Deceiving Authorship Detection

Tools to Write Anonymously & Current Trends in Adversarial Stylometry.

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26C3/28C3 Diff

- Review & Updated Analysis of 26C3 Material
 - New Corpus (45 authors)
 - New Method (Writeprints)
 - Much more robust results.
- The tools we discussed are now built!
 - JStylo
 - Anonymouth
- Detecting Deception in Adversarial Writing



An Overview

- What is "Authorship Recognition" and "Adversarial Stylometry"?
- What is the anonymity threat?
- Analyzing & Deceiving Authorship Recognition
- Two Tools
 - JStylo
 - Anonymouth
- Detecting Deception



What is Authorship Recognition?

- The basic question: "who wrote this document?"
- **Stylometry**: The study of attributing authorship to documents based only on the linguistic style they exhibit.
 - "Linguistic Style" Features: sentence length, word choices, syntactic structure, etc.
 - Handwriting, content-based features, and contextual features are not considered.
 - Individuals have unique writing styles because language is learned on an individual basis.
- In this presentation, stylometry and authorship recognition are used interchangeably.



What is Adversarial Stylometry?

- Adversarial Stylometry: Applying deception to writing style in order to affect the outcome of stylometric analysis.
 - But, is writing style modifiable? (Yes!)
 - Is it possible to deceive stylometry through altered writing style?
 (Yes!)
 - What are the implications of looking at stylometry in an adversarial context?



How Can Stylometry be a Threat?

- Supervised Stylometry
 - Given a set of documents of known authorship, classify a document of unknown authorship.
 - Hypothetical Scenario: Alice the Anonymous Blogger vs. Bob the Abusive Employer.
- Unsupervised Stylometry
 - Given a set of documents of unknown authorship, cluster them into author groups.
 - Hypothetical Scenario: Anonymous Forum vs. Oppressive Government.



Purely Hypothetical?

- Previous examples are purely hypothetical. What about a real example?
- From "Inside WikiLeaks" by Daniel Domscheit-Berg:
 - "I nudged Julian with my foot. We exchanged glances and started giggling. If someone had run WikiLeaks documents through such a program, he would have discovered that the same two people were behind all the various press releases, document summaries, and correspondence issued by the project. The official number of volunteers we had was also, to put it mildly, grotesquely exaggerated."





Adversarial Stylometry: A Review

- Understand the threat model
- Build a corpus.
- Evaluate current methods of stylometry against adversarial text.
- Analyze results and develop tools.



Threat Model

- Threat: Authorship recognition can identify you if there are sufficient writing samples and a set of suspects.
 - 6500+ words of training data per author
 - 500+ words of testing data
 - 50 or less suspects
 - These may be different:
 - Tweets (short messages)
 - Large numbers of authors (Writeprints)
- Old assumption: Writing style is invariant.
 - It's like a fingerprint, you can't really change it.



Circumvention Methods

 Challenge: conceive methods of circumventing writing style analysis.

Obfuscation

 An author attempts to write a document in such a way that their personal writing style will not be recognized.

Imitation

 An author attempts to write a document such that the writing style will be recognized as that of another specific author.

Translation*:

 Machine translation is used to translate a document to one or more languages and then back to the original language.



Building a Corpus

- Corpus = Dataset of documents.
- Data sets for adversarial stylometry do not exist. Participants are required to craft intentionally adversarial passages.
- Participation had three parts:
 - Submit 6500 words of pre-existing writing from a formal source.
 - Write a new 500 word obfuscation passage.
 - Task: Describe your neighborhood.
 - Write a new 500 word imitation passage.
 - Task: Imitate Cormac McCarthy, describe your day.
- Authors had no formal training or knowledge in linguistics or stylometry.



Brennan-Greenstadt Corpus

- 12 Individual Authors.
- Participants contacted through classes, colleagues, friends at Drexel University.
 - Motive for proper participation.
- One-on-one interaction with participants.
- Corpus is publicly available at https://psal.cs.drexel.edu
- Good for preliminary results, but we need something better.
 - Too small.
 - Too homogenous.



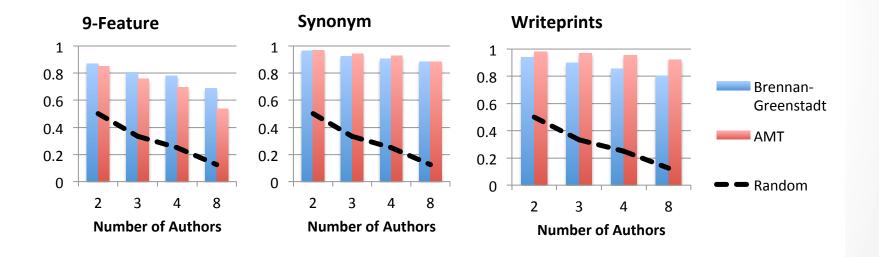
Building a Better Corpus with Amazon Mechanical Turk

- Drexel AMT Corpus
 - AMT = Amazon Mechanical Turk
- Same tasks as previous corpus.
- Only 45 of 101 of submissions are usable!
 - 45 Accepted Submissions.
 - Guidelines without spoiling data set. Must follow directions and:
 - Pre-existing writing must be formal in nature
 - Remove non-content
 - Minimal dialogue/quotations
 - Refrain from submitting: small samples, lab reports, Q&As, etc.
- Released today. Publicly available at https://psal.cs.drexel.edu
- This corpus is large, diverse, and unique.



Original vs. AMT Corpus

- AMT Corpus evaluated just as strongly as Drexel.
 - 9-Features does worse, Synonym does the same, Writeprints does better.





Evaluate Stylometry Methods Against the Corpus

- Three methods of Stylometry
 - 9-Feature / Neural Network
 - Synonym-Based Approach
 - Writeprints / SVM



Method 1: 9-Feature Set Neural Network

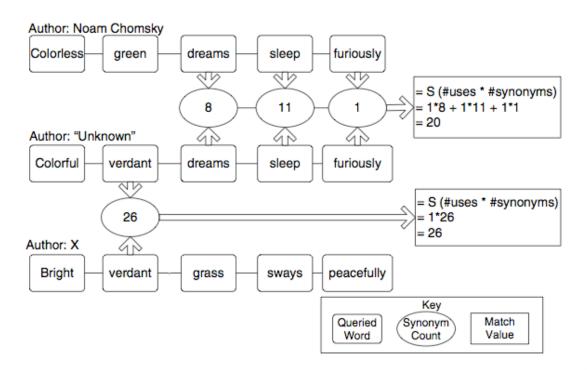
- Simple stylometric approach. Demonstrates potential effectiveness with a small number of obscure metrics.
- 9-Feature Set
 - Unique words, Complexity, Sentence Count, Average Sentence Length, Average Syllable Count, Character Count, Letter Count, Gunning-Fog Readability Index, Flesch Reading Ease Score.
- Neural Network Classifier.



Method 2:

Synonym-Based Approach

- Examines word choices when compared to available synonyms and frequency of use.
- Clark & Hannon, 2007.
- Good demonstration of single feature type stylometry.





Method 3: Writeprints (SVM)

- Based on the Writeprints approach by Abbasi & Chen, 2008.
 - Writeprints Baseline Feature Set.
 - Contains hundreds of features including character and word ngrams, function words, parts-of-speech tags, punctuation, and character level metrics.
- Support Vector Machine Classifier
 - Standard for multi-class classification in stylometry.
- Implementation of the full Writeprints approach uses a more extensive feature set and unique classification approach.

