bit.ly/ona23-beyond-chatgpt

AI tools from before and beyond ChatGPT

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Let’s (mostly) not talk about ChatGPT.

How ChatGPT turned generati

arstechnica.com/ai/2023/08/how-chatgpt-turned-generative-ai-into-an-anything-tool/

Update

ars TECHNICA

BIZ & IT TECH SCIENCE POLICY CARS GAMING & CULTUR

SIGN IN

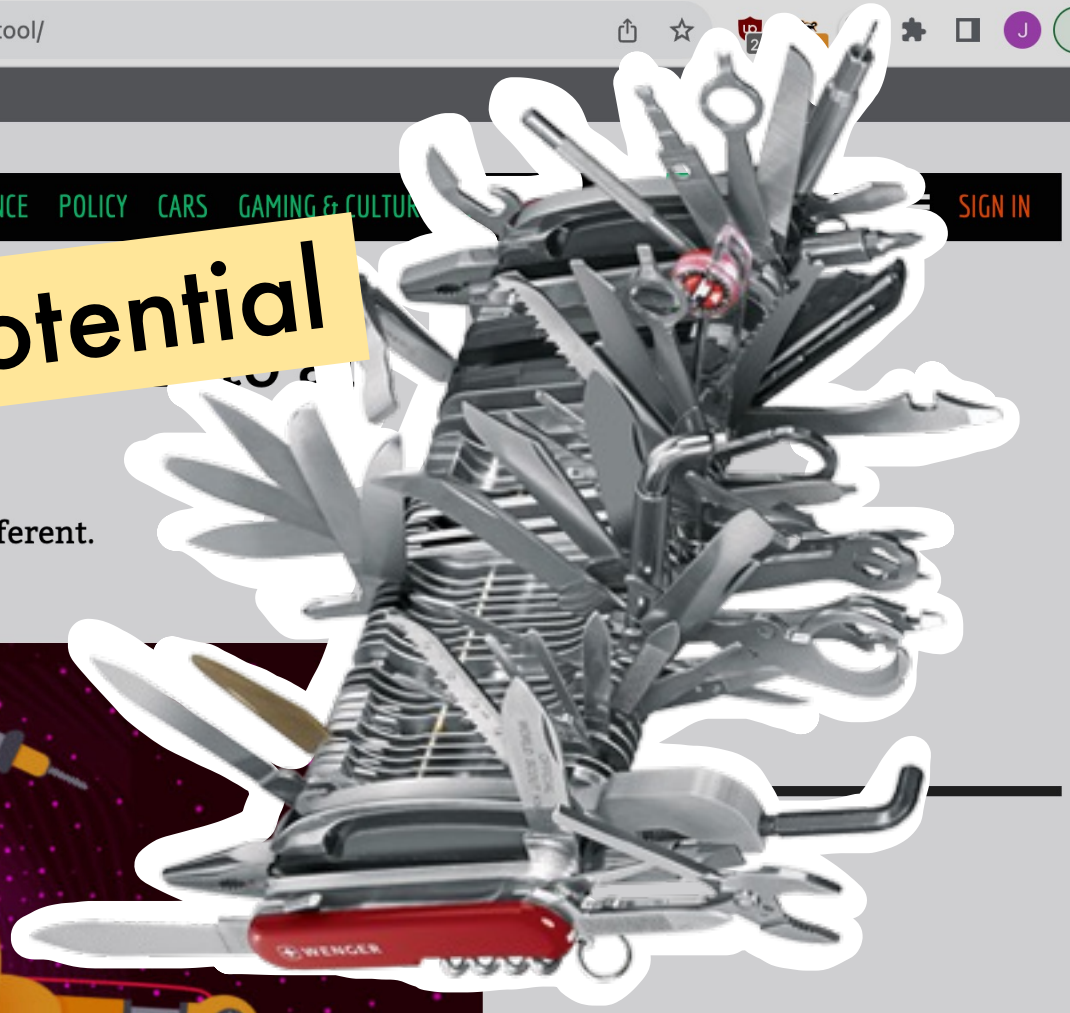
DO ANYTHING —

How Cha

until recently, AI models were specialized tools. Modern LLMs are different.

HAOMIAO HUANG - 8/23/2023, 7:30 AM

tty Images



poor discoverability of potential

https://arstechnica.com/ai/2023/08/how-chatgpt-turned-generative-ai-into-an-anything-tool/

This isn't about *specific tools*

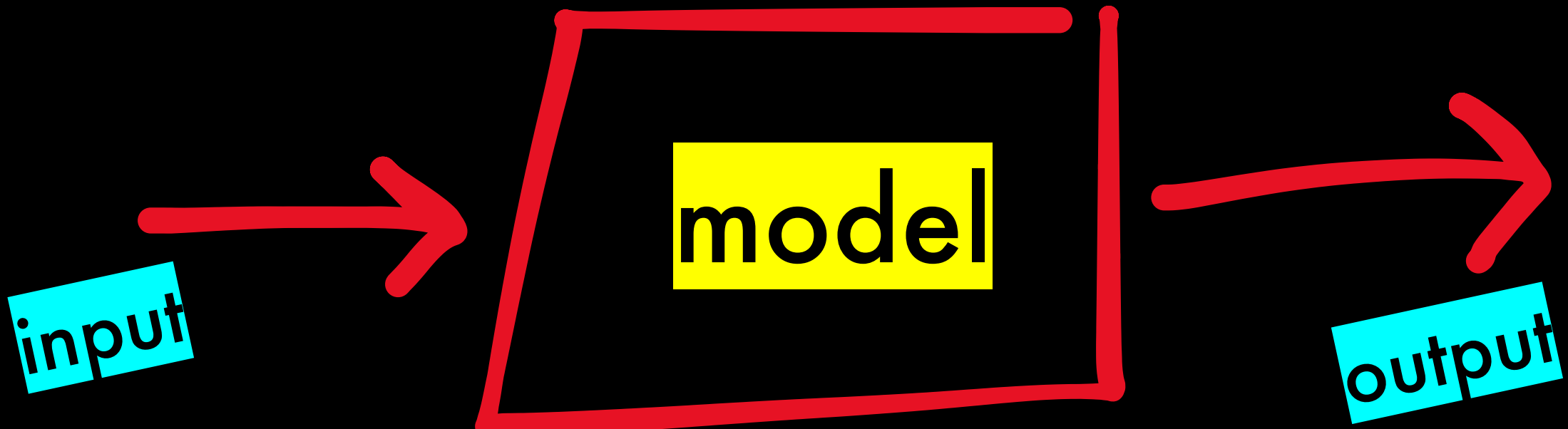
Freedom: It's easy and useful to experiment on your own

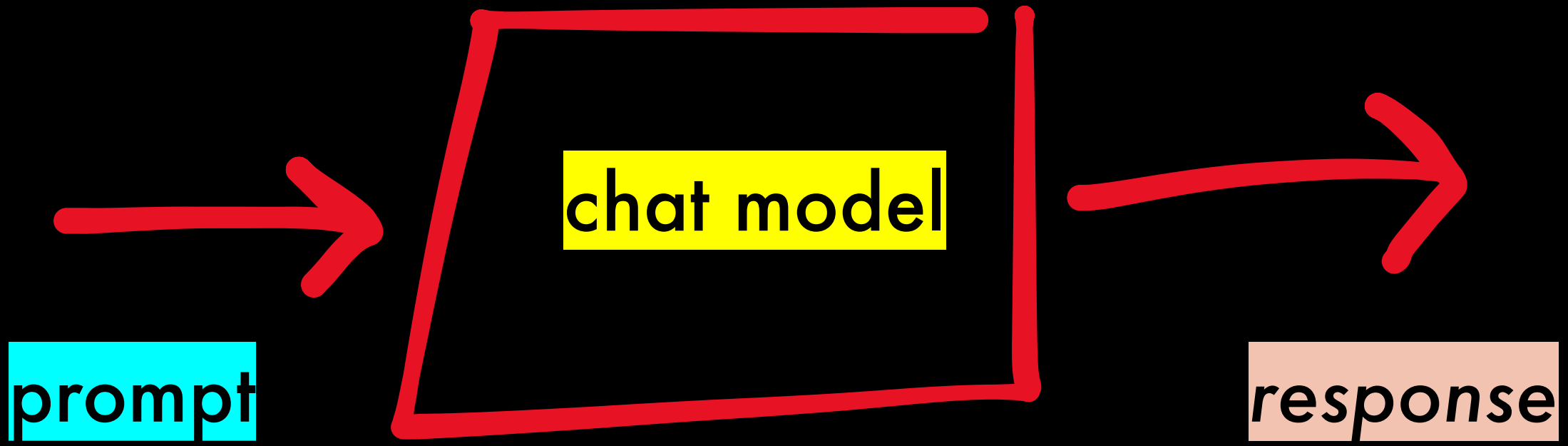
Flexibility: Vendor lock in makes me sad

Lots of investigation, little biz

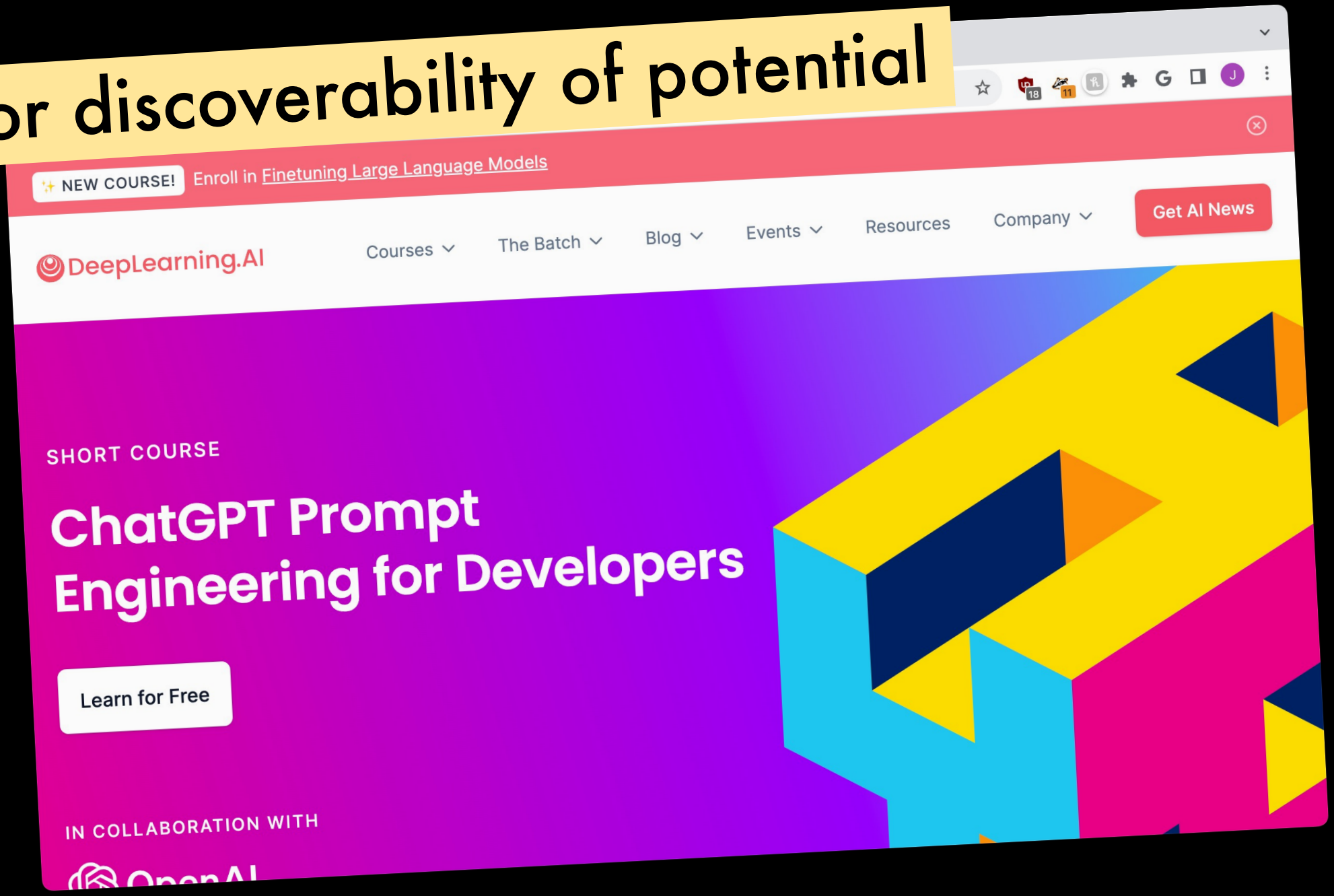
models

~~"tools"~~

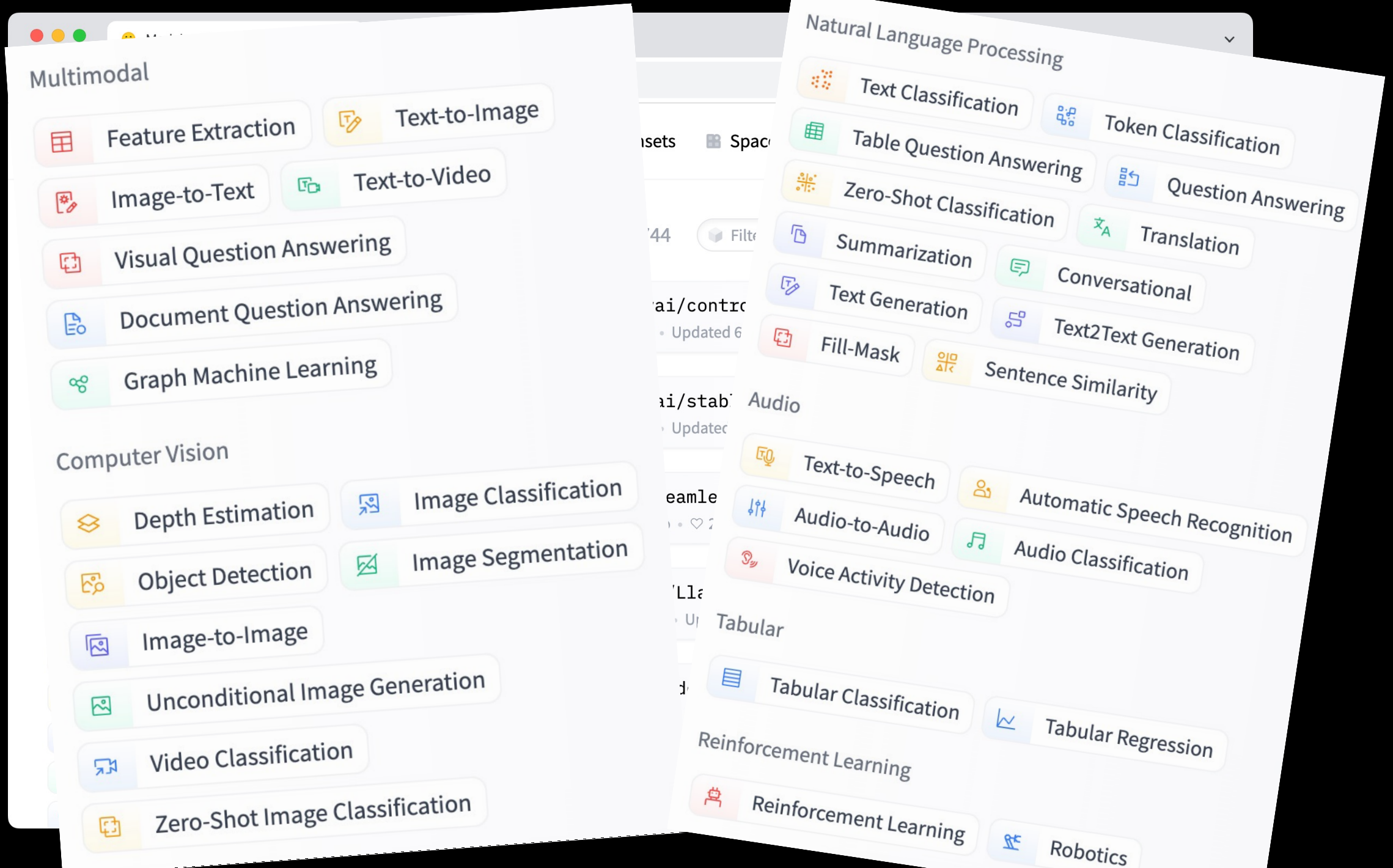

input
model
output


chat model
prompt
response

poor discoverability of potential




text
sentiment analysis model
POSITIVE
or
NEGATIVE


Multimodal
Feature Extraction
Text-to-Image
Image-to-Text
Text-to-Video
Visual Question Answering
Document Question Answering
Graph Machine Learning
Computer Vision
Depth Estimation
Image Classification
Object Detection
Image Segmentation
Image-to-Image
Unconditional Image Generation
Video Classification
Zero-Shot Image Classification
Natural Language Processing
Text Classification
Token Classification
Table Question Answering
Question Answering
Zero-Shot Classification
Translation
Summarization
Conversational
Text Generation
Text2Text Generation
Fill-Mask
Sentence Similarity
Audio
Text-to-Speech
Automatic Speech Recognition
Audio-to-Audio
Audio Classification
Voice Activity Detection
Tabular
Tabular Classification
Tabular Regression
Reinforcement Learning
Reinforcement Learning
Robotics


text
sentiment analysis model
POSITIVE
or
NEGATIVE


classification
model
text
category

Apple's App Store is full of con

washingtonpost.com/technology/2019/11/22/apple-says-its-app-store-is-safe-trusted-place-we-...

