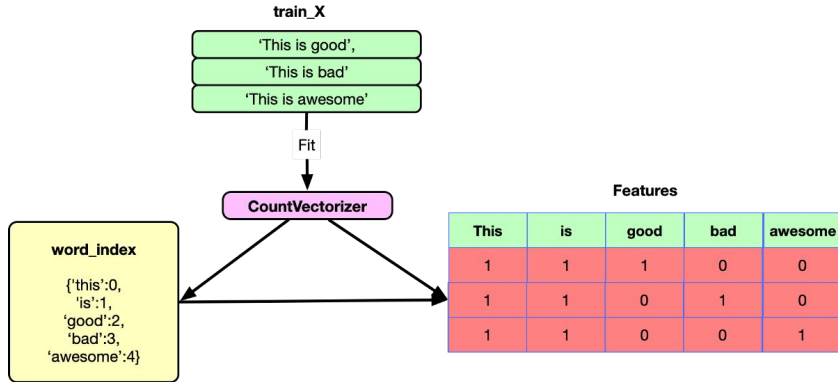
# Natural Language Processing

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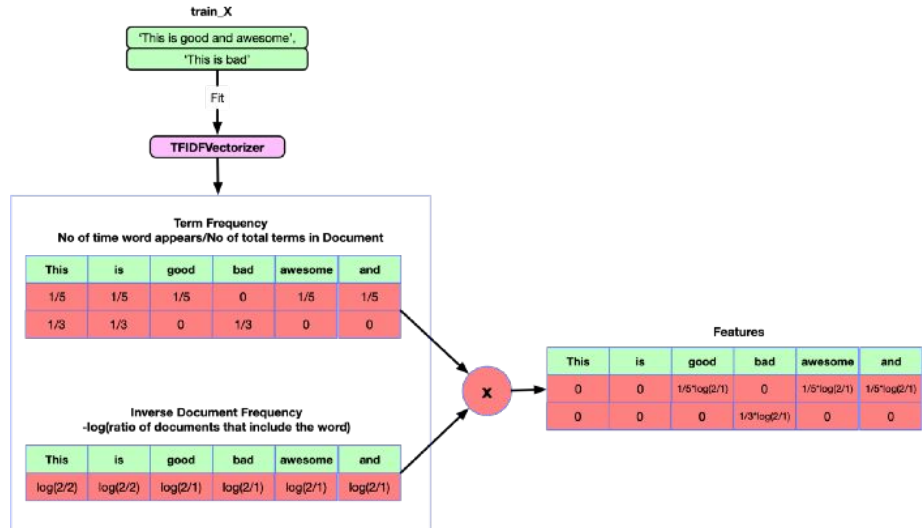
1. Naive Bayes
2. Linear Regression

# Vectorizers

## COUNTVECTORIZER



## TFIDFVECTORIZER



# Vectorizers

- **Bag of Words:**

- Use a vectorizer to split comment into words
- Convert each comment into a vector of word frequencies

- **CountVectorizer**

- Creates pure frequency vector

- **TfidfVectorizer**

- Normalizes frequencies
- Up-weighting rare words
  - Good for jargon identification