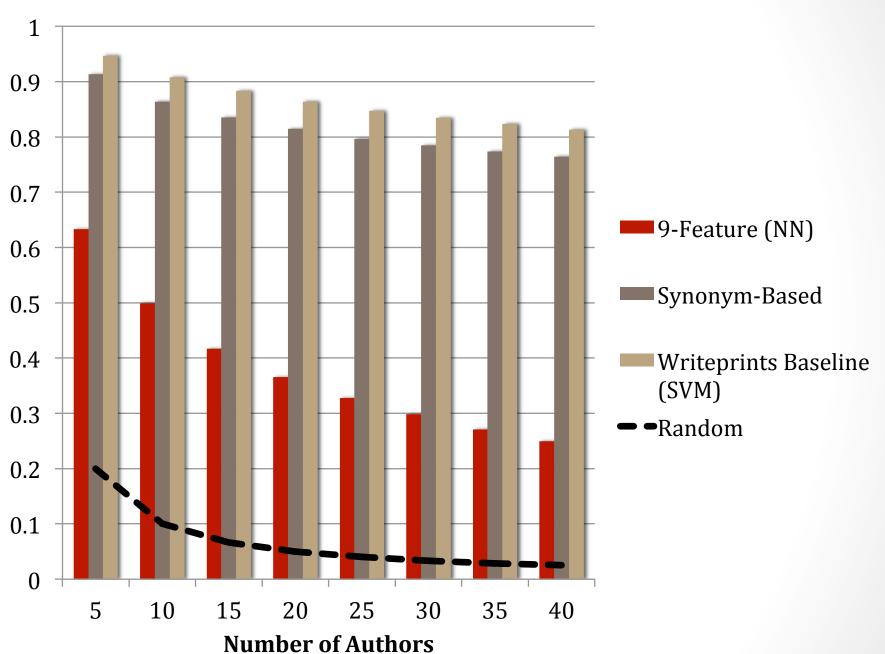


Detecting Authorship

- Four experiments.
 - Baseline: Evaluate the non-adversarial passages to establish a baseline accuracy for each method and feature set.
 - **Obfuscation**: Attempt classification of the obfuscation passages when classifiers are trained on the baseline data.
 - Imitation: Attempt classification of the imitation passages.
 - Imitation Success: Analyze the number of successful imitations.

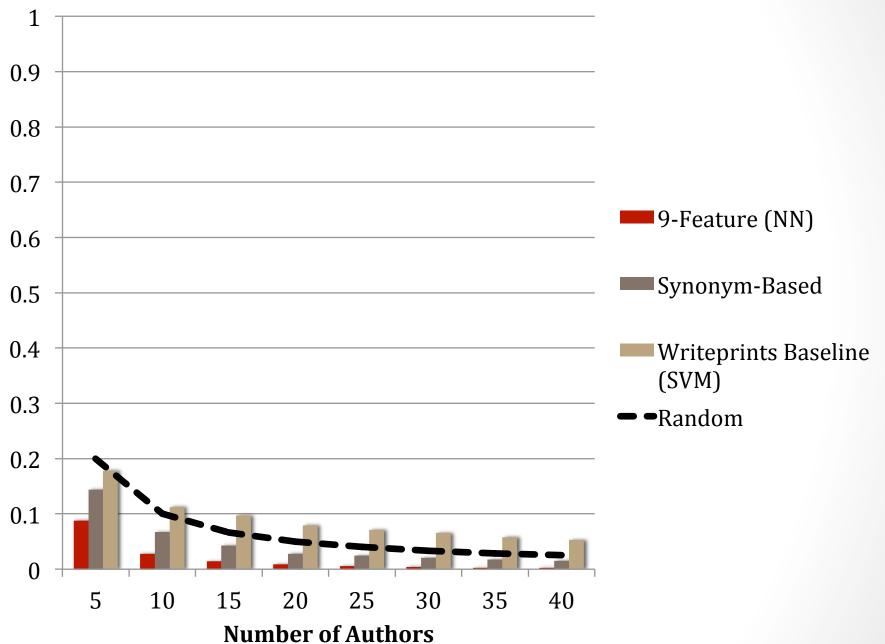


Baseline Precision



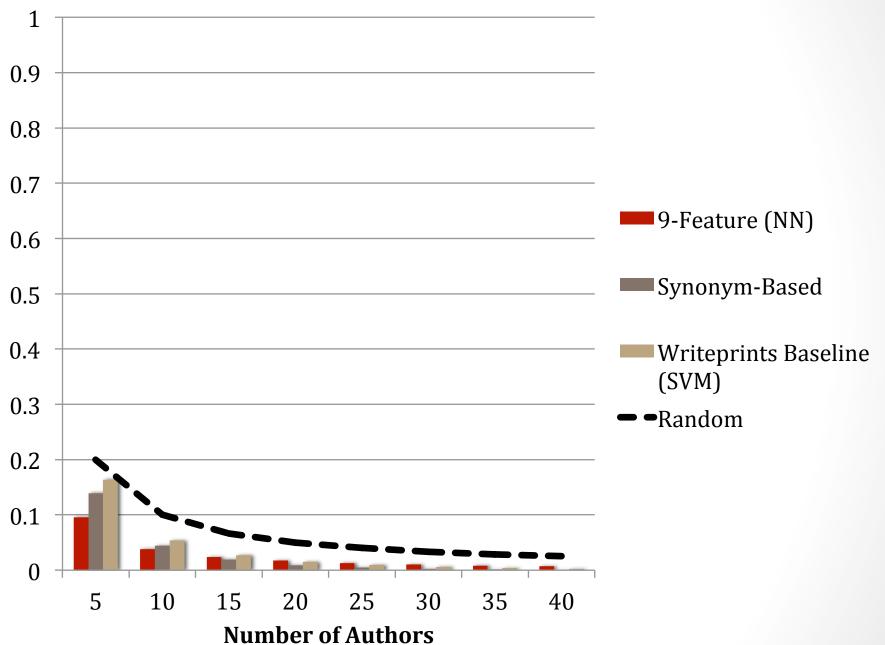


Obfuscation Precision



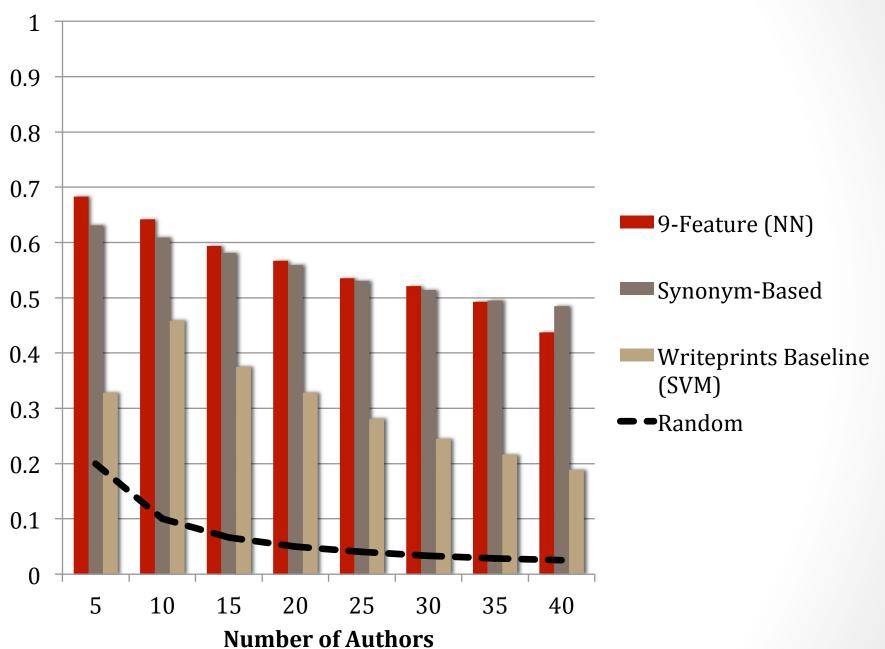


Imitation Precision





Imitation Success (Framing Cormac McCarthy)





Two Tools

- JStylo: Authorship Recognition Analysis Tool.
- Anonymouth: Authorship Recognition Evasion Tool.
- Free, Open Source. (GNU GPL)
- Alpha releases available today at https://psal.cs.drexel.edu
 - Migrating to GitHub soon.







JStylo: The Problem

- Stylometry-based research is difficult.
- Existing tools are good but limited.
 - Weka provides a suite of machine-learning classification tools.
 - Not tailored for text analysis no feature extraction ability.
 - Functions better as an API for software development.
 - JGAAP has a strong basic toolset for stylometry.
 - Limited in running multiple feature sets.
 - Strong API.
 - Extendable. Intended to be used this way.
- Nuances of stylometry are not easy to grasp.
- Many open research questions related to authorship. We need an easy-to-use tool that both researchers and nontechnical users can understand.



