

Formosan Cyber Watchdog

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Team members









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Client & Problem of the client

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Craigslist

Classified AD

jobs, apartments, garage sales, used cars, personal ads etc.

Local focus

Reducing long-distance shipping cost Community connection

■ User-friendly System

So easy, your child can do it

craigslist (4)







208 million

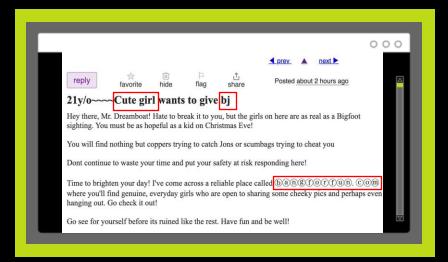
visits per month – as of August 2023





Dark side

Robbery
Sexual Assault
Scammer
Child Trafficking and Slavery









Inappropriate content?



filter

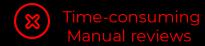
By Keywords

Fail to filter the special symbols and abbreviations



flag

post removed when receiving a sufficient number of flags from different users











Business Analysis



- Ensuring a Safe User Experience
- Compliance with Regulations and Policies
- Improving User Engagement
- Boost Advertiser and Partner Relations





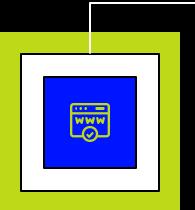






Considerations





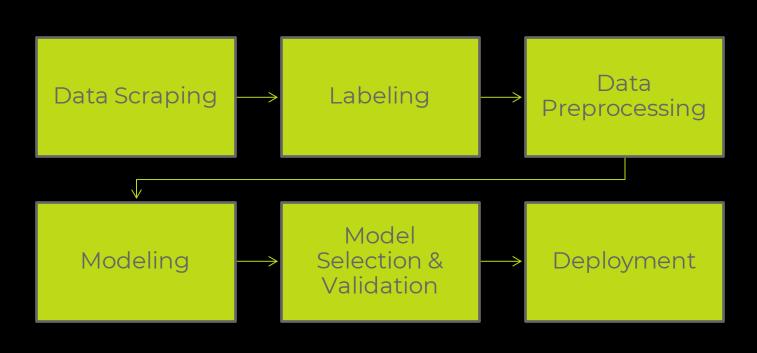
■Scalability & Performance

Find the balance between restricting and website volume





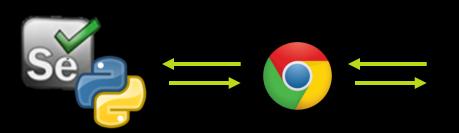
Data Analysis Procedure





Data Scraping

- Utilizing Selenium Library to automate a web browser, execute JavaScript, and collect data
- 11,000 Posts on Craigslist's Community Section



community				
activities artists childcare classes events general groups local news	lost+found missed connections musicians pets politics rants & raves rideshare volunteers			





- Label prediction: supervised learning
- Applied LLM(Large language models) to label posts
- Manually tried different models, Google: PaLM 2 Chat 32k, OpenAI: GPT-3.5, Mistral 7B, Zephyr 7B, etc.
 - Decision: Google PaLm2 Chat 32K
- Sent prompts through API and labeled posts based on the responses

You

You are a language expert.

Please help me to check if the following text contains sexually implicit or explicit content: {This probably sounds crazy but you just delivered 3 heavy items to me in Overland. (12 pack energy drinks) I was in the garage. I had all 3 in my arms and you took one so you could take a pic of it on the porch. We shared a smile together and I felt a connection. You drove a black car but I didn't get a good look at it. Female, wore a hat, I think you had freckles. Black jeans/pants. If you see this, please contact me. I really hope to hear from you. And/or you can stop by the house anytime, ask for me, leave a note, etc. If anyone reads this that might know who I'm talking about, delivery driver, female, freckles (I think), wore a hat, that drives a black car please let her know I'm interested in talking more with her.}.

Please be concise and only answer 'Yes' or 'No'. Do not elaborate."

ChatGPT

No.







