

Classifying Federal Register Documents by Type

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DATS 6103 - Summer 2022

What is the Federal Register?

- Daily journal of the U.S. government
- Print and [online](#)
- 4 main sections
- 4 document types:
 - Notices
 - Proposed Rules
 - Final Rules
 - Presidential Documents



FEDERAL REGISTER
The Daily Journal of the United States Government



The Problem: Uncategorized Documents

- Particularly an issue in the 1990s data
- Available data from 1994-1999:
 - 201,591 total documents
 - 166,031 categorized as 1 of 4 main types
 - 32,468 lack "type" labels
- Problem: Severe undercounting for analysis
- Solution: use labeled documents to build classifier for document type

Revision of Fee Schedules; Fee Recovery for Fiscal Year 2022



A Rule by the Nuclear Regulatory Commission on 06/22/2022

PUBLISHED DOCUMENT

Start Printed Page 37197



AGENCY:

Nuclear Regulatory Commission.

ACTION:

Final rule.

SUMMARY:

The U.S. Nuclear Regulatory Commission (NRC) is amending the licensing, inspection, special project, and annual fees charged to its applicants and licensees. These amendments are necessary to implement the Nuclear Energy Innovation and Modernization Act, which requires the NRC to recover, to the

DOCUMENT DETAILS

Printed version:

[PDF](#)

Publication Date:

06/22/2022

Agency:

[Nuclear Regulatory Commission](#)

Dates:

This final rule is effective on August 22, 2022.

Effective Date:

08/22/2022



Document Type:

Rule

Document Citation:

87 FR 37197

Compatibility of Wireless Services With Enhanced 911



An Uncategorized Document by the [Federal Communications Commission](#) on 06/28/1999



PUBLISHED DOCUMENT

The full text of this document is currently available in [PDF format](#).

The full text of this document is also available in a [basic text format](#).



DOCUMENT DETAILS

Printed version:

[PDF](#)

Publication Date:

06/28/1999

Agency:

[Federal Communications
Commission](#)

Dates:

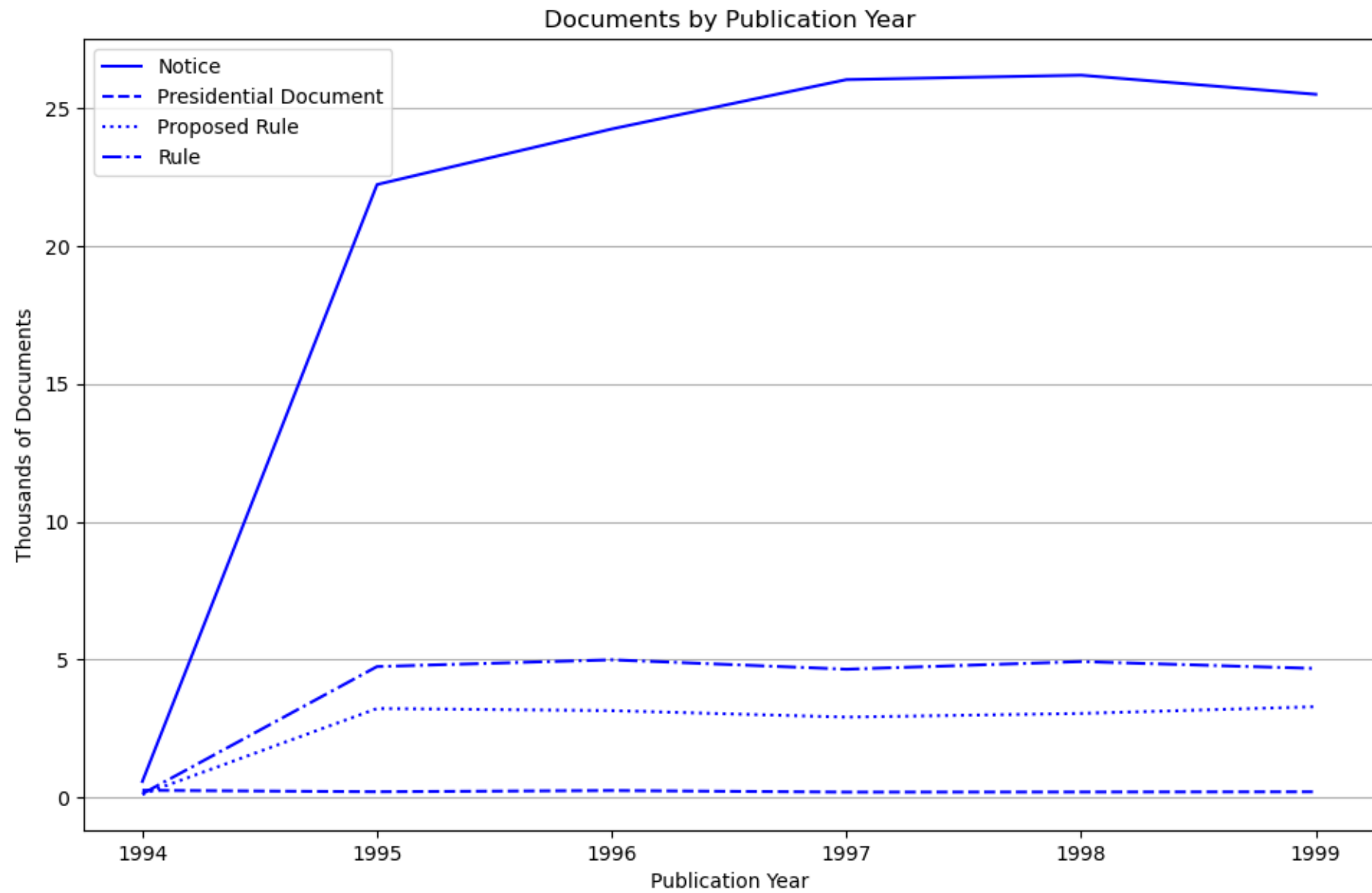
Effective July 28, 1999. This document contains new information collections subject to the Paperwork Reduction Act of 1995 (PRA), which are pending OMB approval. A notice will be placed in the [Federal Register](#) when OMB