jonathan.soma

The Washington Post
Democracy Dies in Darkness

TECH Help Desk Artificial Intelligence Internet Culture Space Tech Policy

TECHNOLOGY

Apple says its App Store ‘is safe and trusted place.’ We found of unwanted sexual beh apps, some targeting m

The prevalence of unwanted sexual content raises questions protective cocoon to its customers as its platform grows

By Reed Albergotti and Al Johri
November 22, 2019 at 7:00 a.m. EST

all this app supports is hate and negativity” “This should be taken off the app “The ones I thought

wp

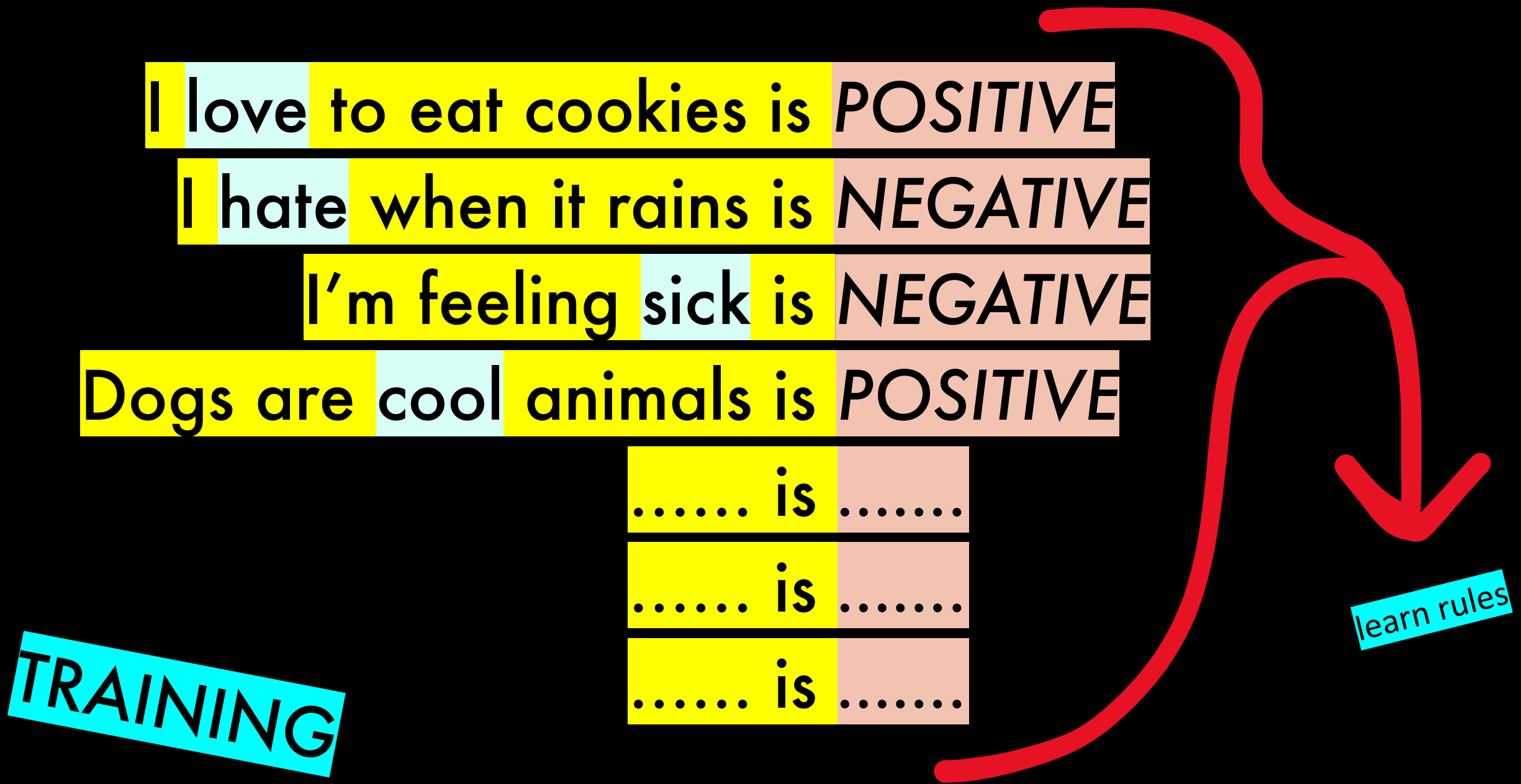
1:05 / 2:53

Chat apps in Apple's App Store are rife with predatory and sexually explicit behavior. The W
used machine learning to see just how bad it is. (Video: Reed Albergotti/T

	polarity	text
2	0	@kconsidder You never tweet
3	0	Sick today coding from the couch.
4	1	@ChargerJenn Thx for answering so quick,I was afraid I was gonna crash twitter
5	1	Wii fit says I've lost 10 pounds since last time
6	0	@MrKinetik Not a thing!!! I don't really have a life.....
7	1	and its pretty cool I never thought I would ever get to this point...
8	0	finallly! twitter wouldnt let me in! x
9	1	Just took a shower, i feel so refreshed now
10	0	Love the show! You and Jasmine got me watching! Every mtv web exclusive has me a great dancer!
11	1	I have a story to tell you... its my birthday tomorrow!! yes thats right tomorrow is th ....Lauren is a TOOL face... HAPPY EOFYS!!
12	1	That came out wrong. I heart @danielradcliffe just as much.
13	0	Just got done with the Fam celebratin Jun bdays, it's sad to see my grandma getting so forgetful, couldn't remember who had bdays n June
14	0	dissapointed about missing the VIP chance on NIN in Chicago.
15	1	@i_am_girlfriday aw, i'm sure you were absolutely cute with those bangs
16	1	@Jefner Thanks for the welcome I hope it does too, it should be fine but I gotta speak on it too
17	1	@iamdiddy I'm here for diddy!
18	1	Fact #1: Between 70-75% of bottled water comes from the ground; the same source where consumers receive tap water. Not PURE SWISS -
19	0	Going to bed now. not gonna wait longer, if the phone rings, then great, if not i'll just be sleeping.
20	1	ohhhhhh im so ready for a relaxing weekend
21	1	http://www.flickr.com/people/bugekgek/ Thy new gallery.
22	0	is up to no good. has to leave for camping in 10 hours ! .. and cannot sleep
23	0	@SaintBastard dude. I'm a Caps fan at heart. 0 cups. Many a heartbreaking loss. My #2 is the Bruins.. so.. yeah.
24	1	happy birthday mother
25	1	@Edwin808 there u go u see RE and Bodysnatching go hand in hand - beautiful spot to be doing either or both actually
26	1	@mrsjackbauer7 haha omg it works now... finally!
27	0	About to spend an evening of unadulterated gaming, the wife's in bed early and the controllers are charged! If only I had a gaming buddy
28	0	I wanted to go eat with you guys but my mom made me go to Coolidge elementary to clean my old teacher's classroom
29	0	@dirty69_4ever my wife and I need 2 get to bed earlier in the eve so she'll be more open to that kind of thing. We NEED our 5hrs of sleep
30	1	. @MaryRSnyder Thank you! I figured you could use the chocolate biscotti after all that shredding!
31	0	@chanicedev0nne wtf niecey r u serious.. dam yo wats happening man? that shit aint right..I dnt get it
32	0	xbox live is down for the day Oh well, gives me a chance to play Chronicles of Riddick or start a new character in Fable 2.

YOU LABEL!

IT LEARNS!


I love to eat cookies is POSITIVE
I hate when it rains is NEGATIVE
I'm feeling sick is NEGATIVE
Dogs are cool animals is POSITIVE
..... is .....
..... is .....
..... is .....
TRAINING
learn rules

Language is weird

fishing	running
fishes	runs
fished	ran

Vectorize our text

We'll be using a stemmed TF-IDF vectorizer to both combine similar words - "pic" and "pics" - as well as have uncommon words carry a little more weight.

```
#!pip install pystemmer
#!pip install sklearn
```

```
%%time

from sklearn.feature_extraction.text import Tfi
import Stemmer

# English stemmer from pyStemmer
stemmer = Stemmer.Stemmer('en')

analyzer = TfidfVectorizer().build_analyzer()

# Override CountVectorizer
class StemmedTfidfVectorizer(TfidfVectorizer):
    def build_analyzer(self):
        analyzer = super(TfidfVectorizer,
        return lambda doc: stemmer.stemWo

# Create a new StemmedCountVectorizer
vectorizer = StemmedTfidfVectorizer()
matrix = vectorizer.fit_transform(known.

# Build a dataframe of words, purely out
words_df = pd.DataFrame(matrix.toarray()
words_df.head(5)
```

```
vectorizer = StemmedTfidfVectorizer(max_features=500, max_df=0.30)
matrix = vectorizer.fit_transform(known.Review)

# Build a dataframe of words, purely out of curiosity
words_df = pd.DataFrame(matrix.toarray(), columns=vectorizer.get_feature_names
words_df.head(5)
```

it was awful

```
%%time

from sklearn.svm import LinearSVC

X = matrix
y = known.sexual

clf = LinearSVC(class_weight='balanced')
clf.fit(X, y)
```

```
CPU times: user 4.92 ms, sys: 2.89 ms, total: 7.81 ms
Wall time: 6.1 ms
```

None of that matters now!

Instead, we (might) fine-tune pre-trained models

Yes, those are both vocabulary words you will be tested on later.

Train a custom model

hire a dumb baby

"Fine-tune" a pre-trained model

hire a grouchy teen

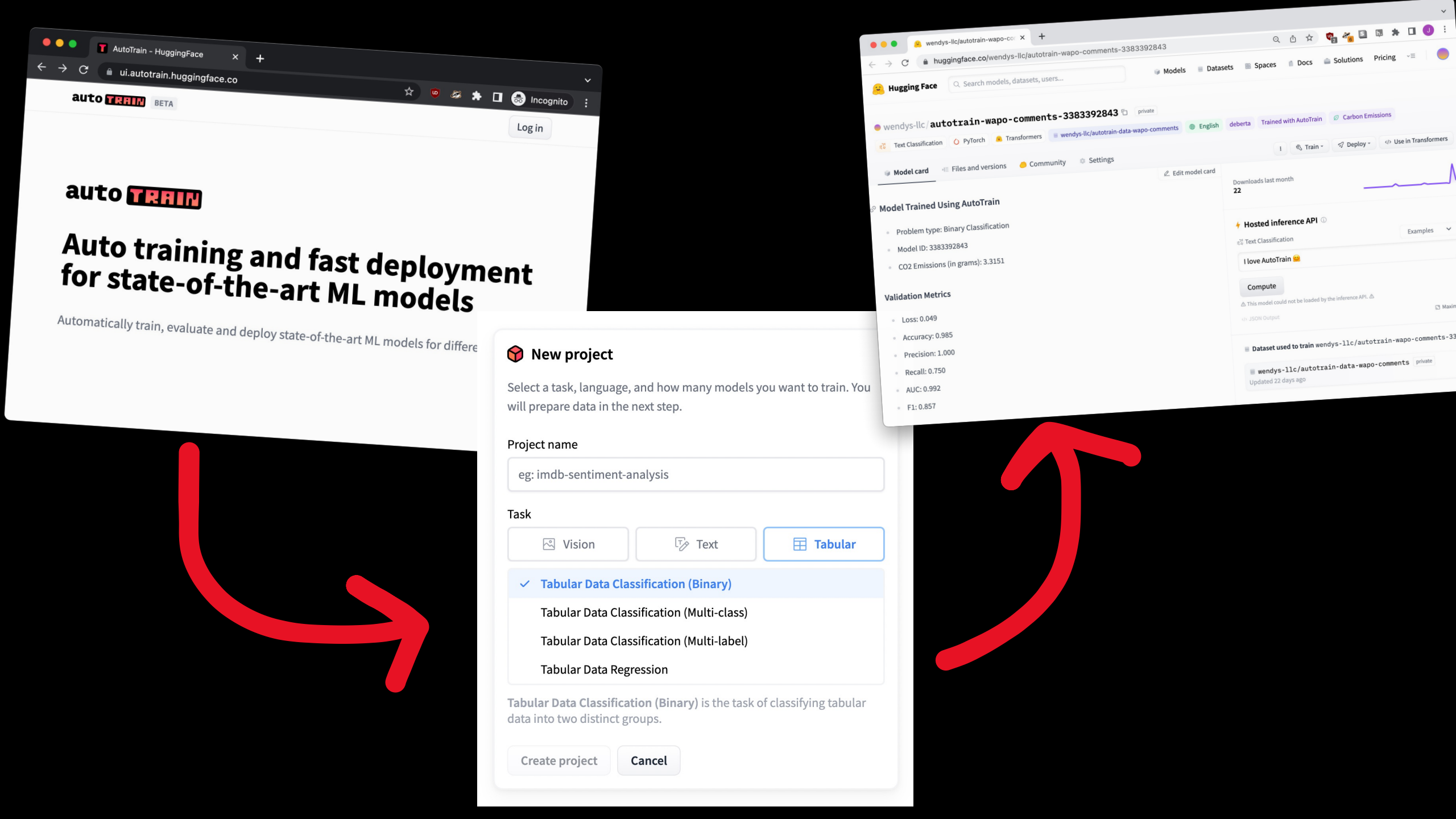
A WHOLE LOT OF TEXT

TRAINING
(to learn language)

I love to eat cookies is *POSITIVE*
I hate when it rains is *NEGATIVE*
I'm feeling sick is *NEGATIVE*
Dogs are cool animals is *POSITIVE*
..... is .....
..... is .....
..... is .....

