



# Hierarchical Approaches to Text-based Offense Classification

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### Motivation



Offense classification plays critical roles for society, research, and public administration:

- 1. Criminal background checks (employment, public benefit eligibility, security credentialing, firearm purchases)
- 2. Evaluating sentencing disparities, testing behavioral theories, etc
- 3. Public funding, resource allocation, election outcomes

Yet, no comprehensive standard currently exists to cover the universe of illicit activity nor a mechanism to map free-entry descriptions into systematic codes, leading to:

- More discretion with little oversight
- Inconsistent definitions and mappings
- Irreproducible research



In this paper, we do the following:

- 1. Introduce the Uniform Crime Classification Standard (UCCS)
  - Schema satisfying 4 design principles laid out by the National Academy of Sciences
- 2. Develop the Text-based Offense Classification (TOC) tool
  - Hierarchical machine learning model trained on over 300k hand coded training records
- 3. Experiment with model variation in TOC to assess determinant of prediction success
  - Help identify potential lessons for other classification applications

## Four design principles for offense classification



National Academy of Sciences (2016) laid out four goals of offense schema:

- 1. Encompass new and emerging crime types
- 2. Fully realized classification for statistical purposes
- 3. Attribute-based classification
- 4. Enable comparisons between jurisdictions across time

	Summary Report- ing System (SRS)	National Incident- Based Reporting System (NIBRS)	National Correc- tions Reporting Program (NCRP)	Uniform Crime Classification Standard (UCCS)
Principle 1:	х	$\checkmark$	X	$\checkmark$
Principle 2:	x	×	×	$\checkmark$
Principle 3:	$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
Principle 4:	×	×	×	$\checkmark$

# Uniform Crime Classification Standard

#### Each UCCS code has four digits:

- First digit  $\rightarrow$  Broad code
- Digits 2 and 3  $\rightarrow$  Offense code
- Fourth digit  $\rightarrow$  Offense modifier

#### Example UCCS codes:

UCCS Code/Description		Broad Code/Description		Offense Code/Description		Offense Modifier Code/Description	
1010	Murder	1	Violent	01	Murder	0	
1011	Attempted Murder	1	Violent	01	Murder	1	Attempt
1012	Conspiracy to Commit Murder	1	Violent	01	Murder	2	Conspiracy
1020	Unspecified Homicide	1	Violent	02	Unspecified homicide	0	
1021	Unspecified Homicide, Attempted	1	Violent	02	Unspecified homicide	1	Attempt
1022	Unspecified Homicide, Conspiracy	1	Violent	02	Unspecified homicide	2	Conspiracy
1030	Voluntary Manslaughter	1	Violent	03	Voluntary/nonnegligent manslaughter	0	
1031	Voluntary Manslaughter, Attempted	1	Violent	03	Voluntary/nonnegligent manslaughter	1	Attempt
1032	Voluntary Manslaughter, Conspiracy	1	Violent	03	Voluntary/nonnegligent manslaughter	2	Conspiracy
1040	Vehicular Manslaughter	1	Violent	04	Voluntary/nonnegligent manslaughter	0	
1041	Vehicular Manslaughter, Attempted	1	Violent	04	Voluntary/nonnegligent manslaughter	1	Attempt
1042	Vehicular Manslaughter, Conspiracy	1	Violent	04	Voluntary/nonnegligent manslaughter	2	Conspiracy





TOC tool maps free-entry offense description fields to UCCS codes

Built through a partnership between Measures for Justice (MFJ) and CJARS

- MFJ hand coded significant volume of offense descriptions, creating training data to develop TOC
- 313,209 hand-coded offense descriptions from 24 states

Freely available to general public online: https://cjars-toc.isr.umich.edu

### Text-based Offense Classification tool





INSTRUCTIONS



DOCUMENTATION

METHODOLOGY REGISTRATION



Designed to harmonize criminal justice offense codes. If you are working with a dataset that contains thousands of rows of cases, each with its own unique charge description associated with it, TOC is a game changer.

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Citation Publications and research reports based on the outputs from TOC should be cited as: Choi, J., Kilmer, D., Mueller-Smith, M., & Taheri, S. (2022). Hierarchical Approaches to Text-based Offense Classification. Unpublished Working Paper.