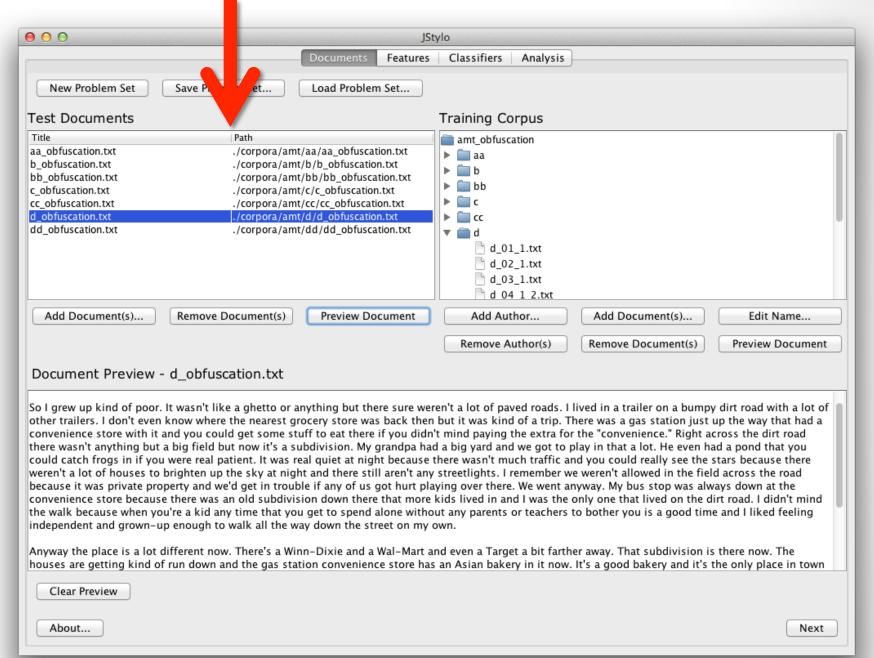
JStylo

- JStylo is an authorship recognition analysis tool. It is built upon a framework of:
 - JGAAP (Java Graphical Authorship Attribution Project)
 - Weka 3 Data Mining Software
- Features
 - Two existing adversarial corpora, featured in this presentation, and new corpus building functionality.
 - Wide selection of writing feature extractors and ability to add new extractors.
 - Wide selection of machine learning based classifiers.
 - Intuitive GUI.
- Alpha Release Available Now: https://psal.cs.drexel.edu

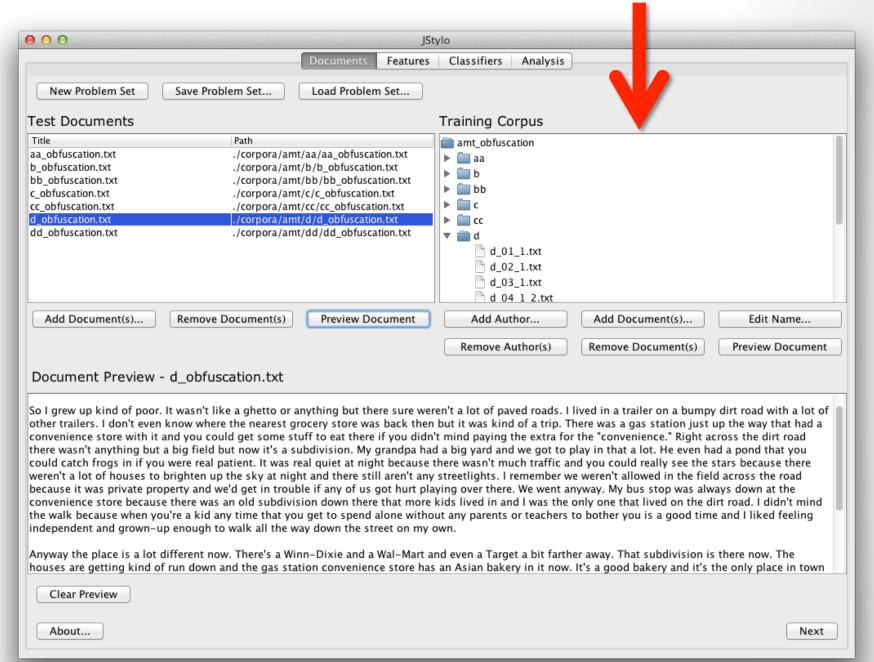


JStylo Demo





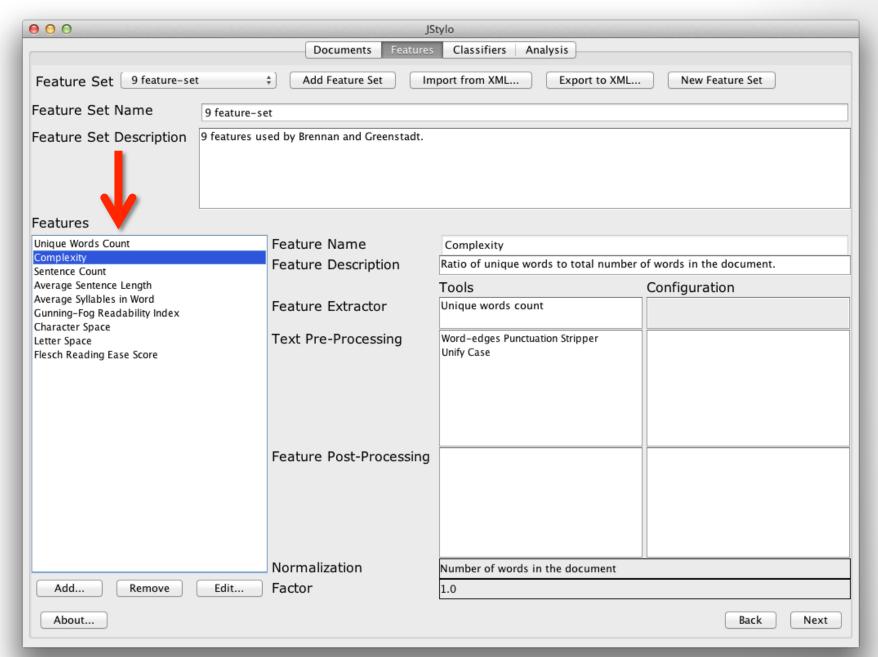




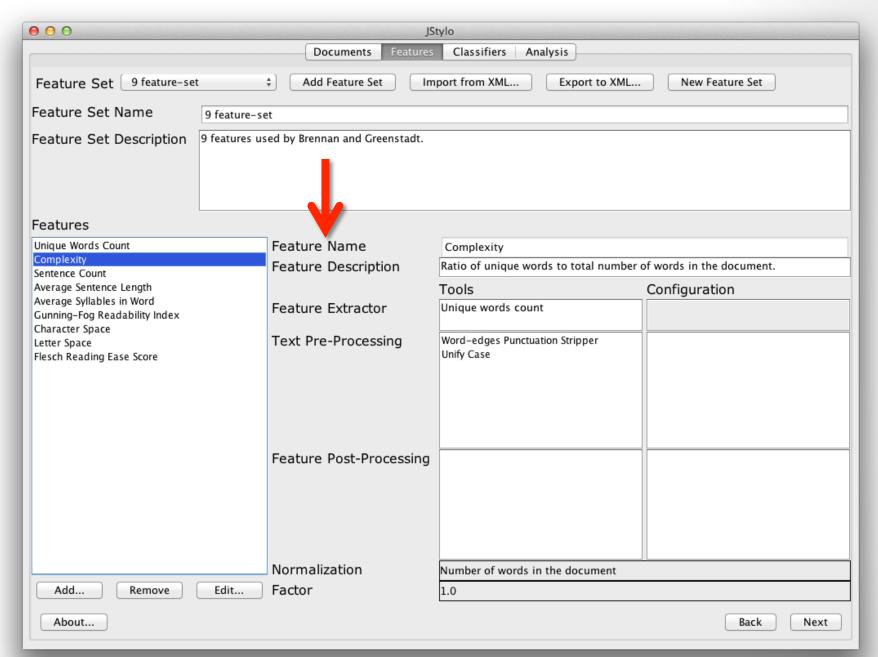


000	JStylo				
Documents Features Classifiers Analysis					
Feature Set 9 feature-set	Add Feature Set Import from XML Export to XML New Feature Set				
Feature Set WritePrints WritePrints (Limited)					
Feature Set Description					
Features					
	Feature Name				
	Feature Description				
		Tools	Configuration		
	Feature Extractor				
	Text Pre-Processing				
	Feature Post-Processing				
	Normalization				
Add Remove Edit	Factor				
About			Back Next		