FINE TUNING
(to learn rules)

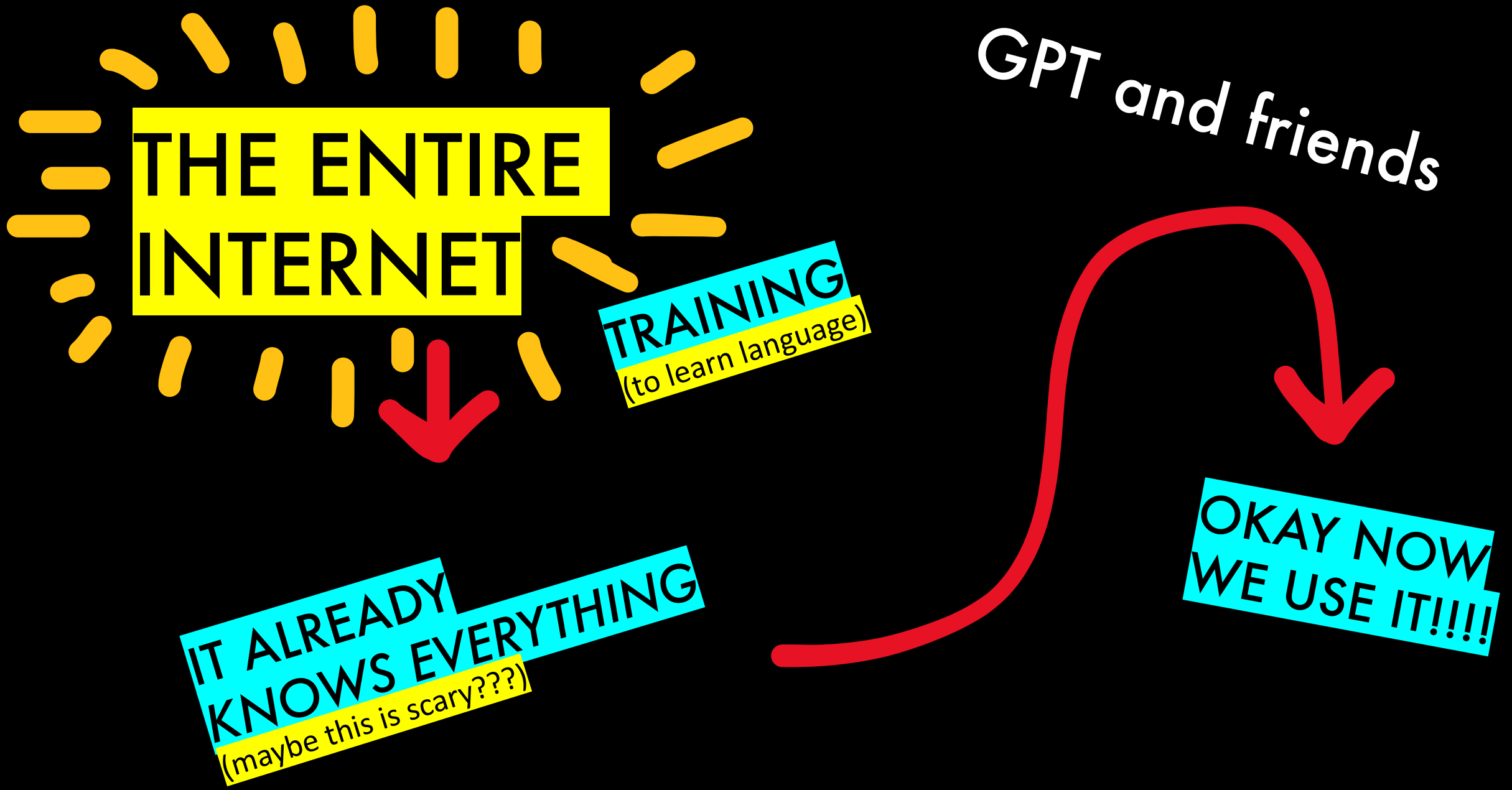
OKAY NOW WE USE IT!!!!


AutoTrain - HuggingFace
ui.autotrain.huggingface.co
auto TRAIN BETA
Incognito
Log in
auto TRAIN
Auto training and fast deployment for state-of-the-art ML models
Automatically train, evaluate and deploy state-of-the-art ML models for differe
New project
Select a task, language, and how many models you want to train. You will prepare data in the next step.
Project name
eg: imdb-sentiment-analysis
Task
Vision
Text
Tabular
Tabular Data Classification (Binary)
Tabular Data Classification (Multi-class)
Tabular Data Classification (Multi-label)
Tabular Data Regression
Tabular Data Classification (Binary) is the task of classifying tabular data into two distinct groups.
Create project
Cancel
huggingface.co/wendys-llc/autotrain-wapo-comments-3383392843
Hugging Face
Search models, datasets, users...
Models
Datasets
Spaces
Docs
Solutions
Pricing
wendys-llc / autotrain-wapo-comments-3383392843
private
English
deberta
Trained with AutoTrain
Carbon Emissions
Text Classification
PyTorch
Transformers
wendys-llc/autotrain-data-wapo-comments
Train
Deploy
Use in Transformers
Model card
Files and versions
Community
Settings
Edit model card
Model Trained Using AutoTrain
Problem type: Binary Classification
Model ID: 3383392843
CO2 Emissions (in grams): 3.3151
Validation Metrics
Loss: 0.049
Accuracy: 0.985
Precision: 1.000
Recall: 0.750
AUC: 0.992
F1: 0.857
Downloads last month
22
Hosted inference API
Text Classification
Examples
I love AutoTrain
Compute
This model could not be loaded by the inference API.
JSON Output
Dataset used to train wendys-llc/autotrain-wapo-comments-33
wendys-llc/autotrain-data-wapo-comments
private
Updated 22 days ago


```
from transformers import pipeline

creepy_pipeline = pipeline(model="wendys-llc/creepy-wapo")
data = [
    "I love the app, talking to people is fun",
    "Be careful talking to men, they all want nudes :("
]

creepy_pipeline(data)
```

```
[{'label': '0.0', 'score': 0.998849630355835},
 {'label': '1.0', 'score': 0.8436576128005981}]
```


classification
model
text
category


GPT and friends
THE ENTIRE INTERNET
TRAINING
(to learn language)
IT ALREADY KNOWS EVERYTHING
(maybe this is scary???)
OKAY NOW WE USE IT!!!

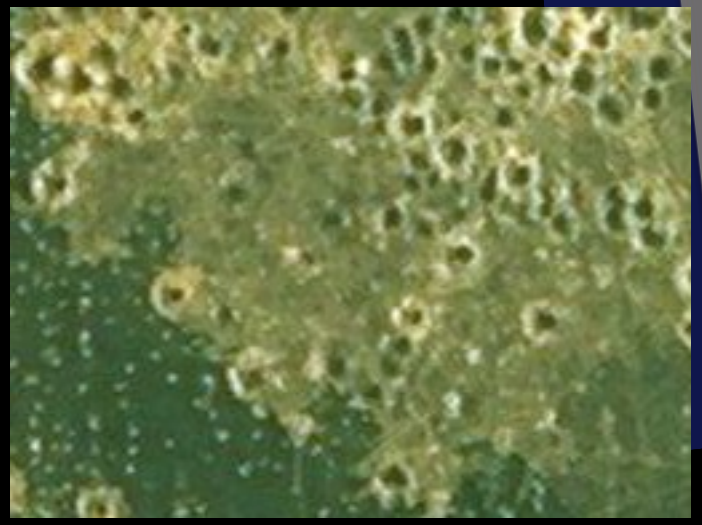
Pure magic: zero-shot classification

```
prompt = """
Categorize the following text as being about ENVIRONMENT, GUN CONTROL,
or IMMIGRATION. Respond with only the category.

Text: A Bill to Regulate the Sulfur Emissions of Coal-Fired Energy
Plants in the State of New York.
"""

response = llm.predict(prompt)
print(response)
```

```
ENVIRONMENT
```

Also "few-shot" if it's nuanced

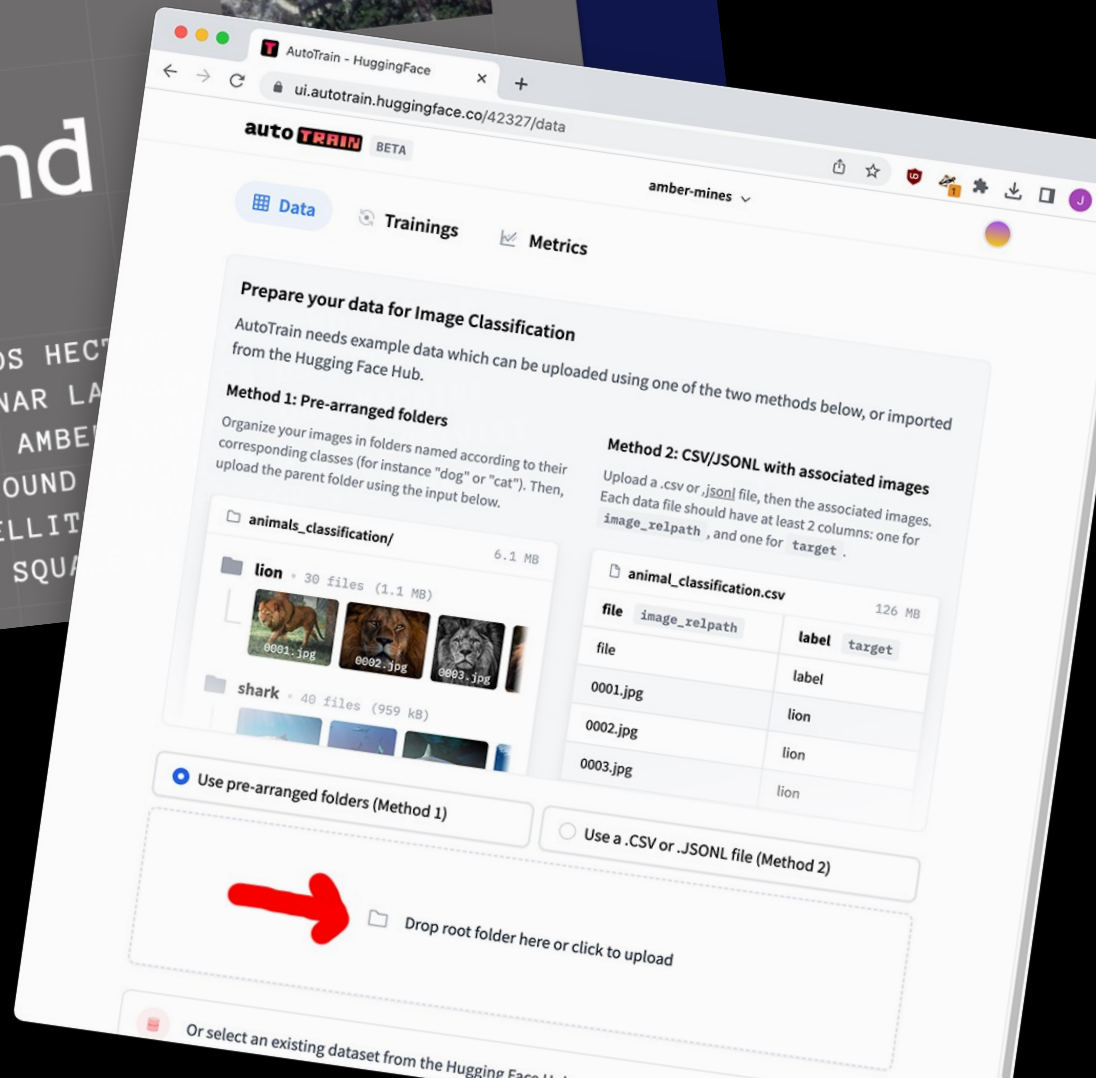
Classification works with images, too



Computer vision

Leprosy of the land

THOUSANDS HEC
INTO LUNAR LA
ILLEGAL AMBE
MODEL FOUND
ON SATELLIT
70,000 SQU



https://texty.org.ua/d/2018/amber_eng/

There's also word classification

But no one in history has ever called it that

Named Entity Recognition (NER)

```
import spacy
from spacy import displacy

text = """
Yevgeny Prigozhin
```

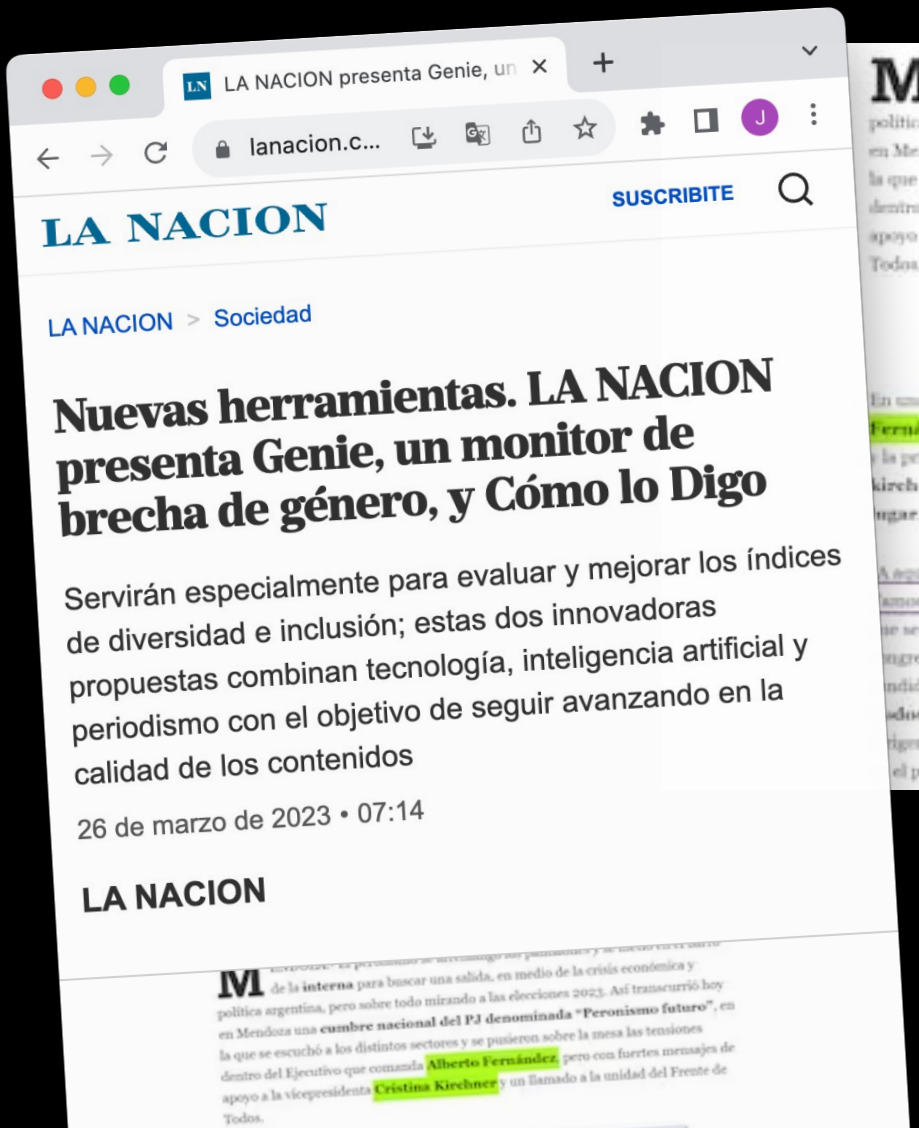
Yevgeny Prigozhin PERSON , Renegade Mercenary PRODUCT Chief Who Rattled the Kremlin ORG