Data

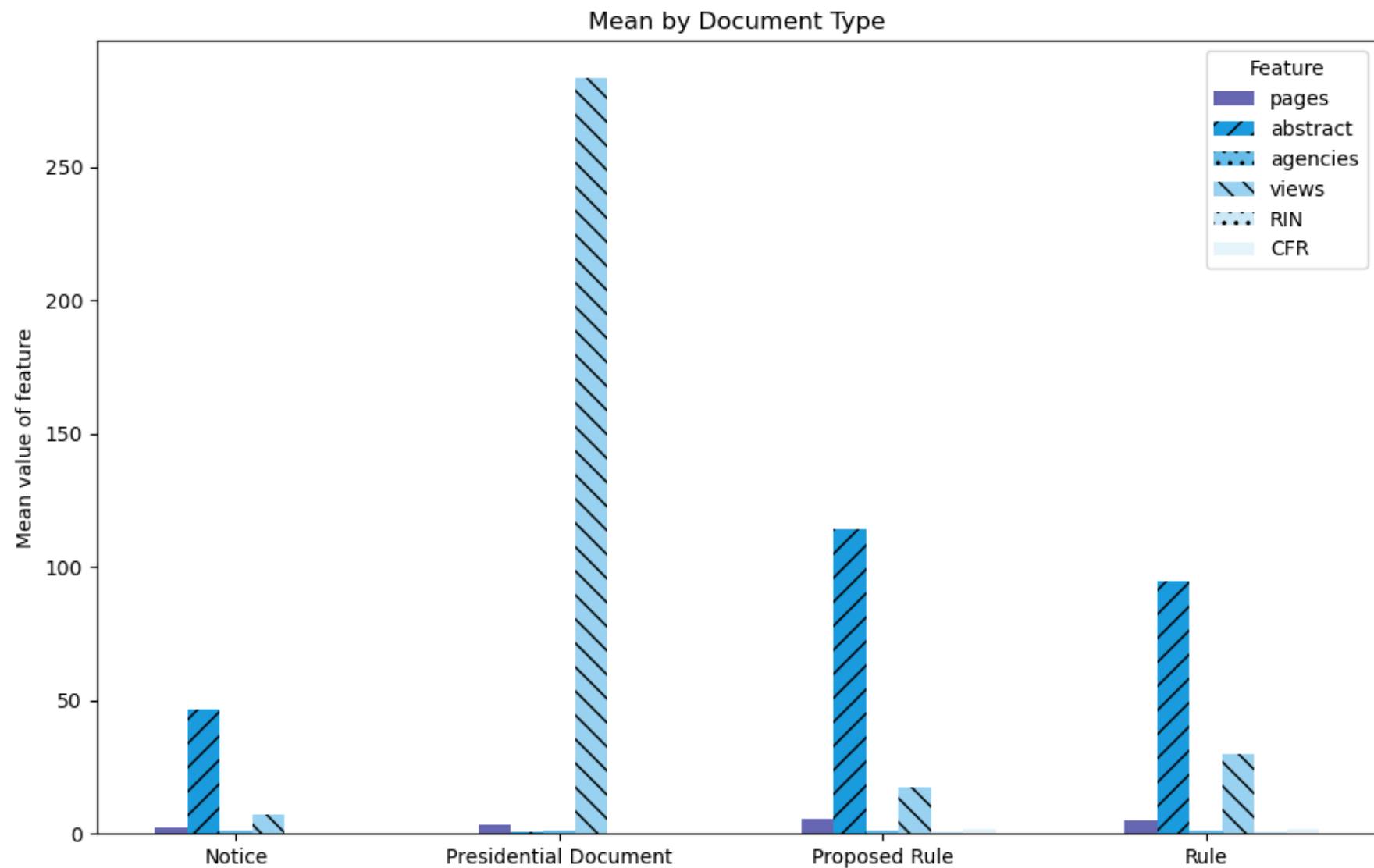
Processed dataset:

- 166,031 documents
- 6 numeric variables
- 5 categorical variables
- 3 text variables

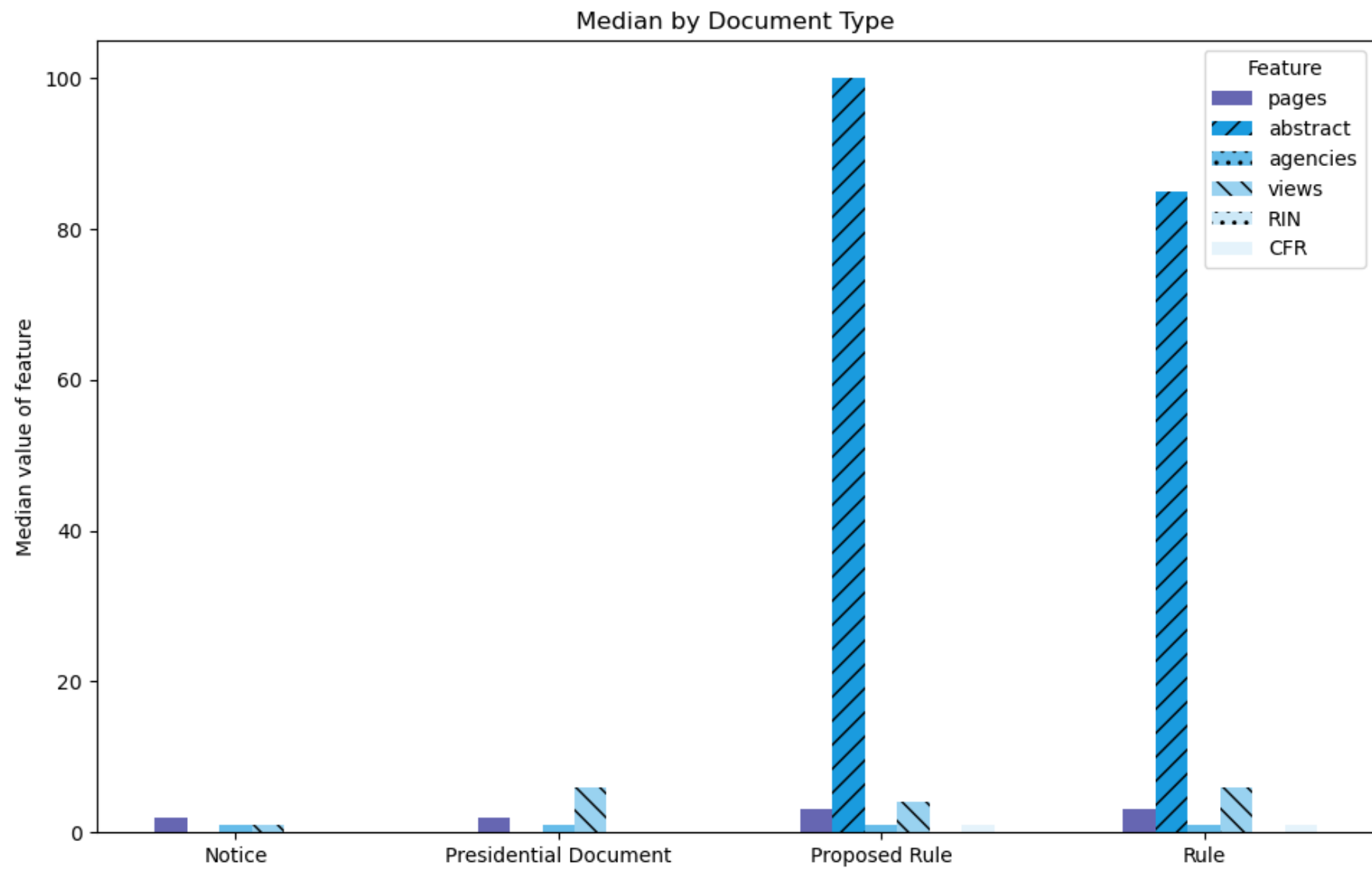
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 166031 entries, 0 to 166030
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   page_length           166031 non-null float64
1   agencies_count_uq     166031 non-null int64
2   abstract_length       166031 non-null int64
3   page_views_count      166031 non-null int64
4   RIN_count             166031 non-null int64
5   CFR_ref_count         166031 non-null int64
6   sig                  166031 non-null float64
7   effective_date_exists 166031 non-null int64
8   comments_close_exists 166031 non-null int64