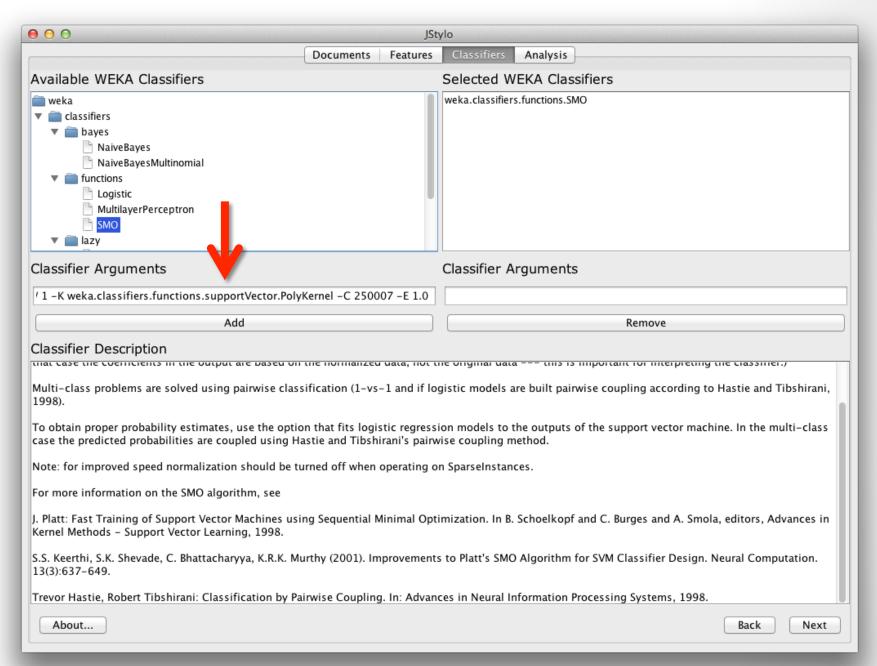




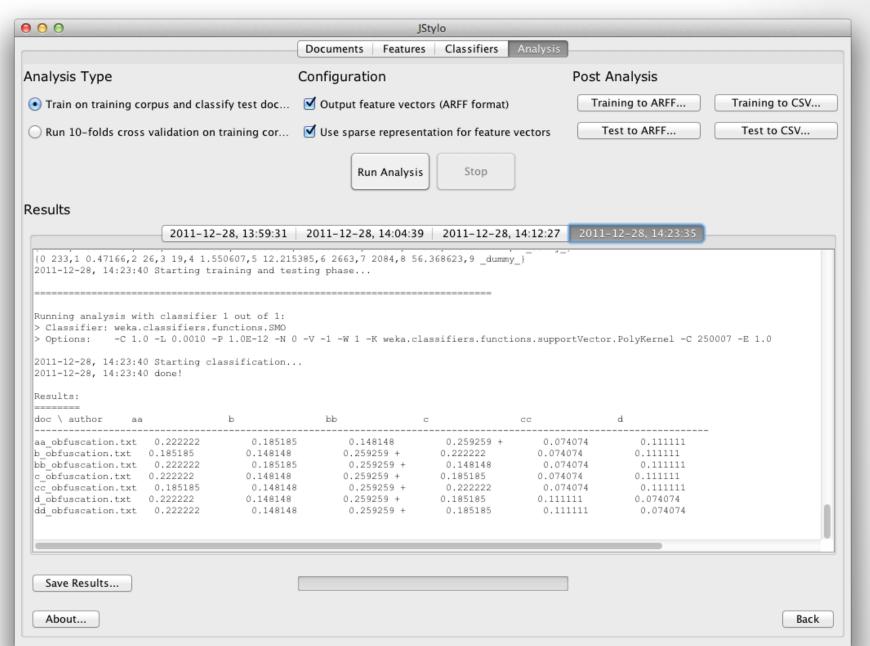








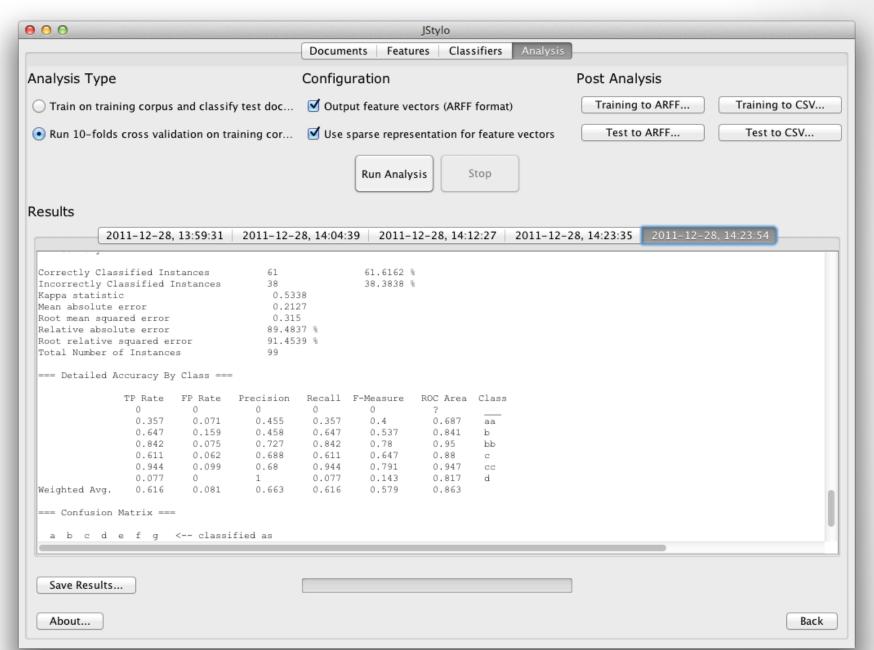






Results:				
======				
doc \ author aa		b	bb	C
aa obfuscation.txt	0.222222	0.185185	0.148148	0.259259 +
b obfuscation.txt	0.185185	0.148148	0.259259 +	0.222222
bb obfuscation.txt	0.222222	0.185185	0.259259 +	0.148148
c obfuscation.txt	0.222222	0.148148	0.259251 +	0.185185
cc obfuscation.txt	0.185185	0.148148	0.2592 9 +	0.222222
d obfuscation.txt	0.222222	0.148148	0.259259 +	0.185185
dd obfuscation.txt	0.222222	0.148148	0.259259 +	0.185185
_				







```
=== Summary ===
Correctly Classified Instances
                                  61
                                                    38.3838 %
                                   38
Incorrectly Classified Instances
                                    0.5338
Kappa statistic
Mean absolute error
                                     0.2127
Root mean squared error
                                    0.315
Relative absolute error
                                   89.4837 %
Root relative squared error
                                   91.4539 %
Total Number of Instances
                                    99
=== Detailed Accuracy By Class ===
                       FP Rate
                                Precision
                                          Recall F-Measure
                                                             ROC Area
                                                                      Class
             TP Rate
               0.357
                        0.071
                                  0.455
                                           0.357
                                                     0.4
                                                               0.687
                                                                       aa
               0.647
                        0.159
                                  0.458
                                        0.647
                                                     0.537
                                                               0.841
                                                                       b
               0.842
                        0.075
                                  0.727
                                        0.842
                                                     0.78
                                                               0.95
                                                                       bb
               0.611
                     0.062
                                  0.688
                                           0.611 0.647
                                                               0.88
                                                                       C
               0.944
                        0.099
                                           0.944 0.791 0.947
                                  0.68
                                                                       CC
                                            0.077
                                                    0.143 0.817
               0.077
                        0
                                                                       d
                                            0.616
                                                     0.579
Weighted Avg.
             0.616
                        0.081
                                                               0.863
```



JStylo Dev Goals

- Wider selection of classification methods and features.
 - Writeprints, Synonym-based, more Weka methods.
 - Ensemble classifiers, weighted averaging.
 - Greater pre and post-processing options.
- Easier to use and understand for non-technical users.
 - Adding an online tutorial.
 - GUI installs of new feature extractors and classifiers.
- Logging and graphing results over multiple experiments.
 - Visualization of documents, authors, and classifications.



Anonymouth: The Problem

- Authorship recognition can be a legitimate threat to privacy and anonymity.
- Intuition in changing writing style goes a long way, but may not be enough and may not be sustainable over multiple documents.
 - We already see methods that offer some resistance to adversarial passages.
- Fully automated text anonymization is an intractable problem.
 - We need a solution that explains authorship recognition nuances as needed and assists the authoring making the most useful changes towards anonymity.