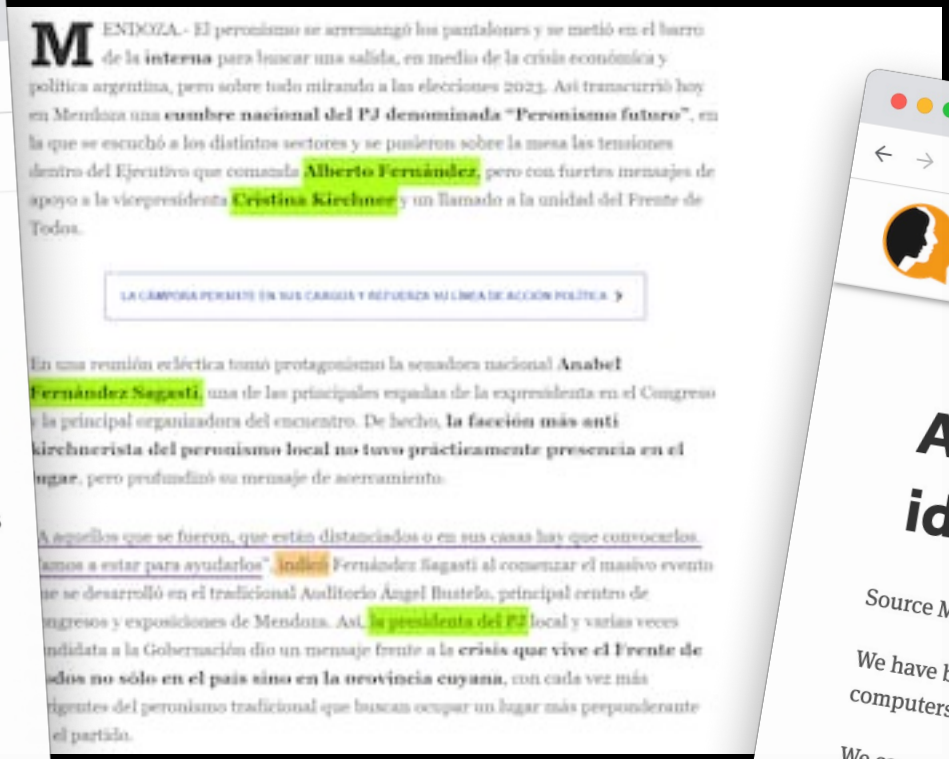
A tycoon and Putin PERSON ally, his paramilitary force fought by Russia GPE 's side

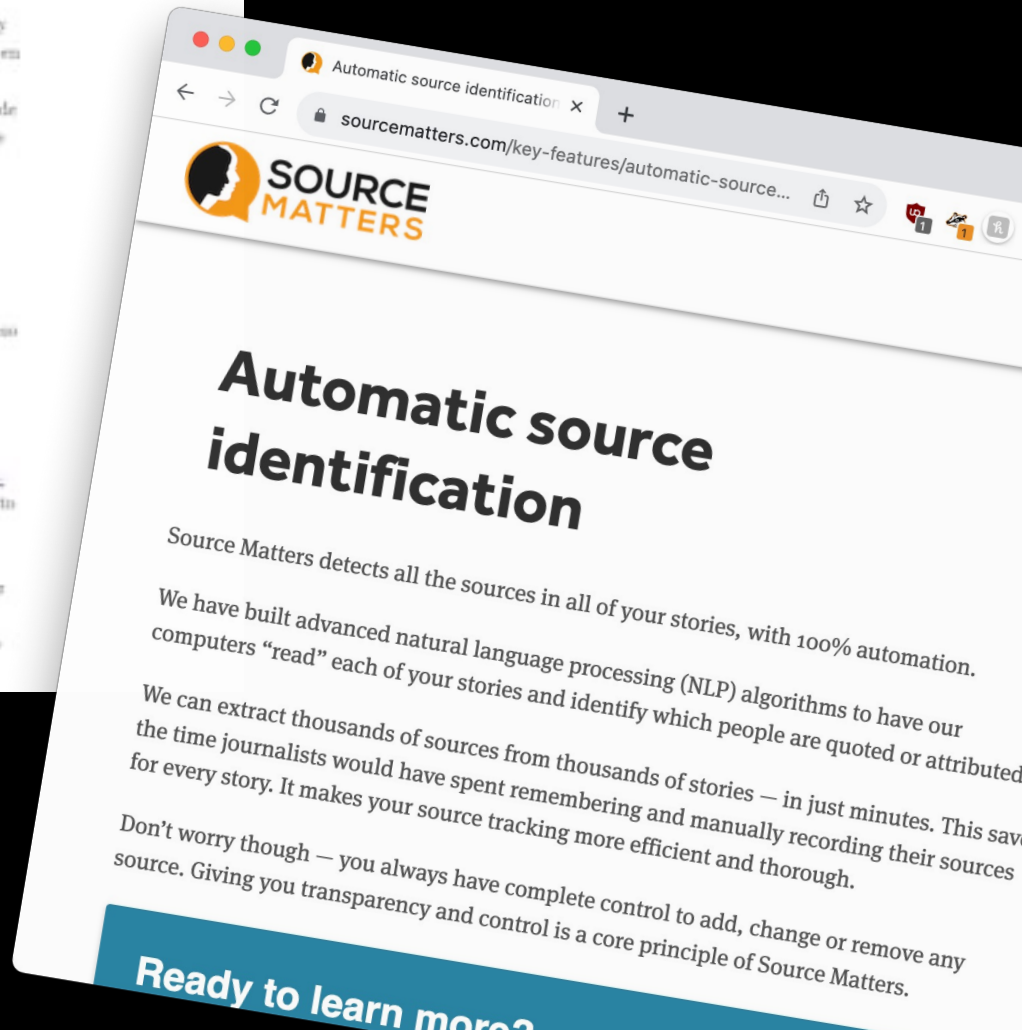
even as he castigated its military leaders. He is believed dead at 62 CARDINAL in a plane crash.

```
displacy.render(doc, style='ent', jupyter=True)
```

Use case: source diversity







JS

List the named entities in the text below. Use a comma to separate the entity and the type of entity. Valid entity types are PERSON, EVENT, ORGANIZATION, and LOCATION.

