

Classification of U.S. Supreme Court Opinions

*Using various NLP
techniques.*



INTRODUCTION

- Thousands of court opinions are published each year.
- They have to be manually analyzed and categorized to facilitate research.
- If we can automate this process it can dramatically lower costs.





PROCESS

- Full text of some 8000 Supreme Court was gathered
- Labels were added to the opinions which classified into categories.
 - 13 Categories
- We then tried to see if various NLP Algorithms could accurately classify opinions into the right category
- We also tried to use unsupervised learning to recreate these topics.

LABEL DISTRIBUTION

3500
3000
2500
2000
1500
1000
500
0

Criminal
Procedure

Economic
Activity

Civil
Rights

Judicial
Power

First
Amendment

Due
Process

Federalism

Unions

Federal
Taxation

Privacy

Attorneys

Interstate
Relations

Miscellaneous



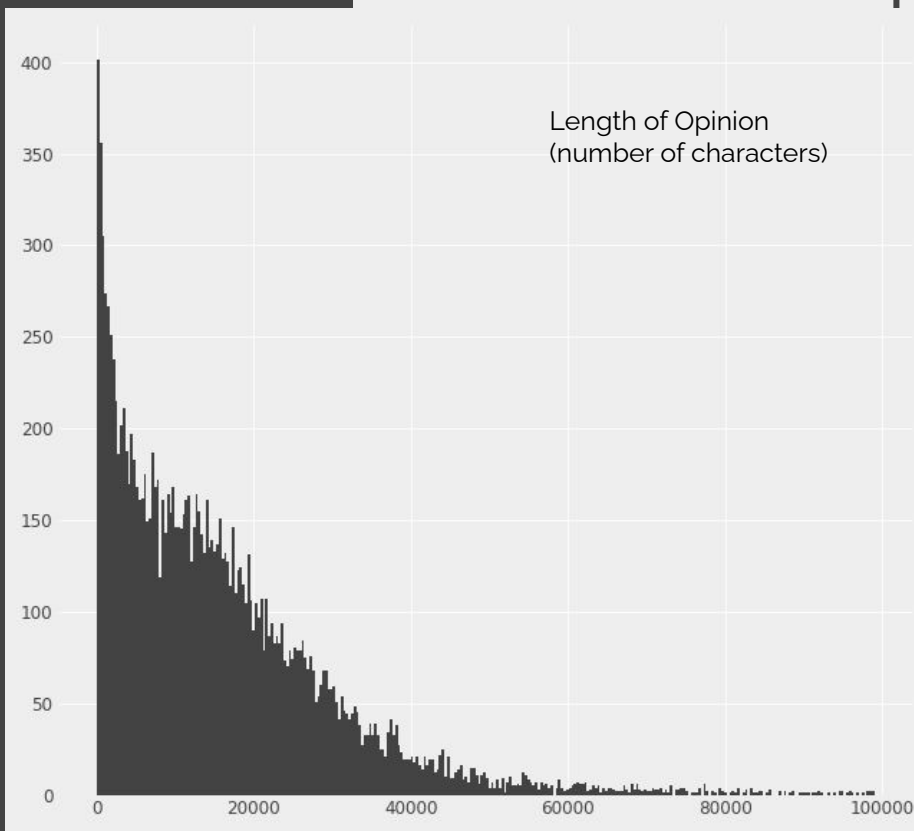
In order to minimize noise and improve model performance, we narrowed down the opinions to build our classifier.