Anonymouth

- Anonymouth is an authorship recognition circumvention tool.
 It is built upon a framework of:
 - JStylo (JGAAP & Weka)
 - WordNet
- Features
 - Corpora, feature extractor, and classifier functionality from JStylo.
 - Suggestion system for modifying documents to evade authorship detection. Ideal value for each feature is calculated, existence of the features is highlighted, user is assisted in changing them.
 - Iterative approach to anonymizing writing style.
 - Dictionary / Synonyms / Interactive Editing Console
- Alpha Release Available Now: https://psal.cs.drexel.edu

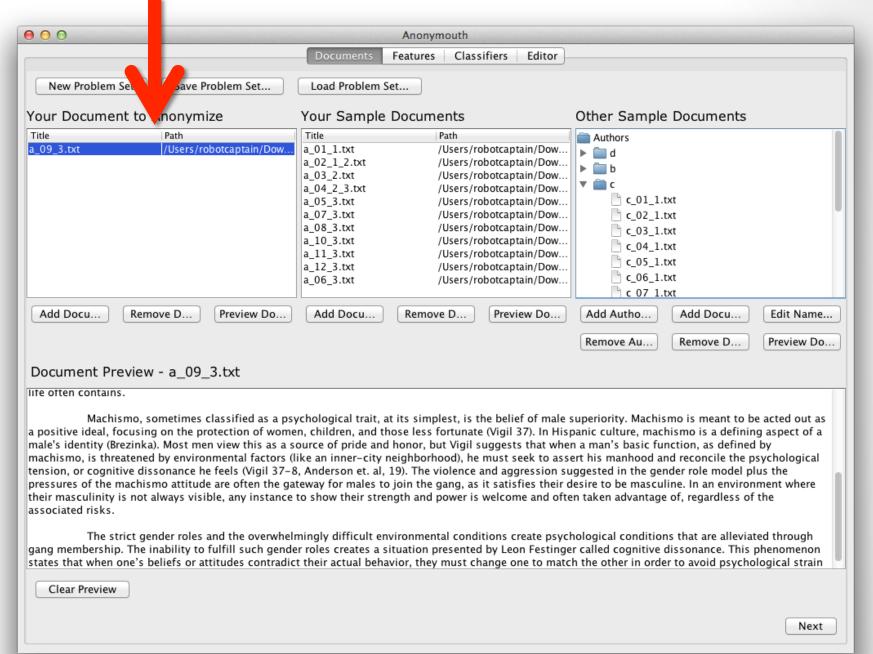


Anonymouth Demo

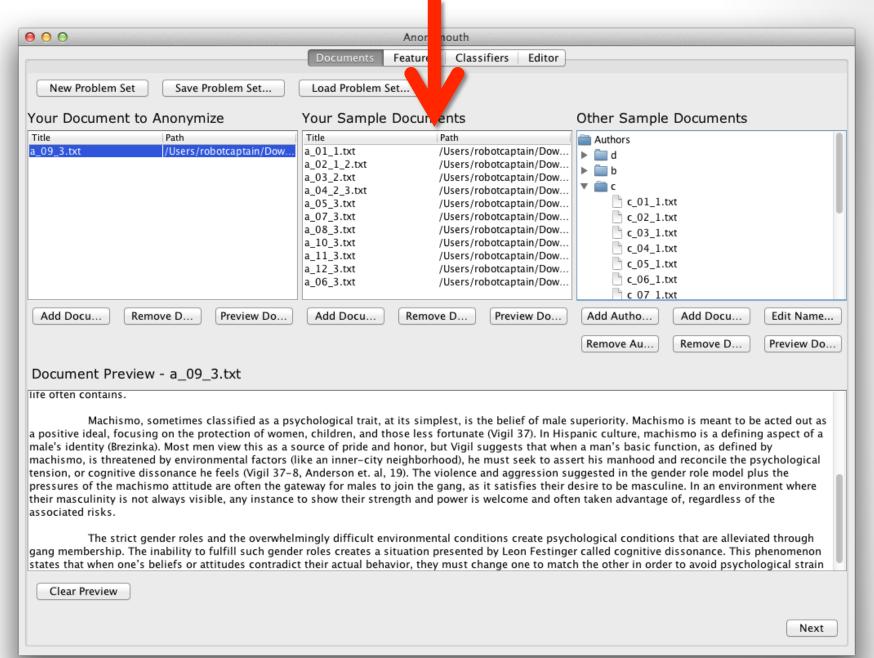


0 0	Anonymouth	
	Documents Features Classifiers Editor	
New Problem Set Save Problem Set	Load Problem Set	
Your Document to Anonymize	Your Sample Documents	Other Sample Documents
Title Path	Title Path	Authors
Add Docu Remove D Preview Do Document Preview	Add Docu Remove D Preview Do	Add Autho Add Docu Edit Name Remove Au Remove D Preview Do
Clear Preview		
		Next