$$w_{x,y} = \text{tf}_{x,y} \times \log\left(\frac{N}{\text{df}_x}\right)$$

**TF-IDF**

Term  $x$  within document  $y$

$\text{tf}_{x,y}$  = frequency of  $x$  in  $y$

$\text{df}_x$  = number of documents containing  $x$

$N$  = total number of documents

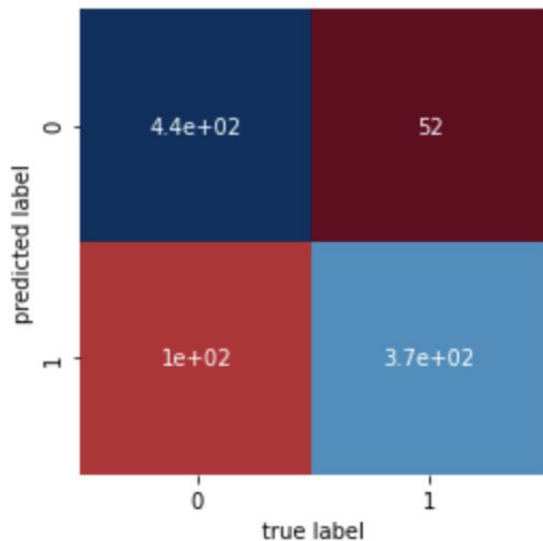
# Modeling with Naive Bayes

## LEMMATIZED

**Accuracy score: 0.84**

**Precision score: 0.88**

**Recall score: 0.78**

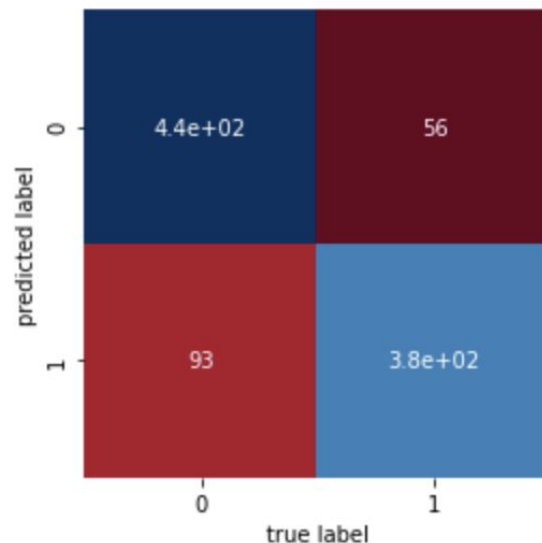


## STEMMED

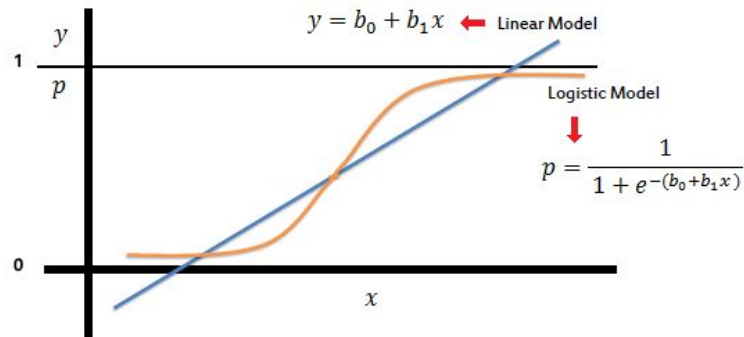
**Accuracy score: 0.85**

**Precision score: 0.87**

**Recall score: 0.80**



# Logistic Regression: CountVectorizer



- **Logistic regression: classic technique**
  - Here with Stemmed data
  - Highly interpretable
- Can look at words or n-grams the model associates most to a subreddit

	precision	recall	f1-score	support
Chemistry	0.85	0.88	0.86	493
Physics	0.87	0.83	0.85	469
accuracy			0.86	962
macro avg	0.86	0.86	0.86	962
weighted avg	0.86	0.86	0.86	962

	Predicted Physics	Predicted chemistry
Actual Physics	435	58
Actual chemistry	79	390

Outperformed Naive Bayes (recall score .80)

# Logistic Regression: TfidfVectorizer

- **TfidfVectorizer:**

- Evaluates how relevant a word is to a collection of documents

	precision	recall	f1-score	support
Chemistry	0.90	0.92	0.91	493
Physics _	0.92	0.89	0.90	469
accuracy			0.91	962
macro avg	0.91	0.90	0.91	962
weighted avg	0.91	0.91	0.91	962

## Best performance so far

- Uses Tfidf Vectorization with normalization
- Ridge regulation with a strength of  $\alpha = 1$
- Excludes stop words
- Includes all words (n-grams that appear in at least one comment)

	Predicted Physics	Predicted chemistry
Actual Physics	455	38
Actual chemistry	53	416