Text: Hi [illegible] NA23 in

Jonathan Soma - PERSON,
ONA23 - EVENT,
Philadelphia - LOCATION

The next big secret is
text embeddings

cat

dog

wolf

tiger

cat

cat

tiger

cat

tiger

dog

cat

tiger

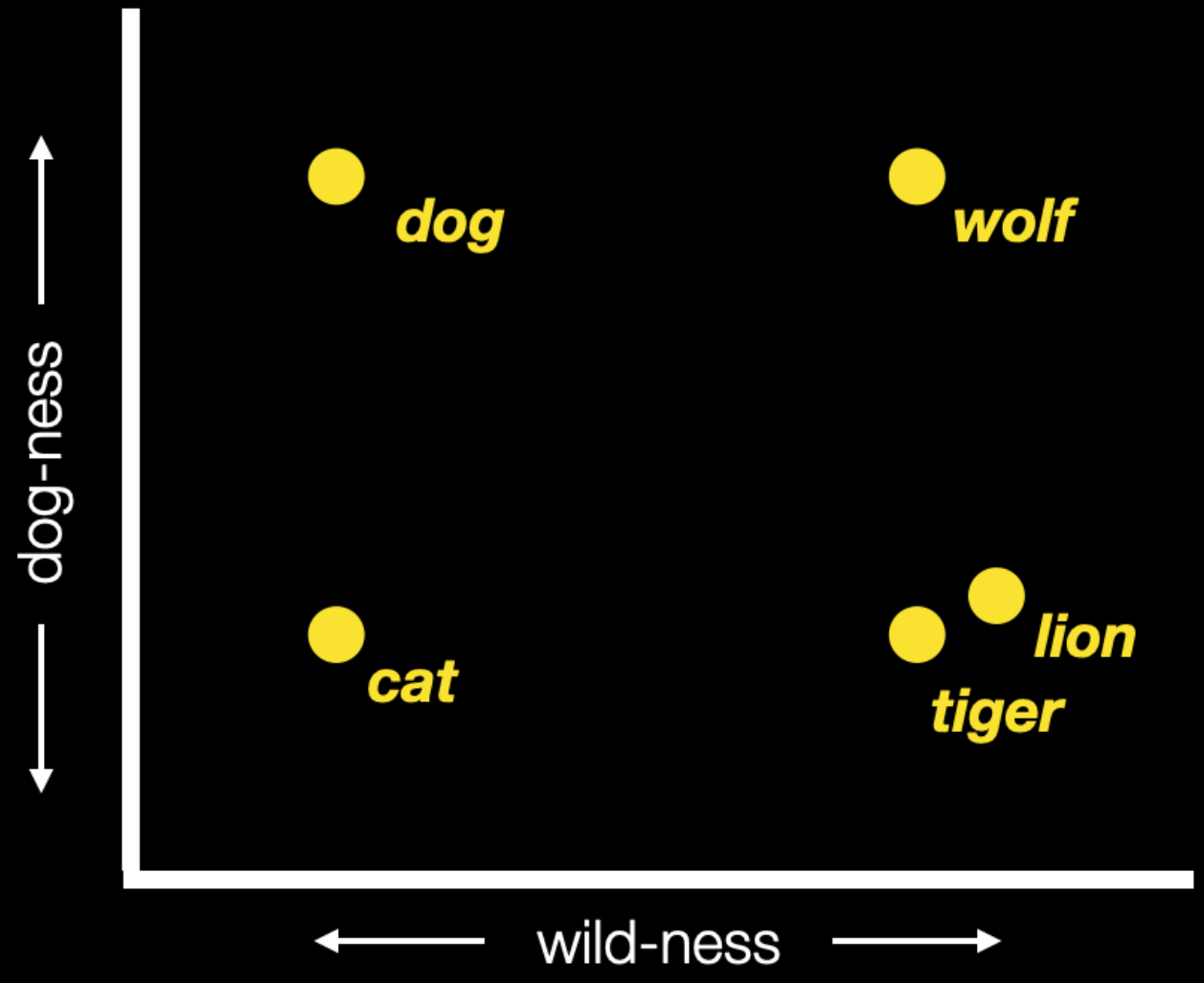
dog

wolf

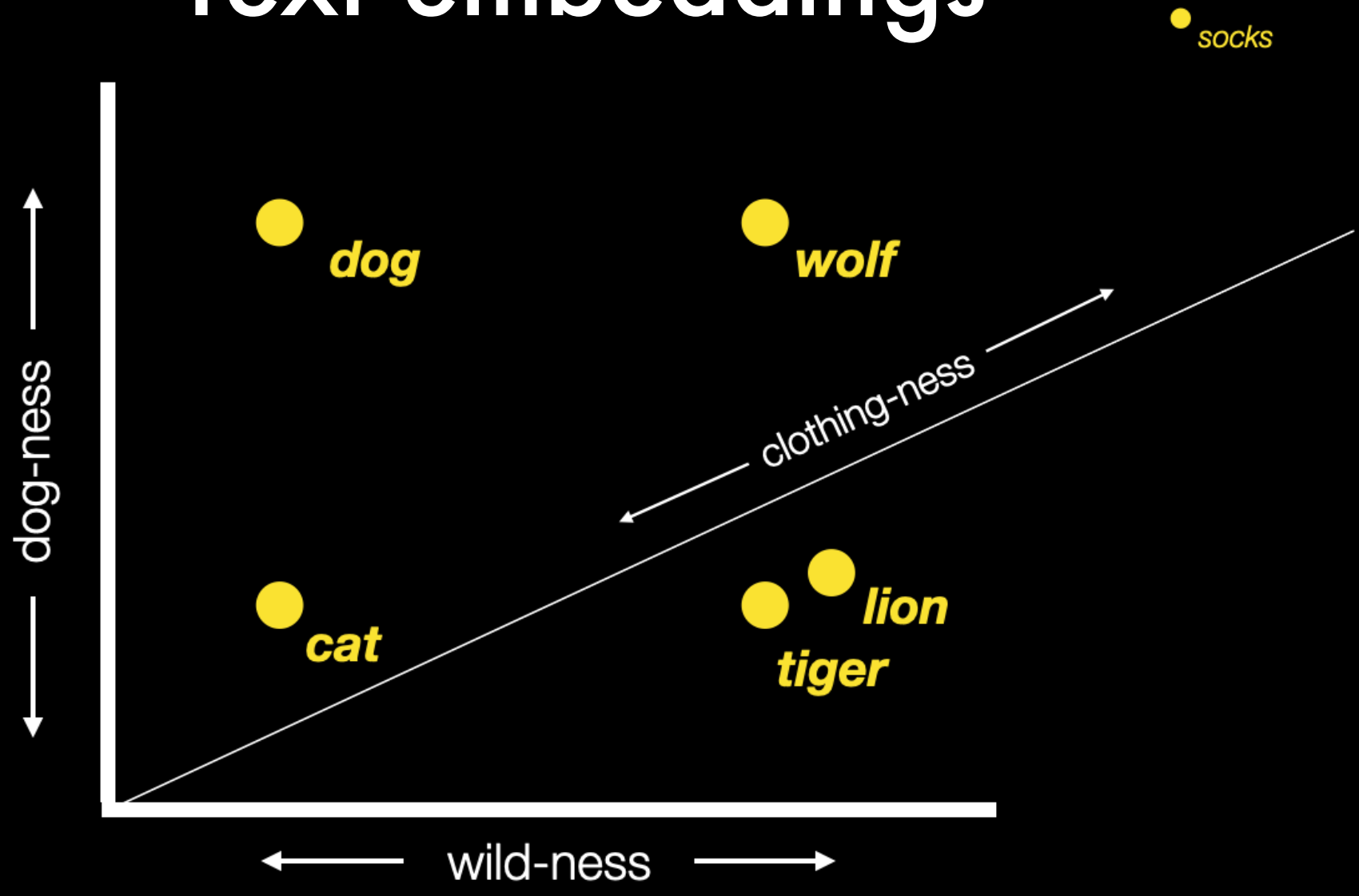
cat

tiger

Text embeddings



Text embeddings



	dog	wild	clothing	fuzzy	code	.....
cat	1	3	1	4	5	.....
dog	10	2	0	6	0	.....
wolf	9	9	1	2	0	.....
lion	2	0	1	6	2	.....
tiger	1	0	2	3	3	.....
socks	0	2	9	7	1	.....

```
array([-0.15067  , -0.024468 , -0.23368  , -0.23378  , -0.18382  ,   -2.8174   , -0.38319  , -0.022373 ,  0.56376  ,  0.40131  ,
        0.32711  , -0.22084  , -0.28777  ,  0.12759  ,  1.1656   ,   -0.42131  , -0.11311  , -0.17317  ,  0.1411   , -0.13194  ,
       -0.64163  , -0.098455 , -0.62397  ,  0.010431 , -0.25653  ,    0.18494  ,  0.097692 , -0.097341 , -0.23987  ,  0.16631  ,
        0.31799  ,  0.037779 ,  1.1904   , -0.17714  , -0.2595   ,   -0.28556  ,  0.0038654,  0.53292  , -0.32367  , -0.38744  ,
       -0.31461  ,  0.038825 , -0.15713  , -0.13484  ,  0.36936  ,    0.27011  , -0.34181  , -0.27702  , -0.67279  , -0.10771  ,
       -0.30562  , -0.40619  , -0.38965  ,  0.3686   ,  0.013963 ,   -0.062189 , -0.24783  , -0.070884 , -0.20898  ,  0.062404 ,
       -0.6895   ,  0.004066 , -0.1367   ,  0.32564  ,  0.24688  ,    0.022372 ,  0.13408  ,  0.1305   , -0.19546  , -0.46849  ,
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        0.20526  ,  0.098624 , -0.23877  ,  0.045338 ,  0.43941  ,    0.46565  ,  0.088294 , -0.48496  ,  0.69263  , -0.068977 ,
        0.030385 , -0.013821 , -0.093273 , -0.18178  ,  0.19438  ,   -0.53709  ,  0.20802  , -0.42987  , -0.11921  ,  0.1174   ,
       -0.3782   ,  0.70144  ,  0.16236                                                     0.1236   ,  0.3607   , -0.19608  ,
       -0.13613  , -0.11425  , -0.31598                                                     0.5061   ,  0.14455  , -0.024368 ,
        0.5419   , -0.42413  , -0.599                                                       0.58634  , -0.054461 ,  0.0076487,
        0.14964  ,  0.29287  , -0.31281                                                     0.23096  , -0.29296  , -0.24325  ,
       -0.4408   ,  1.2174   ,  0.51236                                                     0.7089   ,  0.17402  , -0.0037509,
        0.092514 ,  0.71396  , -0.021051 , -0.33704  , -0.20275  ,   -0.46304  ,  0.11806  , -0.16457  , -0.38609  ,  0.14524  ,
       -0.36181  ,  0.22055  , -0.25665  ,  0.28425  , -0.16968  ,    0.098122 , -0.12352  , -0.1047   ,  0.39047  , -0.3063   ,
        0.058029 ,  0.61182  ,  0.31576  , -0.079185 ,  0.35538  ,   -0.65375  , -0.0044248, -0.033876 ,  0.037114 , -0.27472  ,
       -0.51236  ,  0.4235   , -0.30033  , -0.22376  ,  0.15223  ,    0.0053147,  0.30737  ,  0.12528  , -0.19527  , -0.16461  ,
       -0.048292 ,  0.23532  ,  0.46507  , -0.67579  , -0.32905  ,    0.087518 , -0.051107 , -0.16323  ,  0.521    ,  0.10822  ,
        0.08446  , -0.22123  , -0.045333 ,  0.34463  , -0.1455   ,   -0.060379 , -0.71735  , -0.064327 ,  0.37043  , -0.41054  ,
       -0.18047  , -0.17887  ,  0.96879  , -1.0028   , -0.47343  ,   -0.2728   , -0.30217  ,  0.015771 , -0.43056  ,  0.35647  ,
        0.28542  ,  0.56382  , -0.33211  , -0.38275  , -0.2749   ,    0.17188  , -0.54598  , -0.21541  , -0.044889 , -0.10597  ,
       -0.22955  , -0.24265  , -0.37689  ,  0.24822  ,  0.36941  ,   -0.54391  ,  0.53908  ,  0.070938 ,  0.097839 ,  0.097908 ,
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       -2.8174   , -0.38319  , -0.022373 ,  0.56376  ,  0.40131  ,    0.019574 ,  0.24805  ,  0.3144   , -0.29304  ,  0.54235  ,
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        0.18494  ,  0.097692 , -0.097341 , -0.23987  ,  0.16631  ,   -0.33181  , -0.10902  , -0.023324 ,  0.2106   , -0.29633  ,
       -0.28556  ,  0.0038654,  0.53292  , -0.32367  , -0.38744  ,    0.81506  ,  0.038524 ,  0.46004  ,  0.17187  , -0.29804  ],
```

this is “cat”

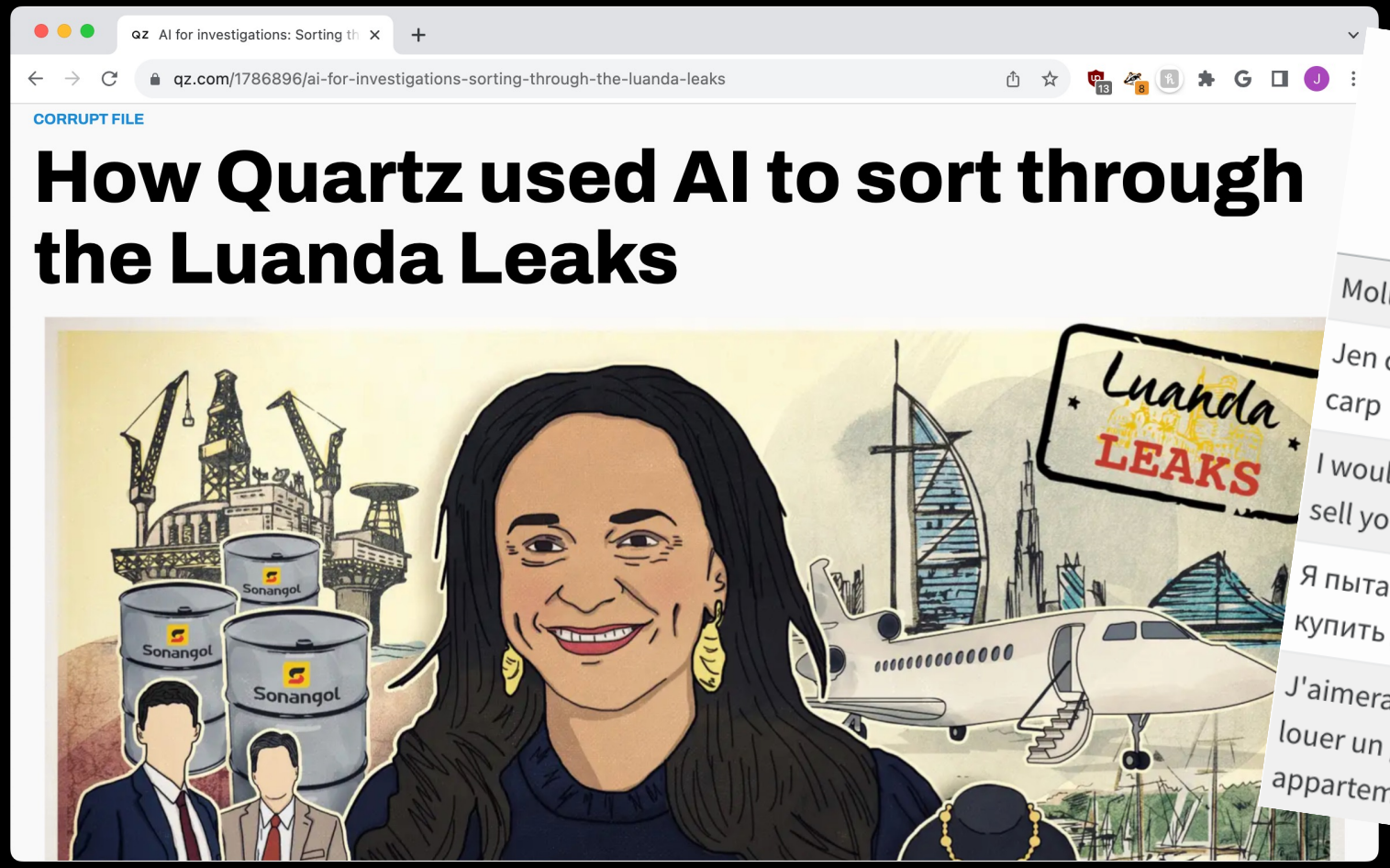
Jen ate a fish

Molly consumed a carp

vibe match

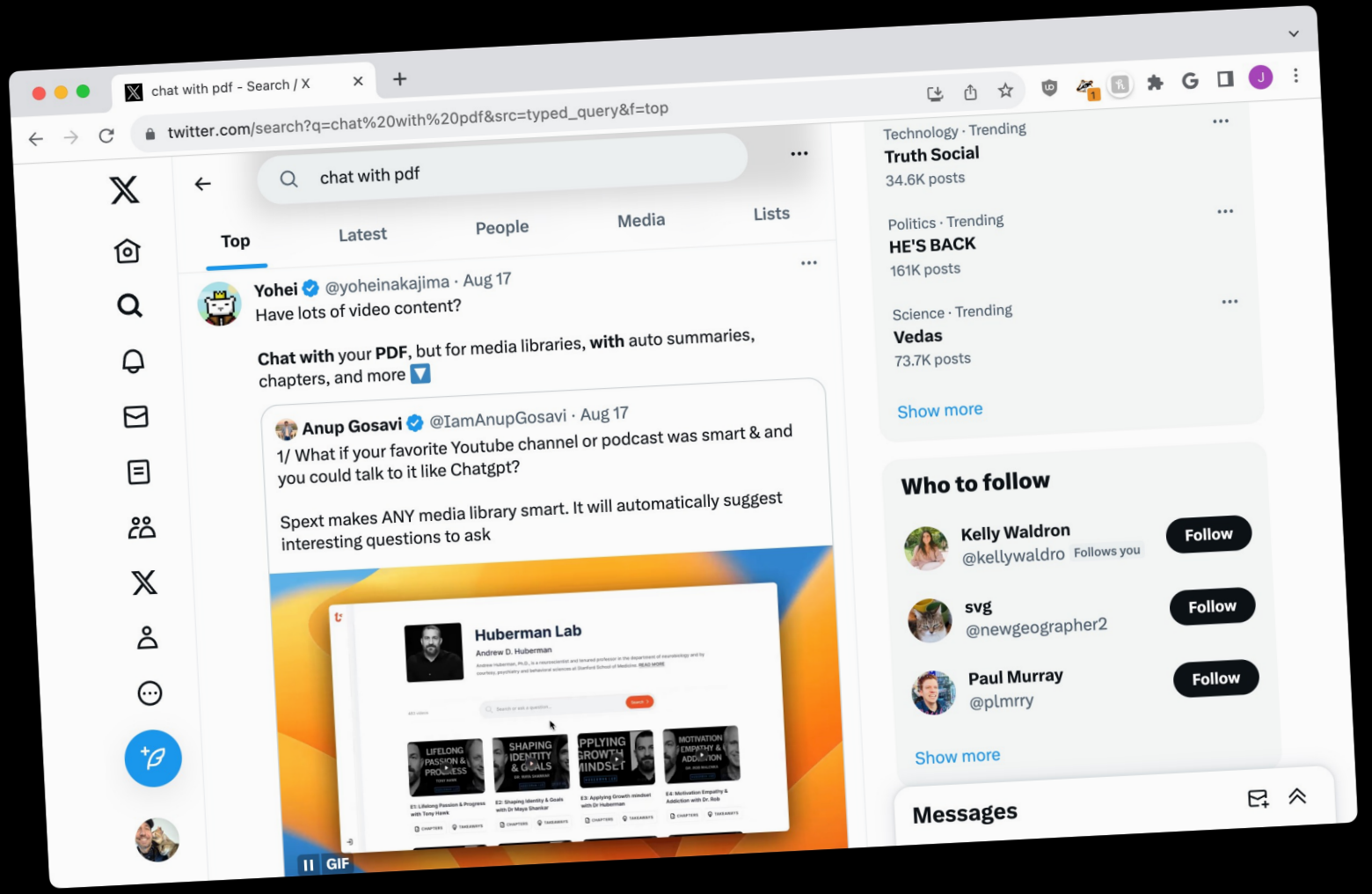
a.k.a. conceptual document similarity

Semantic search



	Molly ate a fish	Jen consumed a carp	I would like to sell you a house	Я пытаюсь купить дачу	J'aimerai vous louer un grand apparteme
Molly ate a fish	1.000000	0.358347	0.058340	0.145439	-0.024103
Jen consumed a carp	0.358347	1.000000	0.059195	0.190241	-0.001941
I would like to sell you a house	0.058340	0.059195	1.000000	0.418692	0.642746
Я пытаюсь купить дачу	0.145439	0.190241	0.418692	1.000000	0.351605
J'aimerais vous louer un grand appartement	-0.024103	-0.001941	0.642746	0.351605	1.000000

“chat with your documents”



Using the texts below, answer the question "what did Zsuzska steal from the devil?"

Hiába tagadta szegény Zsuzska, nem használt semmit, elindult hát nagyszomorúan. Épen éjfél volt, mikor az ördög házához ért, aludt az ördögis, a felesége is. Zsuzska csendesen belopódzott, ellopta a tenger-ütőpálczát, avval bekiáltott az ablakon.– Hej ördög, viszem ám már a tenger-ütő pálczádat is. – Hej kutya Zsuzska, megöletted három szép lyányomat, elloptad atenger-lépő czipőmet, most viszed a tenger-ütő pálczámat, de majdmeglakolsz te ezért.Utána is szaladt, de megint csak a tengerparton tudott közel jutnihozzá, ott meg Zsuzska megütötte a tengert a tenger-ütő pálczával,kétfelé vált előtte, utána meg összecsapódott, megint nem foghatta megaz ördög. Zsuzska ment egyenesen a királyhoz.

De Zsuzska nem adta;,,Tán bolond vagyok, hogy visszaadjam, mikor kivülvagyok már vele az udvaron?!" Az ördög kergette egy darabig, de sehogyse tudta utolérni, utoljára is visszafordult, Zsuzska pedig mentegyenesen a király elibe, od'adta neki az arany fej káposztát. – No felséges király elhoztam már ezt is! A két nénjét Zsuzskának, majd hogy meg nem ütötte a guta, mikormegtudták, hogy Zsuzskának most se' lett semmi baja, másnap megintbementek a királyhoz. – Jaj felséges király van még annak az ördögnek egy arany kis gyermekeis arany bölcsőben, Zsuzska azt beszéli fűnek-fának, hogy ő azt is eltudná lopni. Megint behivatta a király Zsuzskát. – Fiam Zsuzska, azt hallottam, hogy van annak az ördögnek egy arany kisgyermeke is, arany bölcsőben, te azt is el tudod lopni, azt beszélted,azért ha az éjjel el nem lopod, halálnak halálával halsz meg.

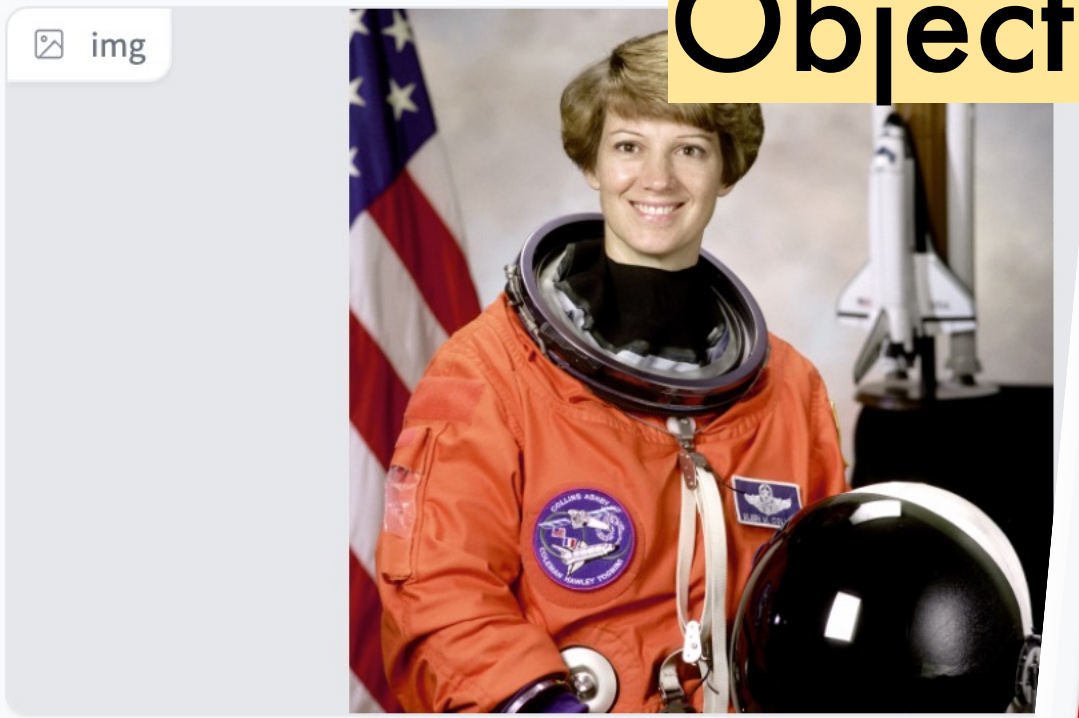
– No felséges király, elhoztam már a tengerütő pálczát is. A király még jobban megszerette Zsuzskát, hogy olyan életre való, de anénjei még jobban irigykedtek rá, csakhamar megint avval árulták be,hogy van annak az ördögnek egy arany fej káposztája is, Zsuzska azt isel tudná lopni, azt mondta. A király megint ráparancsolt Zsuzskára erősparancsolattal, hogy ha a káposztát el nem lopja, halálnak halálával halmeg. Elindult hát szegény Zsuzska megint, el is ért szerencsésen épen éjfélreaz ördög kertjibe, levágta az arany fej káposztát, avval bekiáltott azablakon. – Hej ördög, viszem ám már az arany fej káposztádat is. – Hej kutya Zsuzska, megöletted három szép lyányomat, elloptad atenger-lépő czipőmet, elloptad a tenger-ütő pálczámat, most viszed azarany fej káposztámat, csak ezt az egyet add vissza, soha szemedre sevetem.