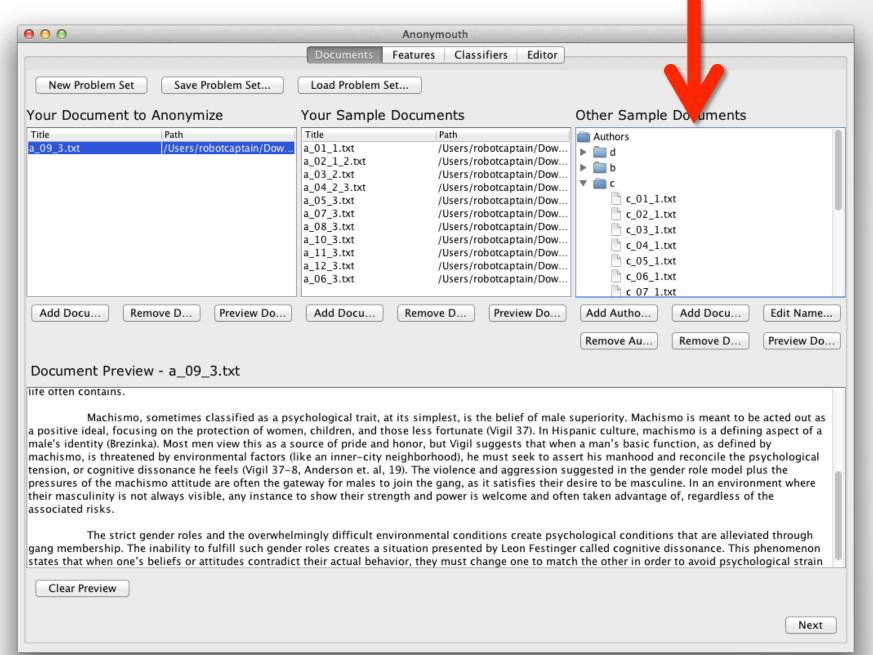




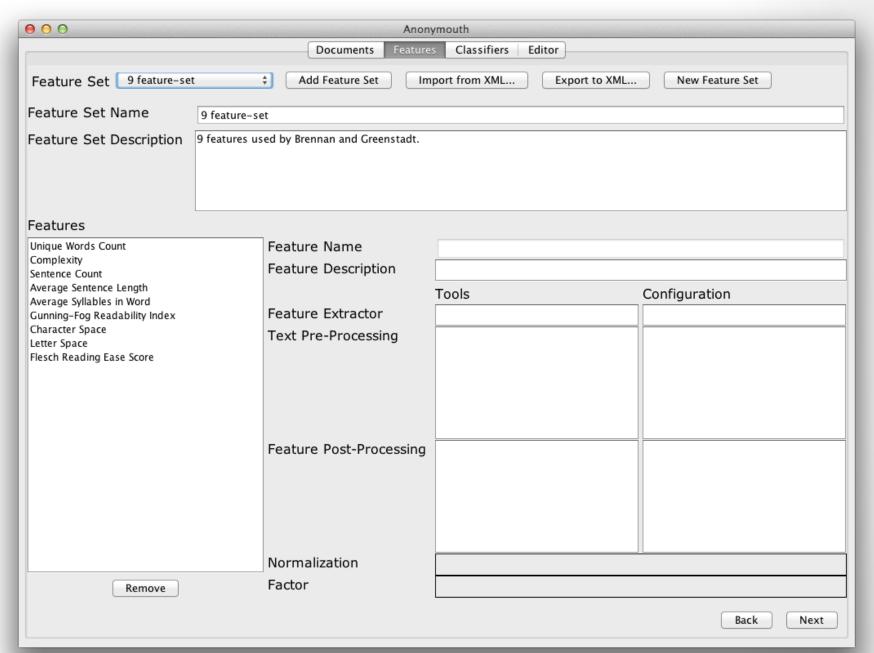




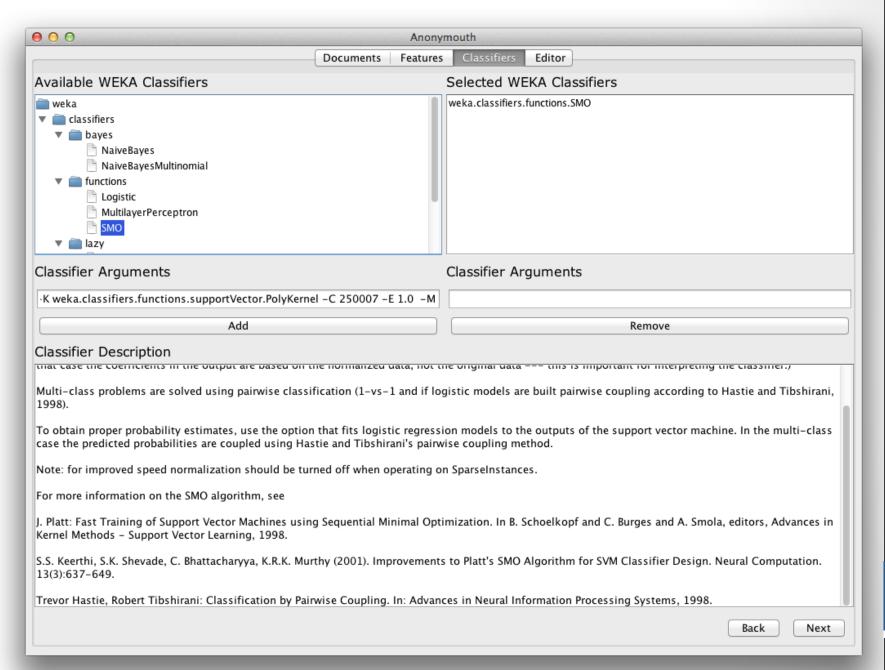




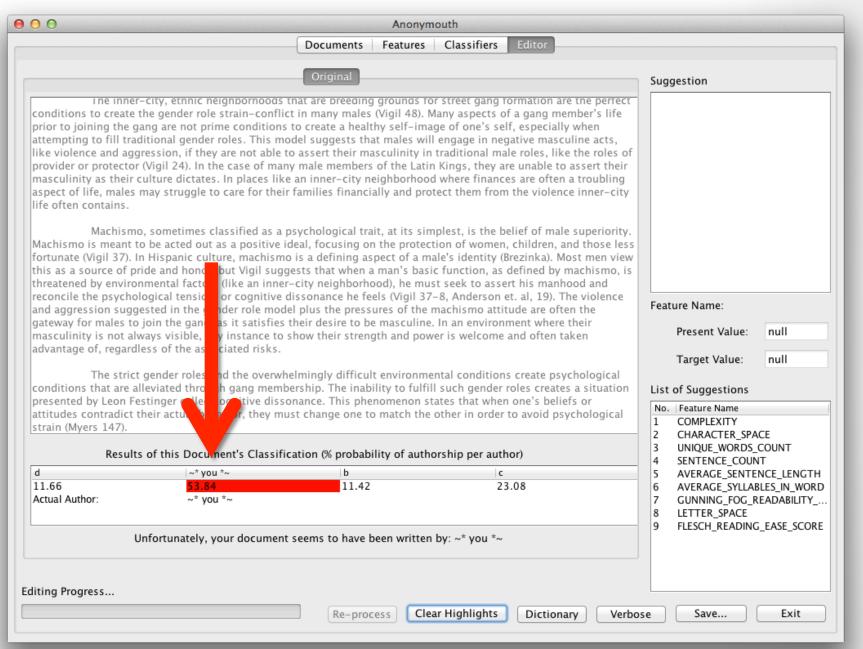




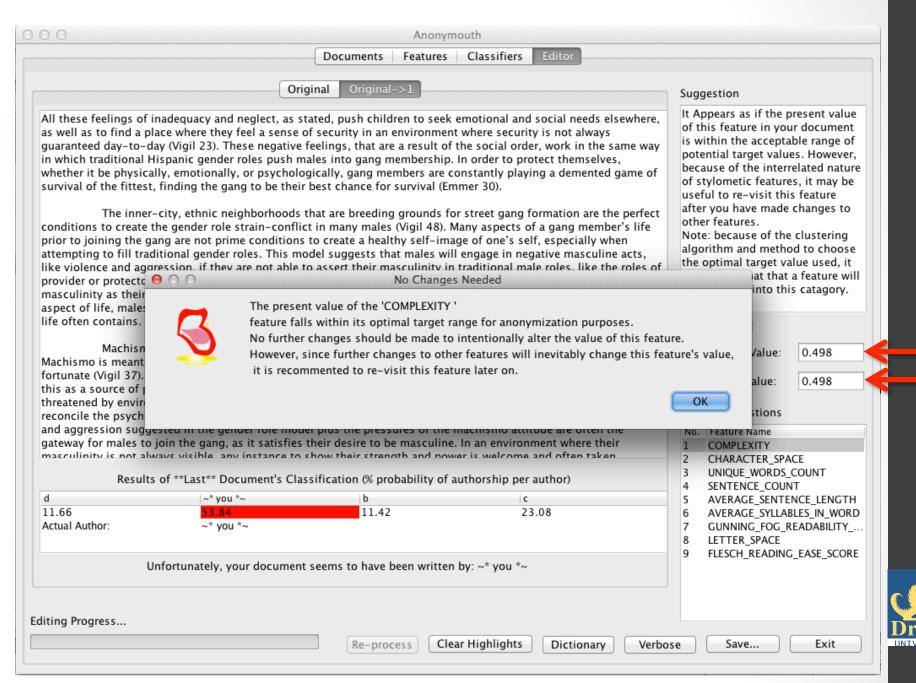


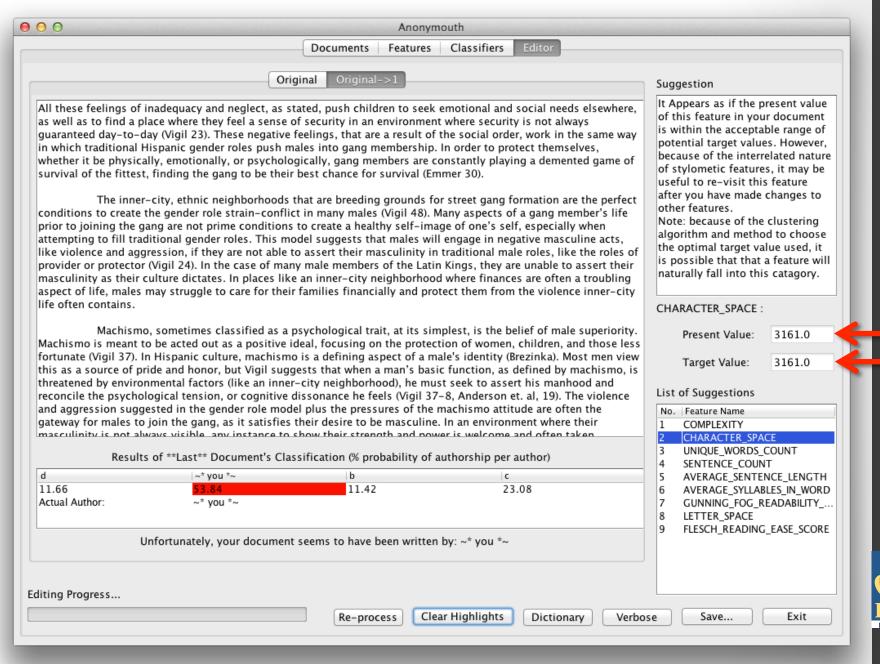


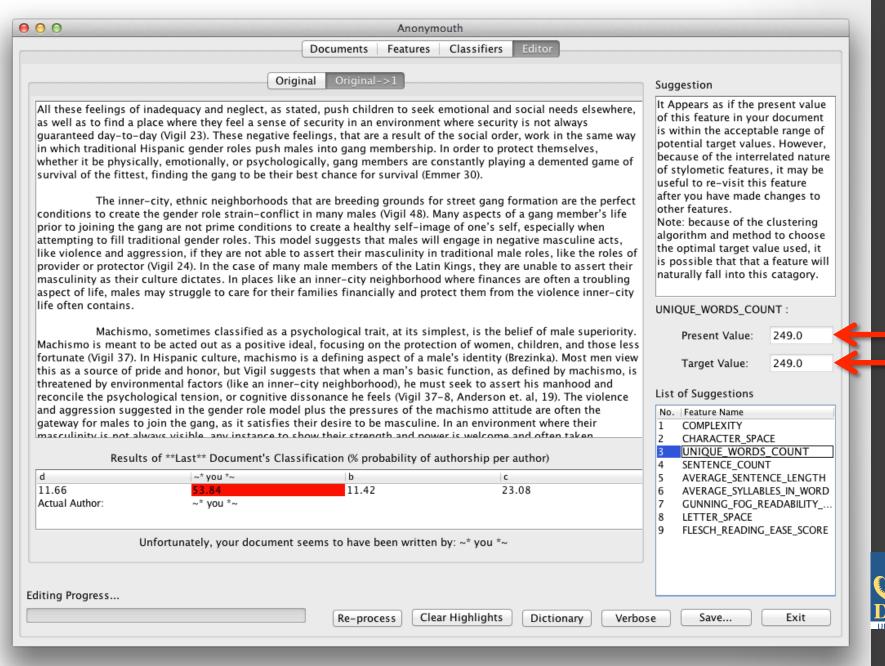


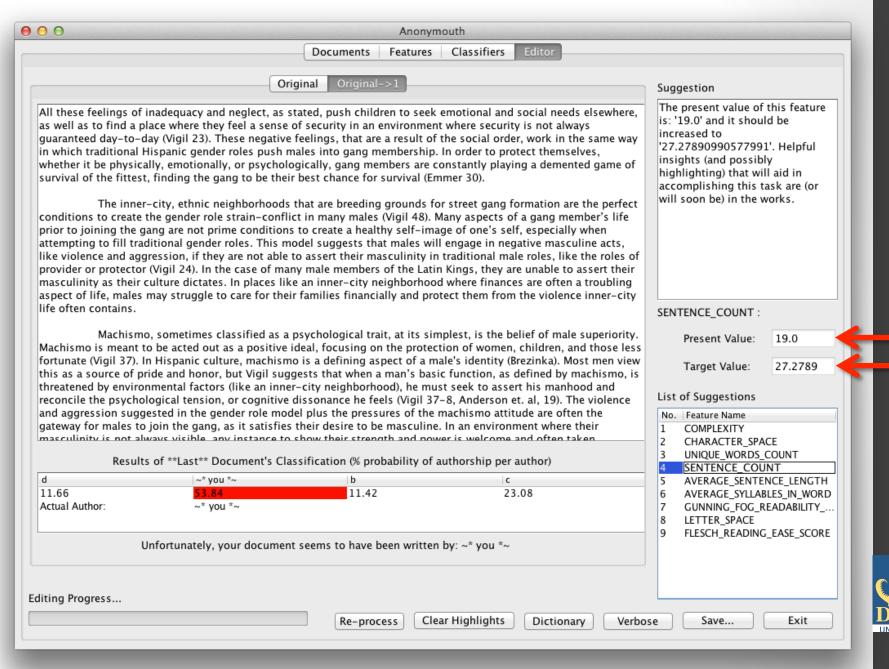


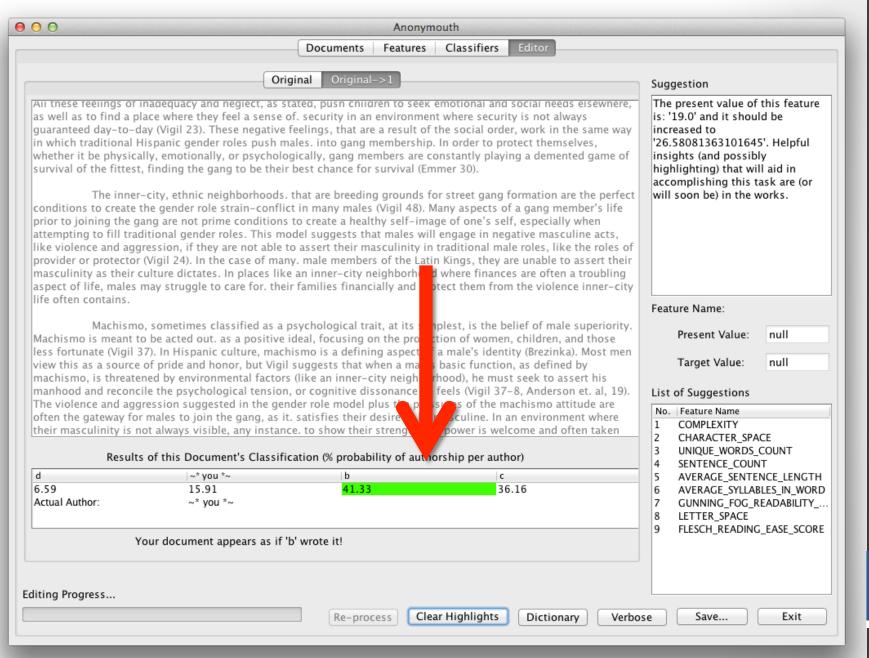




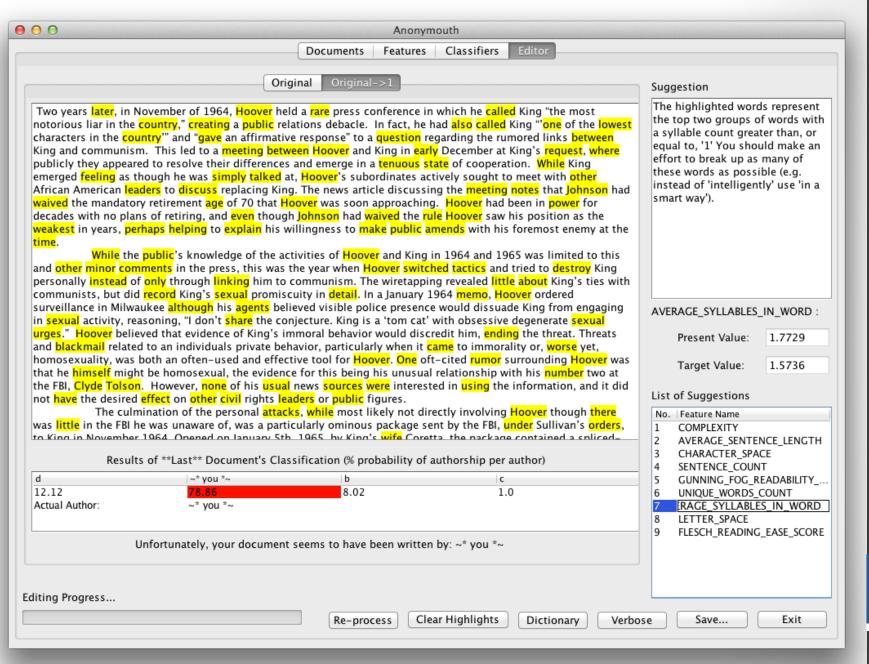














Anonymouth Challenges

- Features are often not independent.
 - Increasing the number of complex words will also increase average syllable count.
 - Reducing the number of times a specific word occurs will also affect the lexical density.
- How can we create an algorithm for anonymity that generates an obfuscated document with minimal effort and without circular feature modification?



Anonymouth Dev Goals

- Streamlined suggestion system.
 - Improved automation on applicable features.
 - Improved clustering algorithm to provide optimal path to anonymity.
- Improved editing interface.
 - Increased phrase and word synonym set support.
 - Edit by blocks of text, not simply feature-by-feature.
- Wider set of features and classification methods.
 - Multi-method and feature collection analysis.
- Usability and anonymity user studies.