9   docket_exists         166031 non-null int64
10  eop                  166031 non-null int64
11  action               166031 non-null object
12  abstract              166031 non-null object
13  title                166031 non-null object
dtypes: float64(2), int64(9), object(3)
memory usage: 17.7+ MB
```



Source: Federal Register API and authors' calculations.



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Preprocessing

- Data cleaning
- Imputing missing values
- Create/extract new variables from document metadata
- Numeric transformer (min-max scaler)
- Categorical transformer (one-hot encoder)
- Label encoder for target
- Text feature extraction (tf-idf vectorizer)
- Train-test split: 70%-30% (116,221 vs. 49,810)

Modeling

1. Complement Naïve Bayes (categorical and numeric)
2. AdaBoost (categorical and numeric)
 - 1000 Complement NB estimators
3. Voting Classifier (categorical and numeric)
 - Complement NB
 - Logistic Regression (balanced weights)
 - KNN → grid search for $k \in [5, 99, 341]$
4. Complement NB (text)
 - tf-idf: term-frequency * inverse document-frequency

Confusion Matrix: Model 1

True label	0	37331	11	33	30
	1	6	400	0	0
	2	3046	0	1537	157
	3	4957	0	120	2182
		0	1	2	3
		Predicted label			

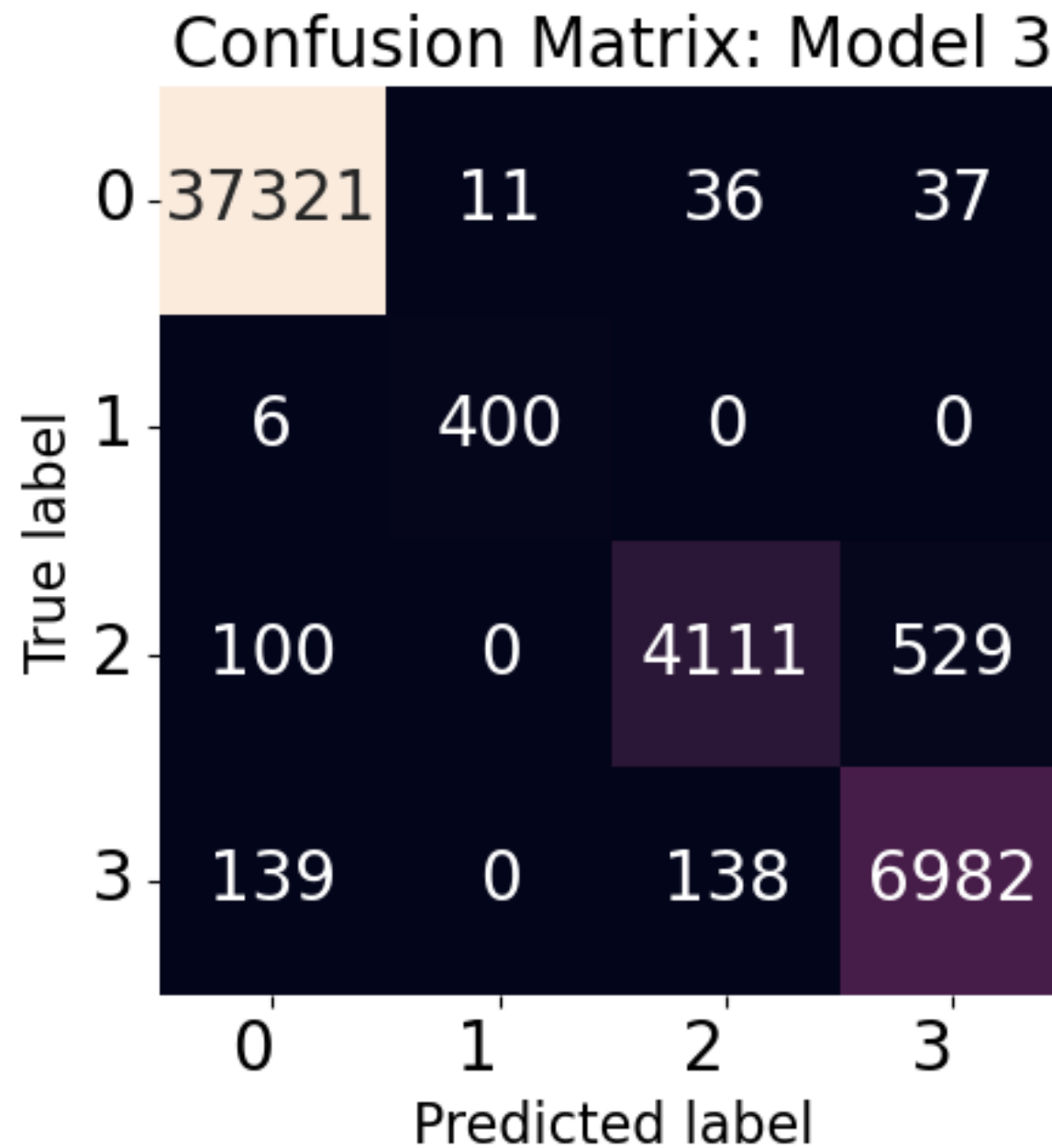
Complement NB
(categorical/numeric)

AdaBoost
(1000 NB estimators)

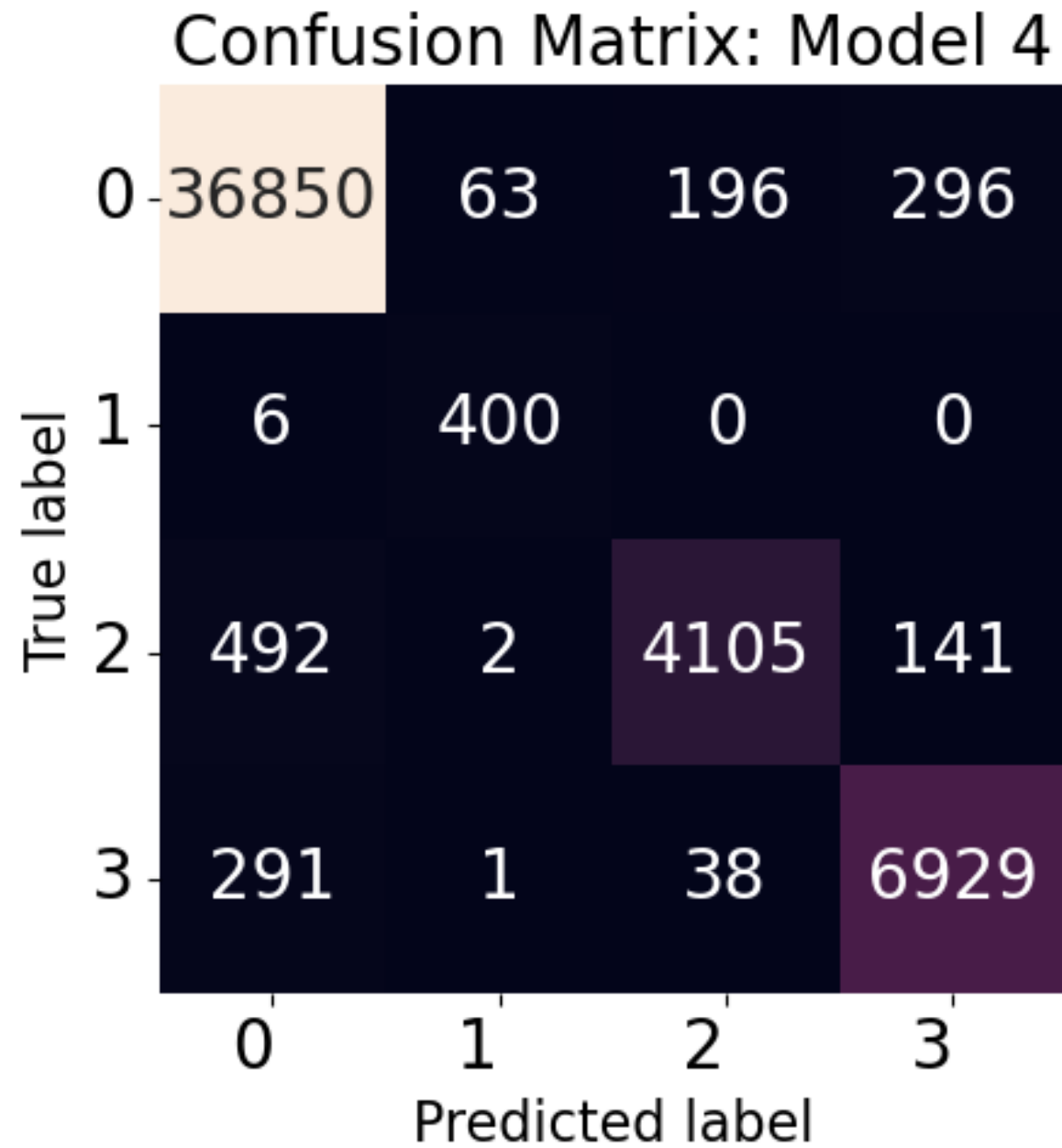
Confusion Matrix: Model 2

True label	0	1	2	3
0	37342	21	12	30
1	406	0	0	0
2	3048	91	1446	155
3	5094	71	18	2076
	0	1	2	3
	Predicted label			

**Voting Classifier
(NB, Logit, KNN)**



**Complement NB
(tf-idf vectors)**



<i>Model</i>	<i>Classifiers</i>	<i>Features</i>	<i>Accuracy</i>	<i>F1-Score</i>
1	NB	5 categorical 6 numeric	0.832	0.797
2	Boost 1000 * NB	5 categorical 6 numeric	0.820	0.781
3	Hard voting NB, Logit, KNN	5 categorical 6 numeric	0.980	0.979
4	NB	1 text	0.969	0.969

Improvements

- Consider different hyper-parameters for AdaBoost (1000x too high?)
- Integrate text features with categorical/numeric features
- Analyze full text of documents → tf-idf vectorizer
- Clean up my code...
- Any other suggestions?

Thank you for listening!

Questions?