Text models can also do automatic translation

I think this is generally an awful idea, though. It feels disrespectful to both your readership and labor as a concept.

Image models: things vs stuff

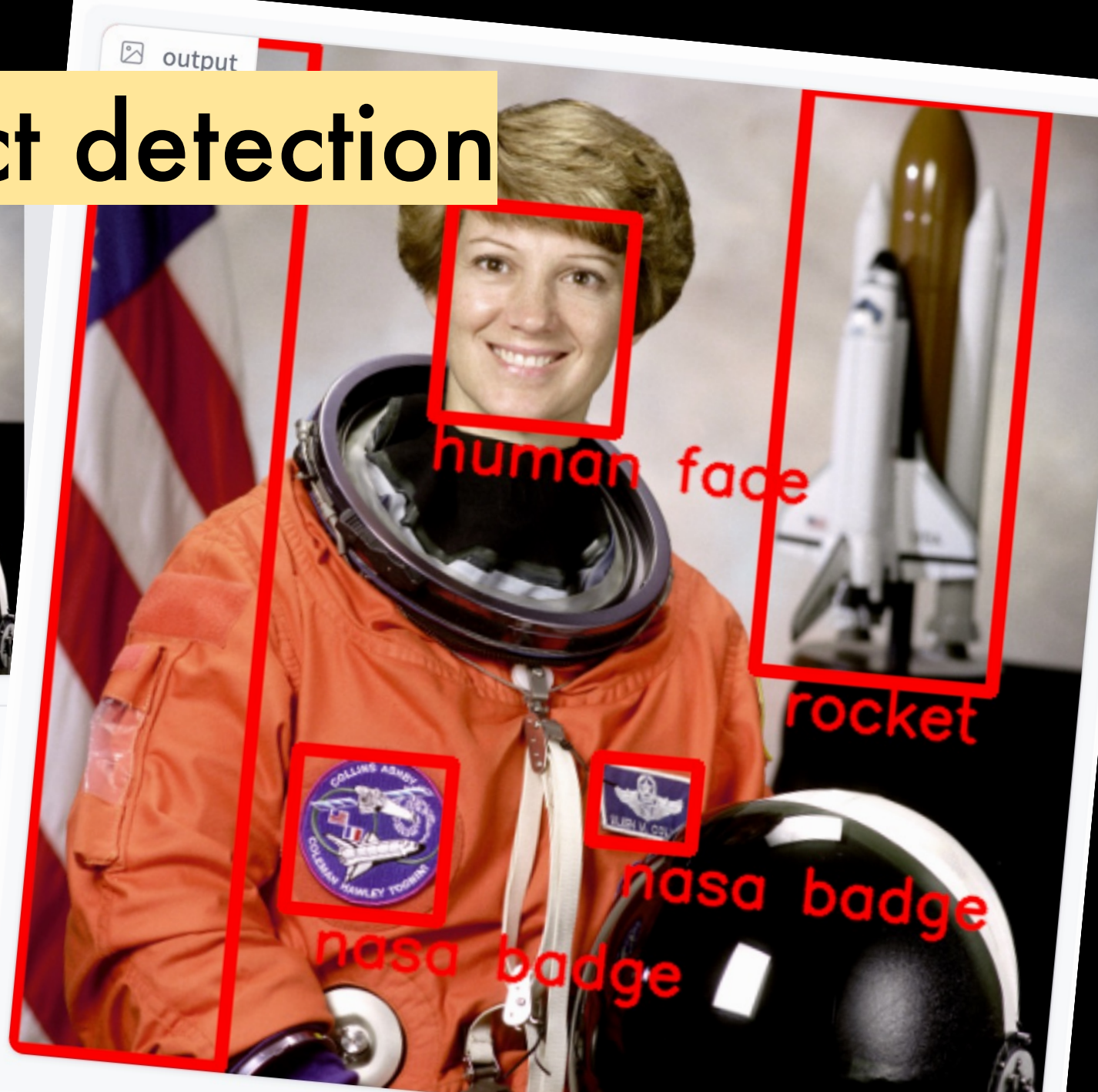
Object detection



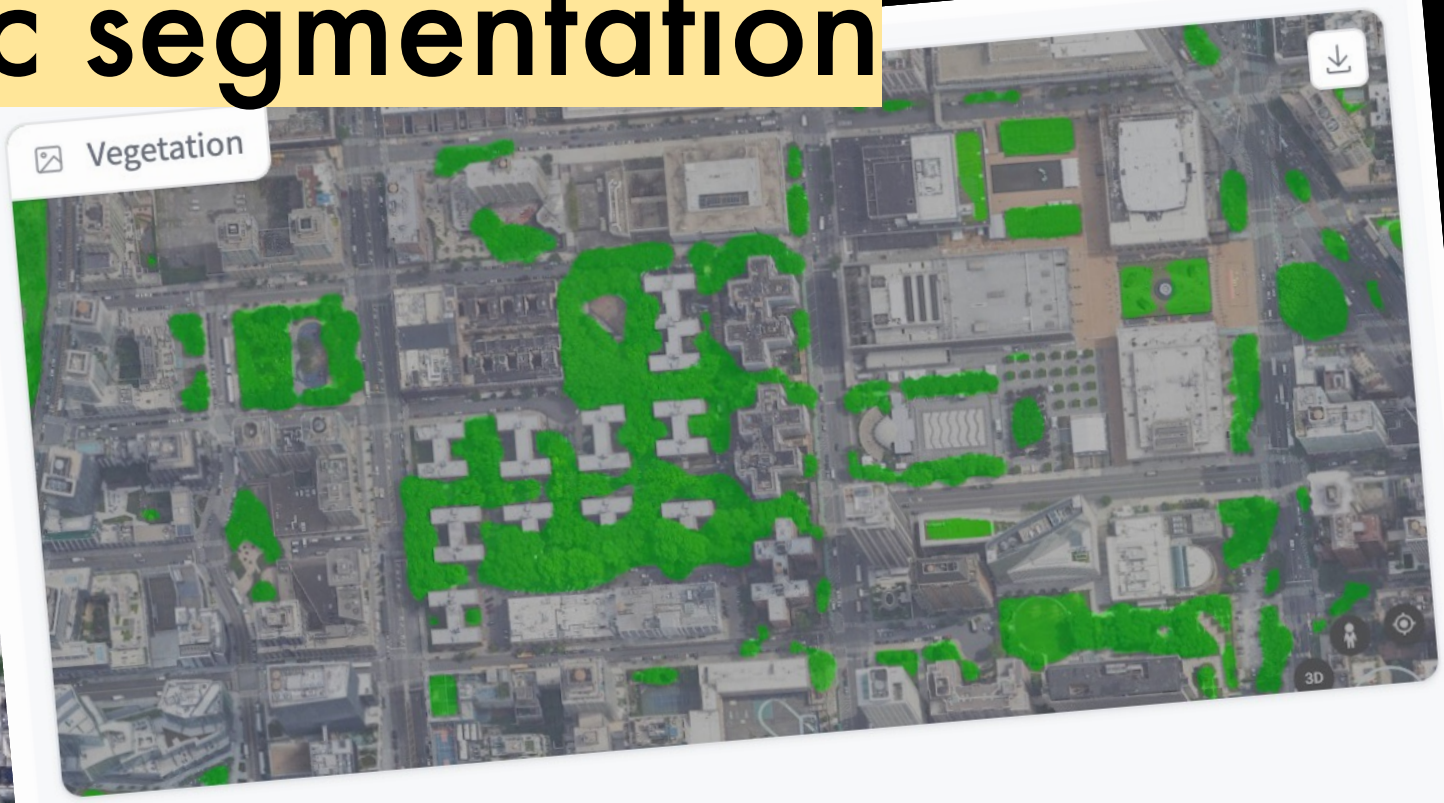
text_queries

human face, rocket, star-spangled banner, nasa badge

score_threshold



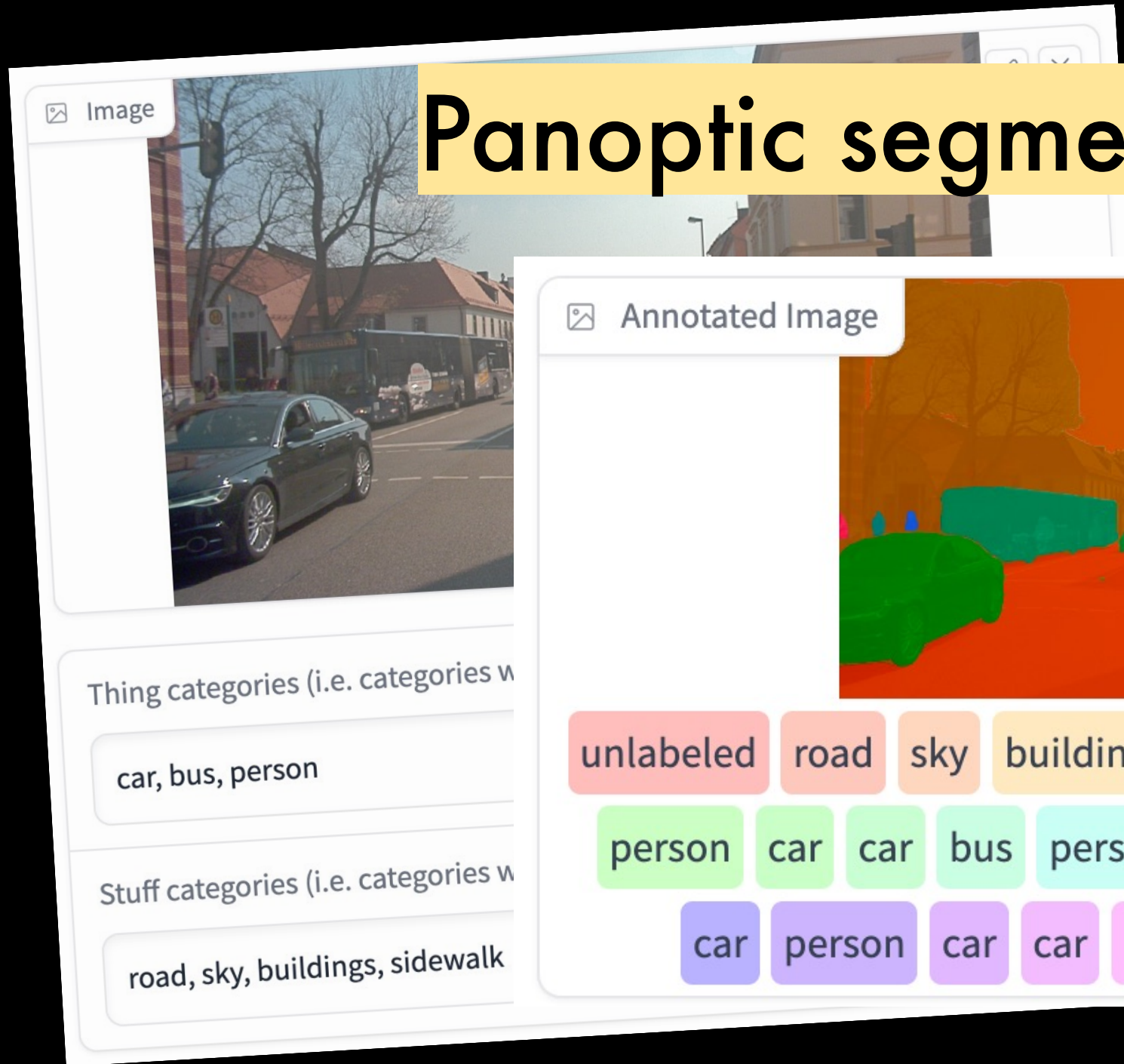
Semantic segmentation

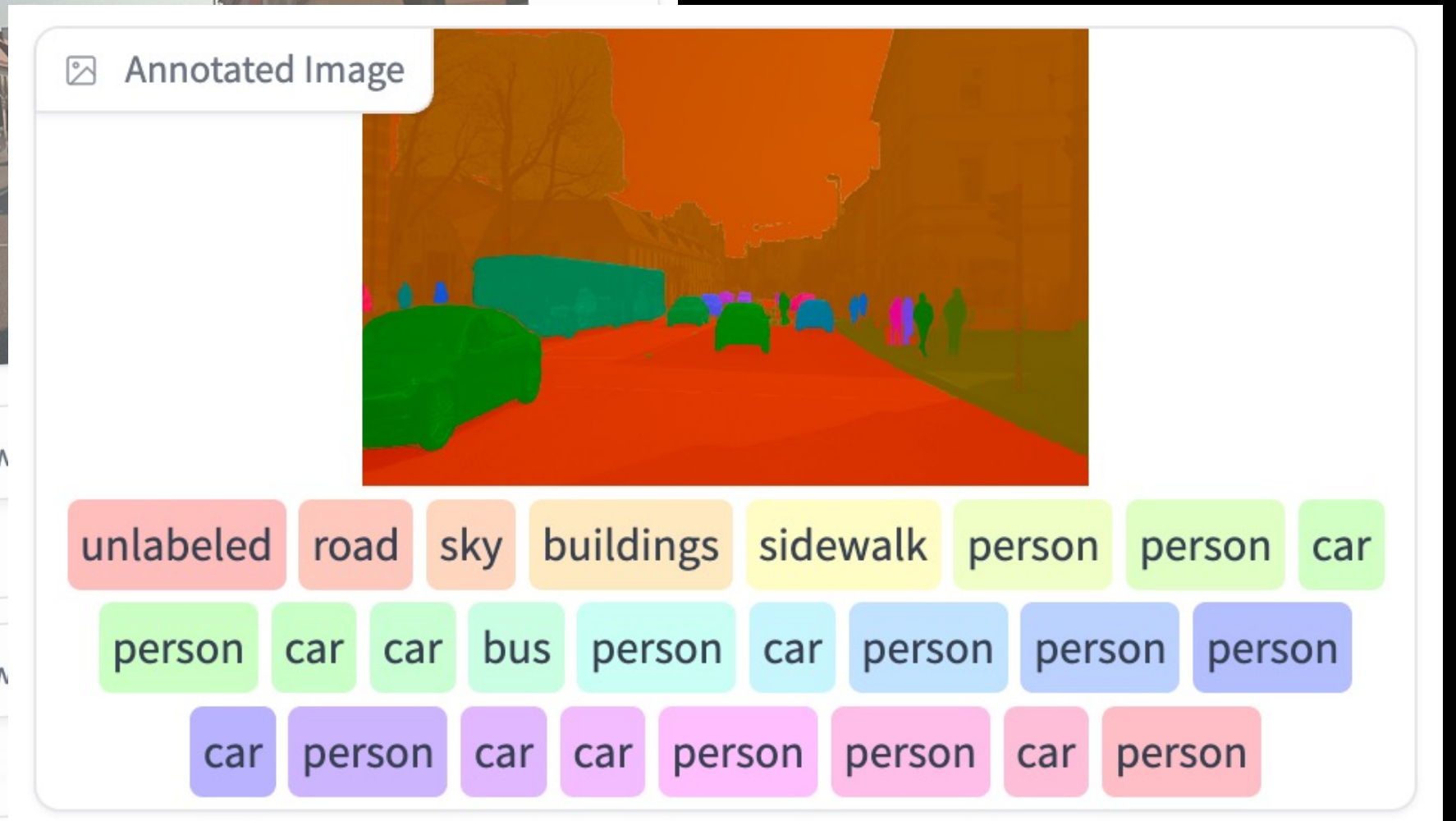


Info

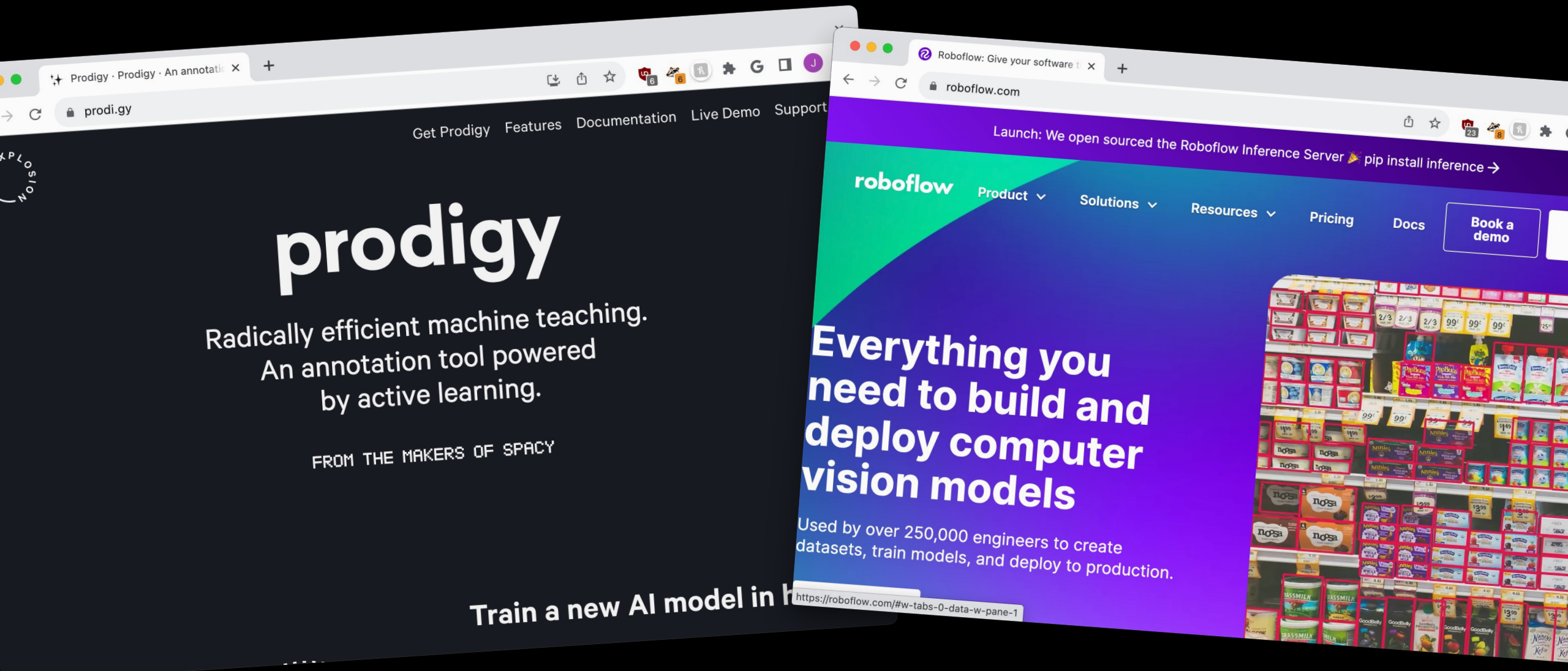
Object Label	Pixel Percent	Square Length
vegetation	15.46 %	0.39 m

Panoptic segmentation



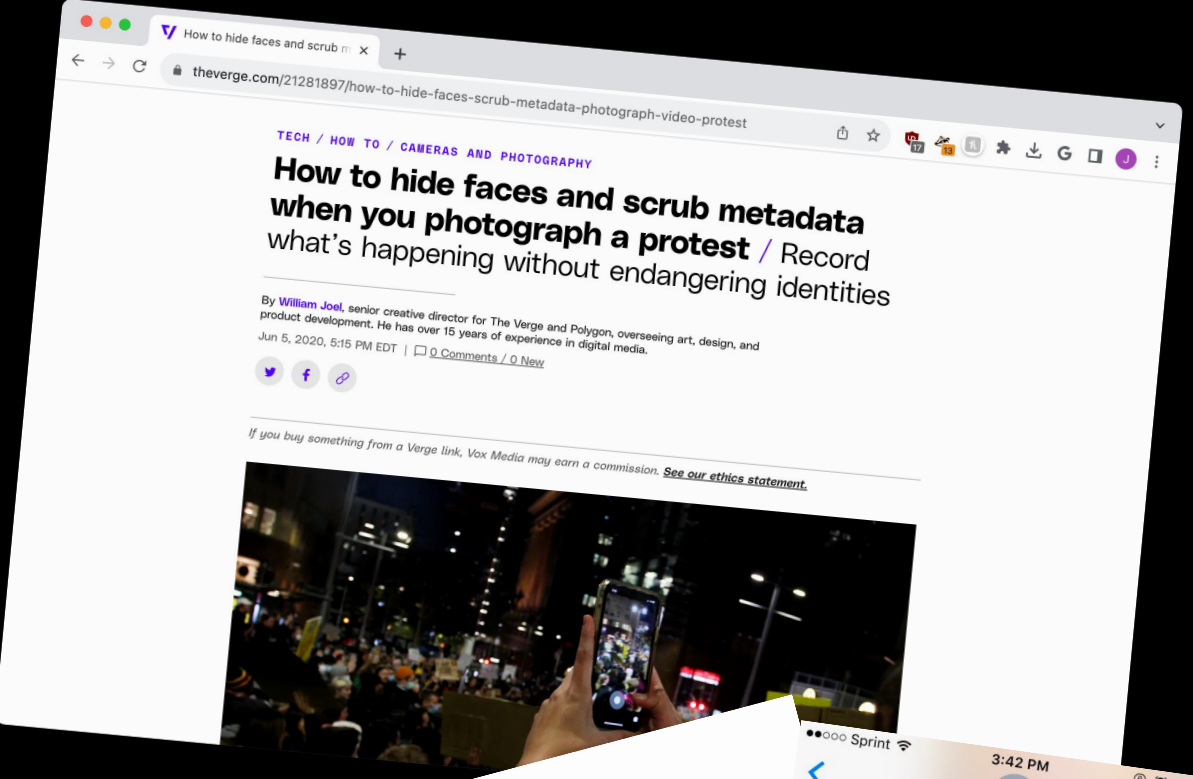


(you might need to fine-tune)



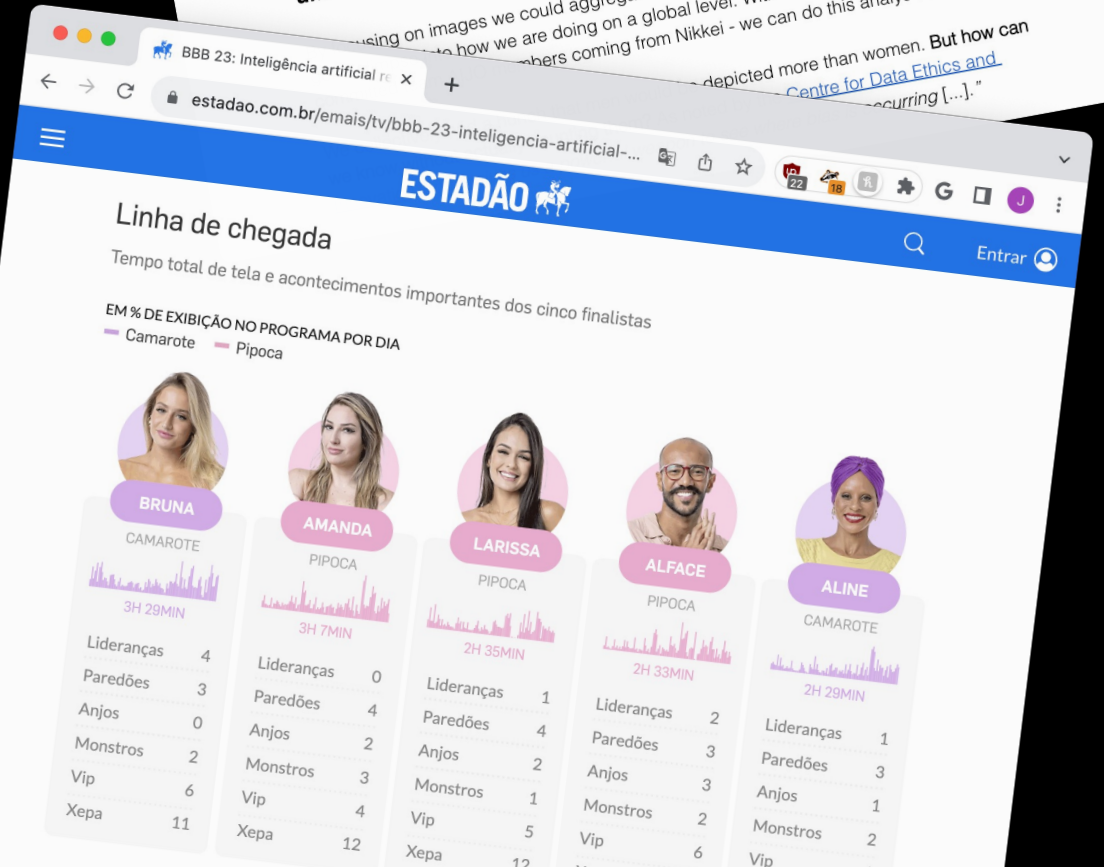
A simple combination

1. Object detection of faces. Cut them out!
2. Add in a classifier for the faces
3. You now have *a million and one possibilities*


How to hide faces and scrub m
theverge.com/21281897/how-to-hide-faces-scrub-metadata-photograph-video-protest
TECH / HOW TO / CAMERAS AND PHOTOGRAPHY
How to hide faces and scrub metadata when you photograph a protest / Record what's happening without endangering identities
By William Joel, senior creative director for The Verge and Polygon, overseeing art, design, and product development. He has over 15 years of experience in digital media.
Jun 5, 2020, 5:15 PM EDT | 0 Comments / 0 New
If you buy something from a Verge link, Vox Media may earn a commission. See our ethics statement.


How The New York Times Uses Software To Recognize Members of Congress
Jeremy Bowers · Follow
Published in NYT Open · 4 min read · Jun 6, 2018
804
3