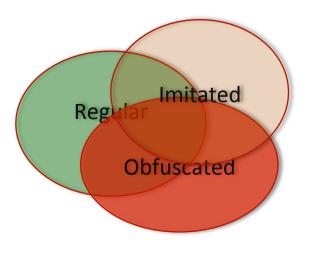


Opening Development

- Project will continue to be developed by PSAL at Drexel, but we welcome collaboration and participation.
- We are interested in...
 - Linguistic Experts
 - Security Advisors
 - UI Experts

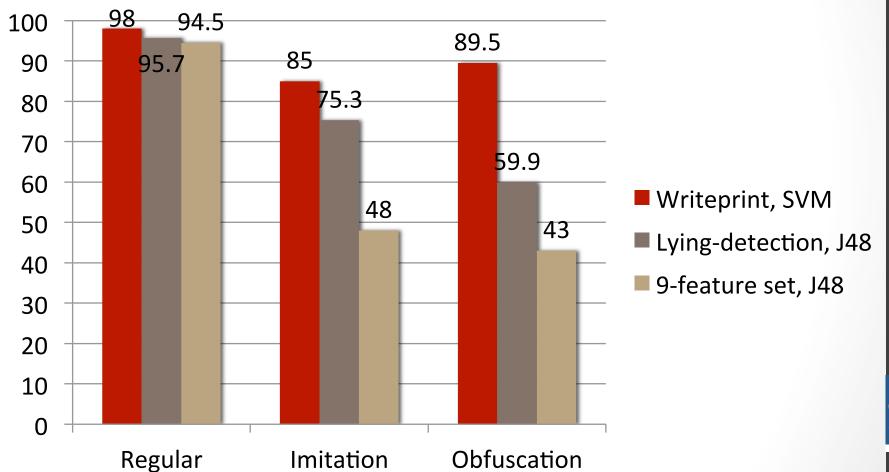


Can we detect stylistic deception?



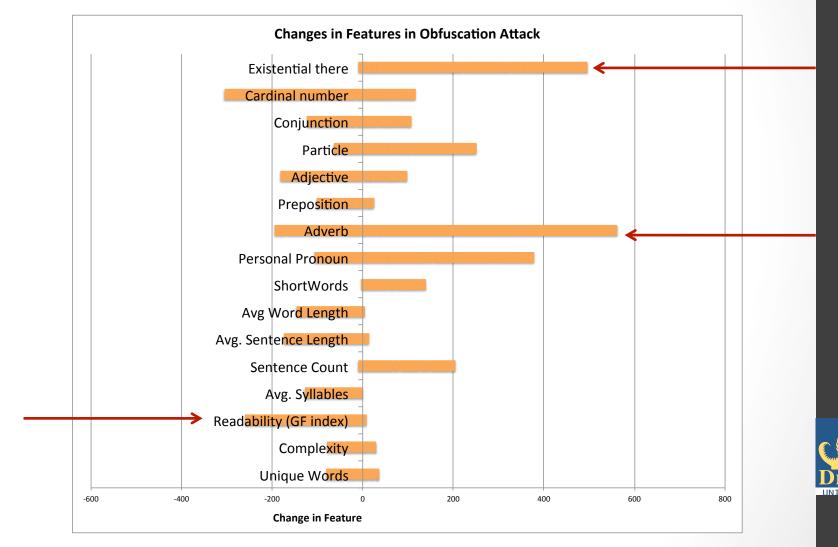


Detecting stylistic deception is possible

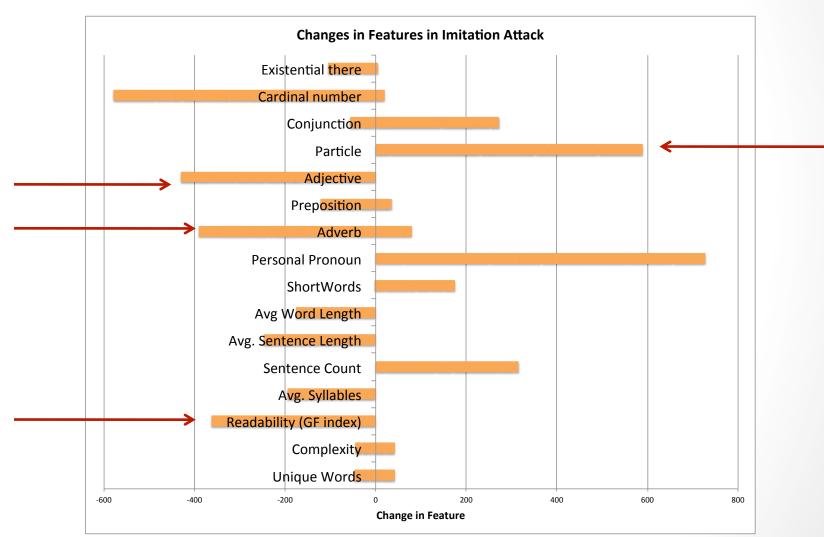




Feature Changes in Obfuscated Passages



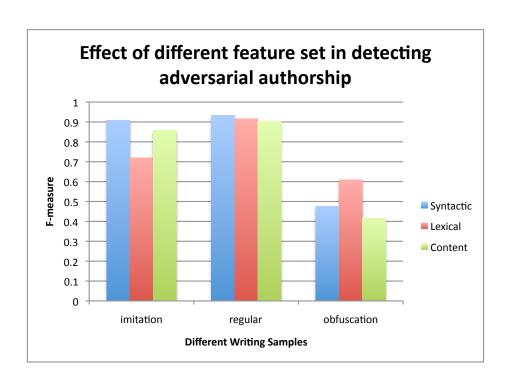
Feature Changes in Imitated Passages





Problem with the dataset: Topic Similarity

- All the deceptive documents were of same topic.
- Non-content-specific features have same effect as contentspecific features.





Hemingway-Faulkner Imitation Corpus

- Articles from the International Imitation Hemingway Contest (2000-2005)
- Articles from the Faux Faulkner Contest (2001-2005)
- Original excerpts of Ernest Hemingway and William Faulkner



Deception detection is possible even when the topic is not similar

81.2% accurate in detecting imitated documents.



Long term deception

- A Gay Girl In Damascus blog:
 - Original author was a 40-year old American citizen, Thomas MacMaster.
 - Pretended to be a Syrian gay woman, Amina Arraf.
 - The author worked for at least 5 years to create a new style.



Long term deception is hard to detect

- None of the blog posts were found to be deceptive.
- But regular authorship recognition can help.
- We tried to attribute authorship of the blog posts using Thomas (as himself), Thomas (as Amina), Britta (Thomas's wife).
- 54.3% of the blog posts were attributed to Thomas (as himself)





Recap

- Available Now:
 - Brennan-Greenstadt Adversarial Stylometry Corpus (12 Authors)
 - Drexel AMT Adversarial Stylometry Corpus (45 Authors)
 - JStylo Alpha Release
 - Anonymouth Alpha Release
- Future Work:
 - Beta releases of JStylo and Anonymouth
 - Academic publication of new results
 - Continued analysis of deception detection and short message classification
 - Continued research on improving partially automated anonymization



Thanks.

- We want to hear from you.
 - Mike Brennan (<u>mb553@drexel.edu</u>)
 - Rachel Greenstadt (greenie@cs.drexel.edu)
 - Ariel Stolerman, JStylo Lead (<u>ariels@drexel.edu</u>)
 - Andrew McDonald, Anonymouth Lead (<u>ams23@drexel.edu</u>)
 - Sadia Afroz, Deception Detection Lead (<u>sa499@drexel.edu</u>)
 - Aylin Caliskan, Translation & Stylometry (<u>ac993@drexel.edu</u>)
- PSAL: https://psal.cs.drexel.edu
- We are looking for interested grad students and post-docs!



Addendum Slides



Research Questions, Practical Implications.

- Our upcoming research questions have substantial practical implications.
- How do you anonymize a document sufficiently in a reasonable period of time?
 - What is *sufficient*? What is *reasonable*?
- Can Anonymouth be used to successfully imitate other authors?
- Can Anonymouth maintain long-term deception? Can its usage be detected?
- JStylo vs. Anonymouth who wins?
 - Based on JStylo, Anonymouth will have everything it needs to help evade detection by the methods it contains.



Two Tools?

- Aren't we creating a tool that enables surveillance and deanonymization?
 - Anonymouth can't exist without JStylo. But it also shows that you can't necessarily depend on stylometry to assign authorship.
 - JStylo allows for easier use of authorship recognition tools, but is extensible and open-source. Implementing a method in JStylo will enable counter-attacks in Anonymouth.
- JStylo vs. Anonymouth who wins?
 - Based on JStylo, Anonymouth will have everything it needs to help evade detection by the methods it contains.
 - Note that nothing prevents others from plugging in proprietary stylometric methods into their version of JStylo.