Contact For any questions, concerns, or inquiries, please reach out directly to the TOC development team (cjarstoc@unich.edu)

### Prediction models





### Prediction models: flat





### Prediction models: flat





### Prediction models: hierarchical





### Prediction models: hierarchical





### Model details



- 75% 25% random split of hand-coded data for training vs testing purposes
  Avoids known problems arising from data leakage
- Modest preprocessing: remove articles, punctuation, capitalization, word normalization
- Tokenization: 4-grams
- Feature selection: Term Frequency-Inverse Document Frequency
- Optimization algorithm: neural network
  - Multi-layer perceptron model with 1 hidden layer and 100 neurons

#### Feature extraction via n-grams



(a) Contiguous sequence of 4 characters

theft from person > 65 value >300k<10k

theft from person > 65 value >300k<10k

th<mark>eft</mark> from person > 65 value >300k<10k ...

theft from person > 65 value > 300 k < 10 k

theft from person > 65 value >300<mark>k<10</mark>k

theft from person > 65 value > 300k<mark><10k</mark>

(b) Extracted features

'thef', 'heft', 'eft ', 'ft f', 't fr', ' fro', 'from', 'rom ', 'om p', 'm pe', ' per', 'pers', 'erso', 'rson', 'son>', 'on>6', 'n>65', '>65', '65 v', '5 va', 'val', 'valu', 'alue', 'lue>', 'ue>3', 'e>30', '>300', '300<', '00<1', '0<10', '<10k', ....





We evaluate three related standard performance statistics:

- Precision: % of positive predictions that are correctly labelled
- Recall: % of labels that are correctly made
- F1 Score: harmonic mean of precision and recall

Model performance is assessed at both the broad crime type and full UCCS code levels



	Broad Crime Type			Full UCCS Code			
	Precision	Recall	F1 Score	Precision	Recall	F1 Score	
All Crime Types	0.983	0.983	0.983	0.963	0.963	0.963	
Broad Crime Type Code:							
Violent	0.997	0.994	0.995	0.993	0.989	0.991	
Property	0.927	0.990	0.957	0.884	0.944	0.913	
Drug	0.999	0.960	0.979	0.862	0.828	0.845	
DUI	0.987	0.986	0.986	0.942	0.941	0.941	
Public Order	0.993	0.938	0.965	0.977	0.923	0.949	
Criminal Traffic	0.987	0.991	0.989	0.986	0.991	0.988	

## Out-of-state performance





Full crime type



13/18

## Comparison with existing resources



Distribution of offense types for felony charges in CJARS



## Comparison with existing resources



Distribution of offense types for misdemeanor charges in CJARS



# Evaluating optimal model parameterization



- Variation in size of training data: 5,000 200,000 training observations (ink)
- Feature unit (n-grams, bag-of-words) and selection mechanism (Count Vectorizer vs. TF-IDF)
- Total number of features: 100 10,000 selected features (ink)
- Machine learning algorithm: Random Forest vs. Neural Network 💷



- 1. Expand geographic coverage of training data to non-covered jurisdictions
- 2. Crowdsource prediction errors to improve corpus of training records via TOC portal
- 3. Incorporate feasibility to process cited statute numbers
  - Requires comprehensive database mapping statutes to offense descriptions
- 4. Develop mechanism to evaluate updates to schema

# Summary and conclusion



This project introduces the UCCS schema and the TOC tool to map free entry offense descriptions to standardized codes

Goal is to raise the bar for research and statistical reporting:

- Reduce researcher discretion as well as data wrangling burden
- Improve common definitions and comparability across jurisdictions
- Expand coverage of historically understudied (but large!) portion of the justice system: misdemeanor caseload
- Lower research barriers to increase diverse perspectives on justice system

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# Appendix Slides

# Variation in size of training data



(b) Full UCCS Code

(a) Broad Crime Type



#### Feature unit selection mechanism



(b) Full UCCS Code

(a) Broad Crime Type



### Total number of features



(a) Broad Crime Type



(b) Full UCCS Code

## Machine learning algorithm



(a) Broad Crime Type



(b) Full UCCS Code