Sprint
3:42 PM
61%
(202) 796-5943
(2/2) certainly - confidence = 100.0)
Cory Gardner (purple, almost certainly - confidence = 100.0)
(1/2) John Cornyn (red, almost certainly - confidence = 100.0)
Mitch McConnell (navy, almost certainly - confidence = 98.0)
John Barrasso (green, almost


Experimenting with Computer
aijoproject.com/post/experiment-documentation
THE AIJO PROJECT
Home
Learnings
Quiz
Team
LSE JournalismAI
Experimenting with Computer Vision
Thanks to the recent technological advancements, we can now use AI to both detect faces and identify their genders. We believe that this type of software can help us identify newsroom gender biases by uncovering who we actually depict.
sing on images we could aggregate news media content from all over the world to how we are doing on a global level. With the help of AI - and two bers coming from Nikkei - we can do this analysis at scale.
depicted more than women. But how can
Centre for Data Ethics and
curring [...]."


BBB 23: Inteligência artificial
estadao.com.br/emais/tv/bbb-23-inteligencia-artificial-...
ESTADÃO
Entrar
Linha de chegada
Tempo total de tela e acontecimentos importantes dos cinco finalistas
EM % DE EXIBIÇÃO NO PROGRAMA POR DIA
Camarote
Pipoca
BRUNA
CAMAROTE
3H 29MIN
Lideranças 4
Paredões 3
Anjos 0
Monstros 2
Vip 6
Xepa 11
AMANDA
PIPOCA
3H 7MIN
Lideranças 0
Paredões 4
Anjos 2
Monstros 3
Vip 4
Xepa 12
LARISSA
PIPOCA
2H 35MIN
Lideranças 1
Paredões 4
Anjos 2
Monstros 1
Vip 5
Xepa 12
ALFACE
PIPOCA
2H 33MIN
Lideranças 2
Paredões 3
Anjos 3
Monstros 2
Vip 6
ALINE
CAMAROTE
2H 29MIN
Lideranças 1
Paredões 3
Anjos 1
Monstros 2
Vip

I don’t know anything about video!!!