Data Prepossessing

- Tokenizing, lemmatizing, and removal of punctuations, stopwords, and numbers.
- Word-embedding:TF-IDF, 1&2-gram, minimum 20 times
- 3,720 words(in 1- or 2-gram) as independent variables
- Data partition: 70% training set & 30% test set







Modeling

- Y = f(X1, X2, X3...), two-class classification
 Appropriate vs. Inappropriate
- Logistic regression
- Naïve Bayes and SVM
- Gradient boosting
- Neural network

	Variable	Coefficient
1265	girl	2.676162
1190	fun	2.593670
400	called	1.975589
1069	find	1.955267
1767	let	1.928546
533	christmas	-1.998285
2993	show contact	-2.044082
2587	puppy	-2.328106
209	band	-2.407132
2715	rehoming	-2.509852

	Feature	Importance
400	called	0.129758
2427	place	0.097166
2658	real	0.044260
1190	fun	0.036258
3571	wasting	0.035527
2992	show	0.027591
2900	see	0.026709
1486	home	0.024740
502	chick	0.023192
1359	guy	0.020294
2715	rehoming	0.019537
3642	woman	0.018684
1027	fee	0.018607
1265	girl	0.018606
2993	show contact	0.016047
2455	please	0.013823
21	actually	0.011650
209	band	0.010891
180	available	0.010033
3308	think	0.009944

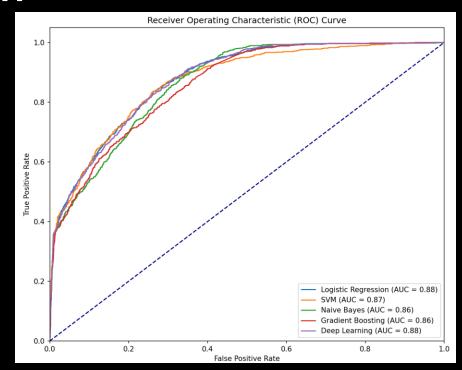




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Model Selection

Model	Criteria: AUC
Logistic Regression	87.8%
Naive Bayes	86.5%
SVM	87.0%
Gradient Boosting	85.9%
Neural Network	87.6%







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Prediction Example:

Demo1	Just GOOGLE the name BIAINGIFIORIFIUN. CIOM and you can view it right in the top While you are ready to stop wasting time, move see for your own. You will be cheerful you did!
Demo2	Don't risk your job, relationship, or freedom replying to these posts! There's still one sight that hasn't been tainted by cheaters or the feds. It's called PUSSYBIZ and it's a haven for everyday girls who are just looking for some company
Demo3	Older Women (Silver Spring) Mid 40's SBM loves the company of women 50+. Extra points if unshaven
Demo4	Pen pal (Akron) Older guy looking for someone to chat with. In my mid 60s and would prefer someone around my age. And preferably a female. Easy to talk to, non judgmental. Someone?



Prediction Demo

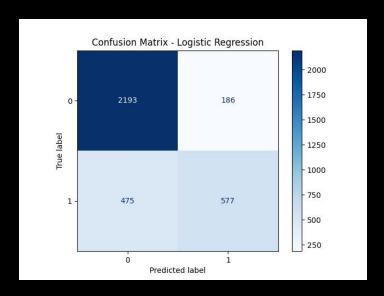
```
Logistic Regression Predictions for new samples: ['1' '1' '0' '0'] SVM Predictions for new samples: ['1' '1' '0' '0'] Naive Bayes Predictions for new samples: ['1' '1' '0' '0'] Gradient Boosting Predictions for new samples: ['1' '1' '0' '0'] Deep Learning Predictions for new samples: ['1' '1' '0' '0']
```



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Confusion Matrix(Classification):

- Type 1 error (Not harmful but predicted as harmful): 186/(2193+186)≈7.82%
- Type 2 error (harmful but predicted as not harmful): 475/(577+475)≈45.15%







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Threshold(0~1):

- Tradeoff between user experience(Type1 error) and restriction level
- By creating a UX, adjust threshold to decrease Type1 error(avoid too much intervention).

	Politics	Activities (children-related)
Threshold (Delete)	0.9	0.8
Threshold (Alert)	0.7	0.6





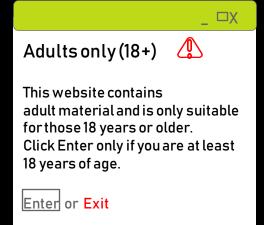


Application



Pop-up Alert Window

Protect children from inappropriate content









Conclusion





Accuracy

New Models: improve the accuracies of the predictions.