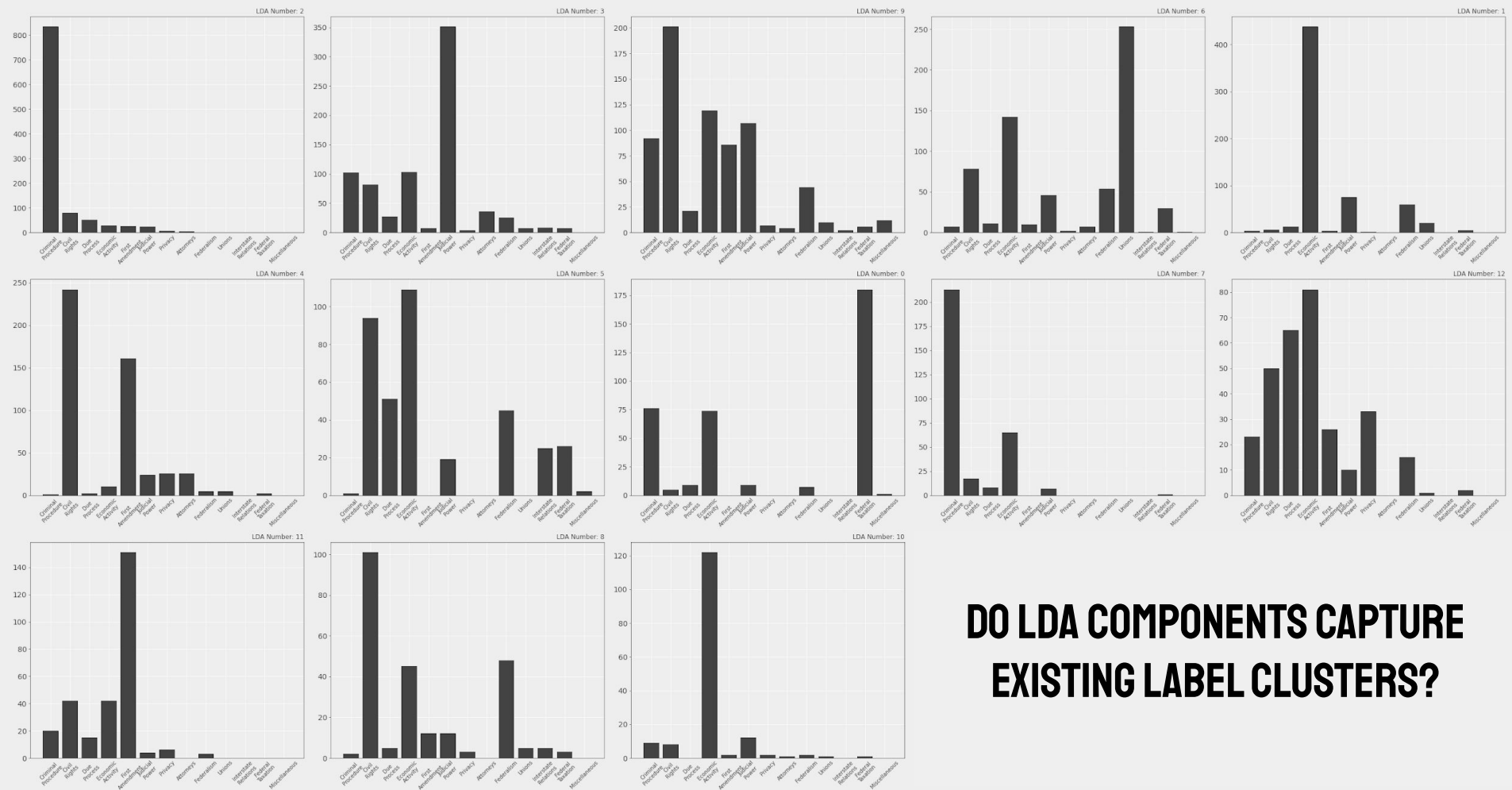
We only kept opinions whose lengths were;

- greater than 5000 characters;
 - A lot of these are per curiam dismissals and affirmations of lower court decisions with no substance.
- less than 85000 characters;
 - This leaves out unusually long opinions

We also filtered out dissents, since they discuss the same subject matter, so seem redundant for the purposes of classification and may add noise.

This leaves us with 6,329 opinions to use for training and testing our model.





**DO LDA COMPONENTS CAPTURE
EXISTING LABEL CLUSTERS?**

175000 -

150000 -

125000 -

100000 -

75000 -

50000 -

25000 -

0 -

court

state

states

act

case

united

federal

law

district

congress

id

petitioner

appeals

statute

2d

government

rules

rule

trial

did

judgment

supra

evidence

new

action

courts

cases

amendment

question

right

CLASSIFICATION PERFORMANCE

0	Base Random Forest	0.616560	0.665296	RandomForestClassifier	CountVectorizer
1	base KNN	0.612368	0.663621	RandomForestClassifier	CountVectorizer
2	Base MNB	0.757261	0.766403	MultinomialNB	CountVectorizer
3	Base SupVector	0.769772	0.773258	LinearSVC	CountVectorizer
4	Sup Vec with TFIDF	0.796732	0.804168	LinearSVC	TfidfVectorizer
5	SVC/TFIDF and english stopwords	0.800698	0.806605	LinearSVC	TfidfVectorizer
6	SVC/TFIDF and common stopwords	0.794210	0.802187	LinearSVC	TfidfVectorizer
7	SVC/TFIDF combined stopwords	0.794670	0.801275	LinearSVC	TfidfVectorizer
8	Neural Net	0.625000	NaN	NeuralNet	padded_sequence

RECOMMENDATIONS & FUTURE WORK



More Training Data

- cases from other courts and jurisdictions



Use advanced pre-trained models

- BERT



Predict other targets.

- Judges Ideological leanings and voting tendencies.



THANKS!

Do you have any questions?

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