# Comparisons

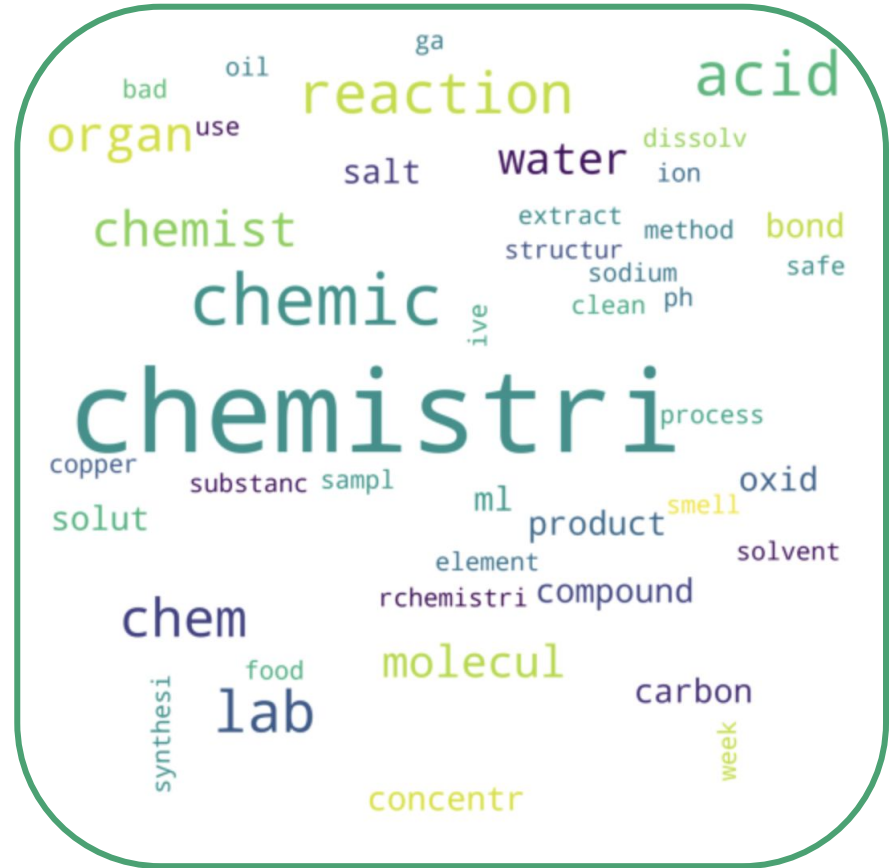
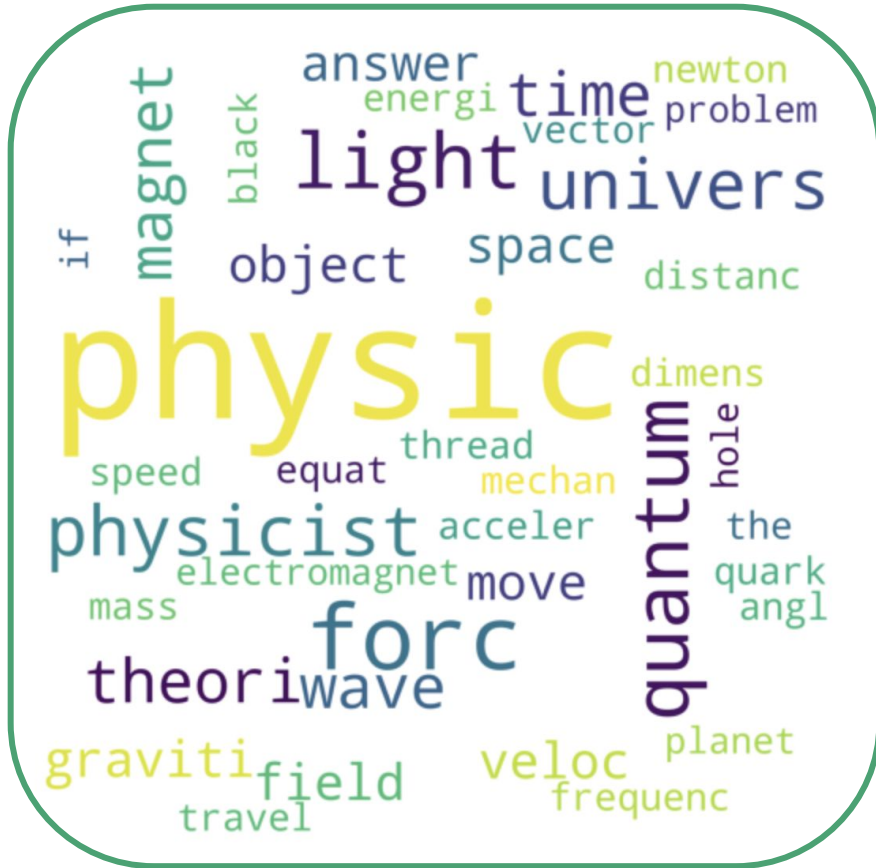
- Naive Bayes:
  - Lemmatizing: .78
  - Stemming: .80
- Logistic Regression:
  - CountVectorizer:
    - Chemistry: .88
    - Physics: .83
  - Tfidf:
    - Chemistry: .92
    - Physics: .89

ngram	coef
physic	7.311782
forc	2.812227
quantum	2.324193
light	2.304542
physicist	2.070253
univers	2.014284
theori	1.970260
wave	1.836353
time	1.811903
magnet	1.735726

ngram	coef
chemistri	-6.834085
chemic	-3.450784
acid	-2.842138
lab	-2.800149
reaction	-2.649828
chem	-2.622460
chemist	-2.275908
organ	-2.254280
water	-2.150410
molecul	-2.099812



# Common Word Roots



# Future Directions

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1. Try with just using 'title' as a feature
2. More classification models:
  - a. Random Forest?
3. Get more data

# Sources

- <https://www.mygreatlearning.com/blog/multinomial-naive-bayes-explained/>
- <https://towardsdatascience.com/naive-bayes-document-classification-in-python-e33ff50f937e>
- <https://monkeylearn.com/blog/what-is-tf-idf/>
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