I just cut it up into images and use image models on it

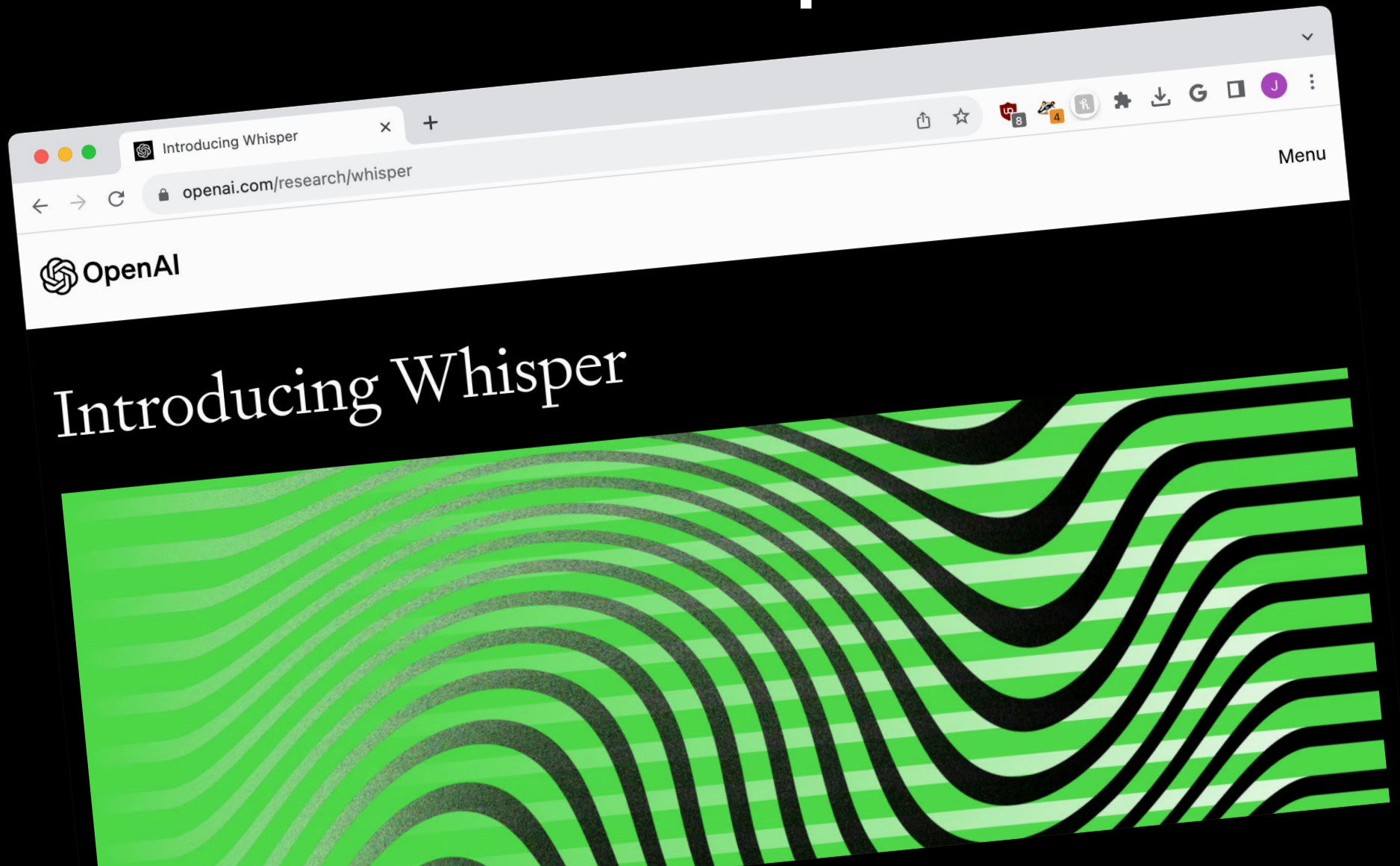
The most fun thing I know of is a tool from a company called Runway

source video + text = new video

Use this to make explainers!!!

Audio models

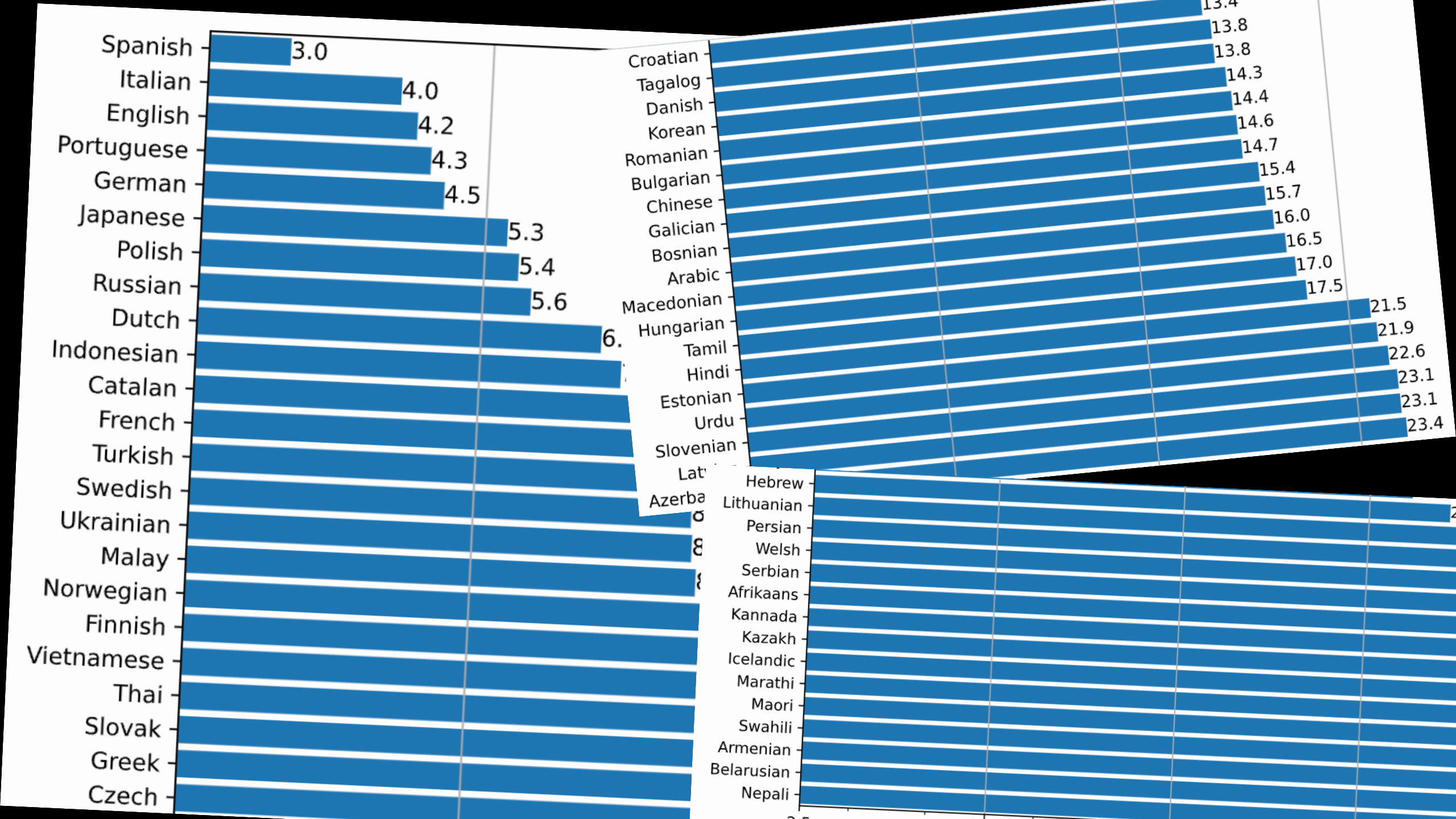
Transcription



```
model = whisper.load_model("base")


result = model.transcribe("6313.mp3")
```

```
' Okay, I was wondering if you could tell me a little bit about the progra
an intensive program designed to last four weeks, each session is four wee
that will be two weeks long. But our general four weeks session involves s
lves starting at nine o'clock in the morning with three hours of grammar i
an hour of group conversation with the same instructor, same class, but we
so students will usually go out to a garden setting because it's a little
h, students return and we have a complete flip of instruction in that ins
have hands on and we offer workshops in Mejibak's drop leaving, regional
have larg…'
```


Spanish 3.0
Italian 4.0
English 4.2
Portuguese 4.3
German 4.5
Japanese 5.3
Polish 5.4
Russian 5.6
Dutch 6.
Indonesian
Catalan
French
Turkish
Swedish
Ukrainian
Malay
Norwegian
Finnish
Vietnamese
Thai
Slovak
Greek
Czech
Croatian 13.4
Tagalog 13.8
Danish 13.8
Korean 14.3
Romanian 14.4
Bulgarian 14.6
Chinese 14.7
Galician 15.4
Bosnian 15.7
Arabic 16.0
Macedonian 16.5
Hungarian 17.0
Tamil 17.5
Hindi 21.5
Estonian 21.9
Urdu 22.6
Slovenian 23.1
23.1
23.4
Hebrew
Lithuanian
Persian
Welsh
Serbian
Afrikaans
Kannada
Kazakh
Icelandic
Marathi
Maori
Swahili
Armenian
Belarusian
Nepali

Audio models can do a *lot*

this is Microsoft's SpeechX

Sample audio 

Generated audio

Text to read

"that summer's emigration however being mainly from the free states greatly changed the relative strength of the two parties"

(please don't fire everyone and replace them with AI voices)

Speaker diarization

This is only a sliver of the types of models and abilities that exist out there.

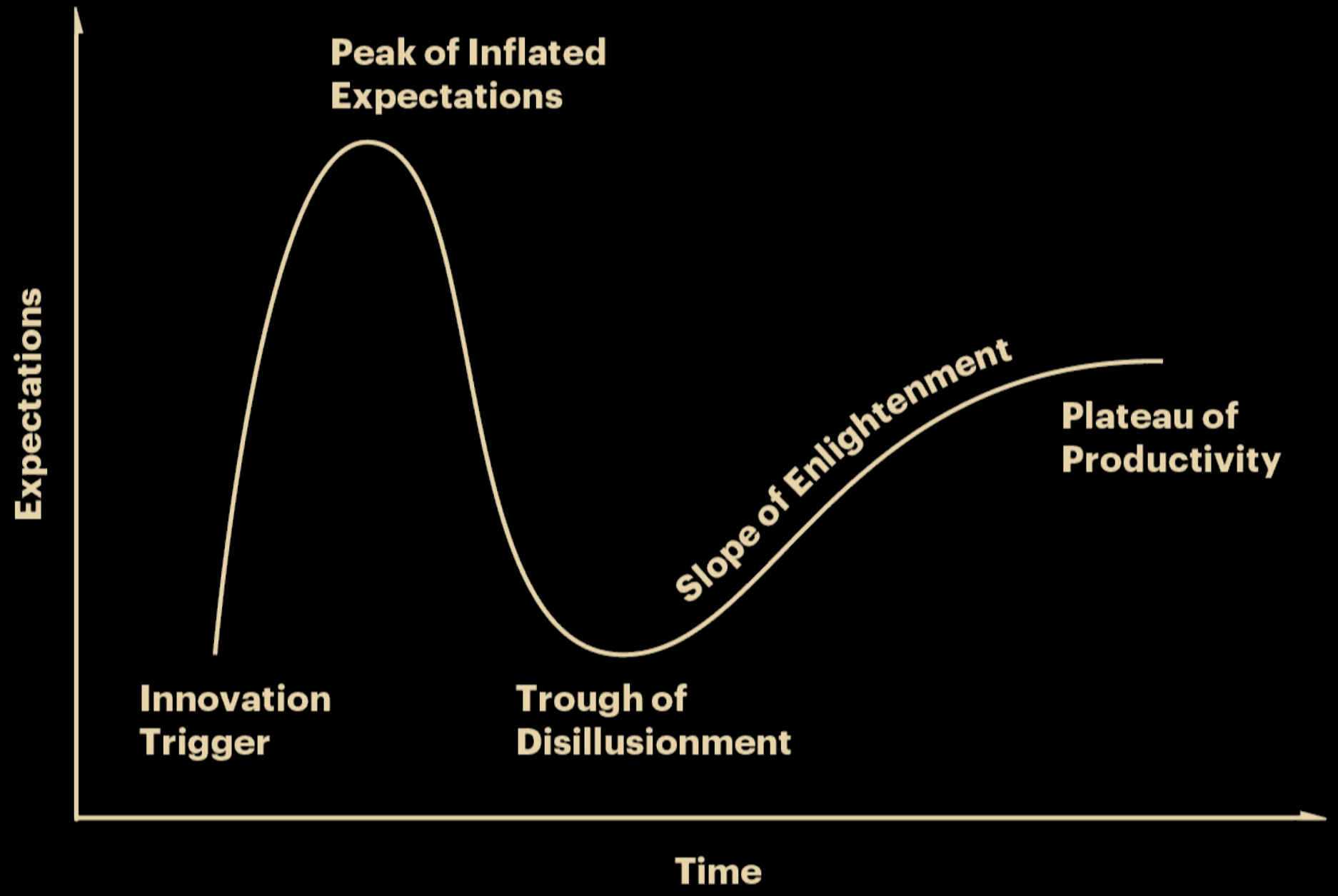
I am begging you to make an
internal playground
of accessible, connected pieces

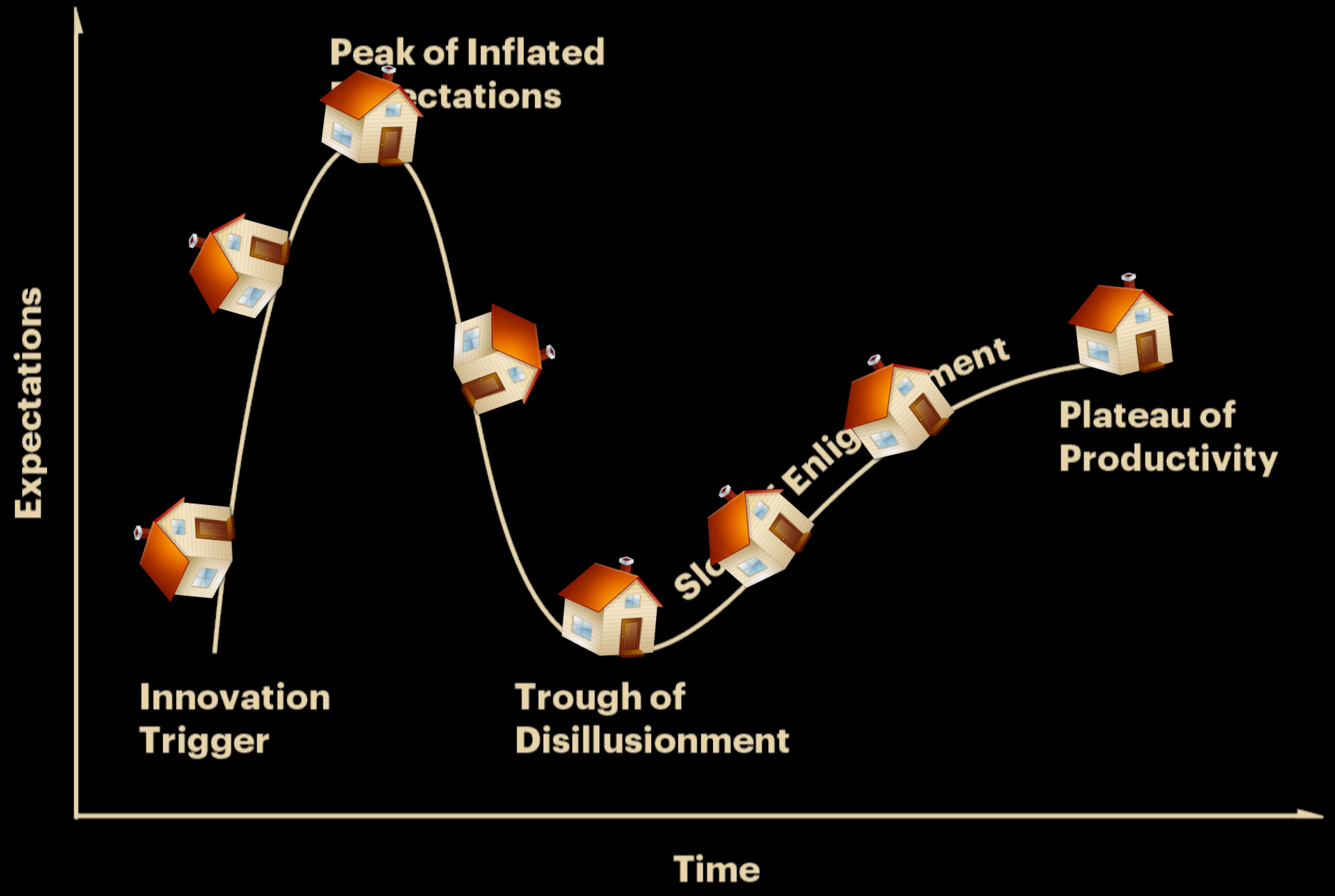
Benefits of a playground

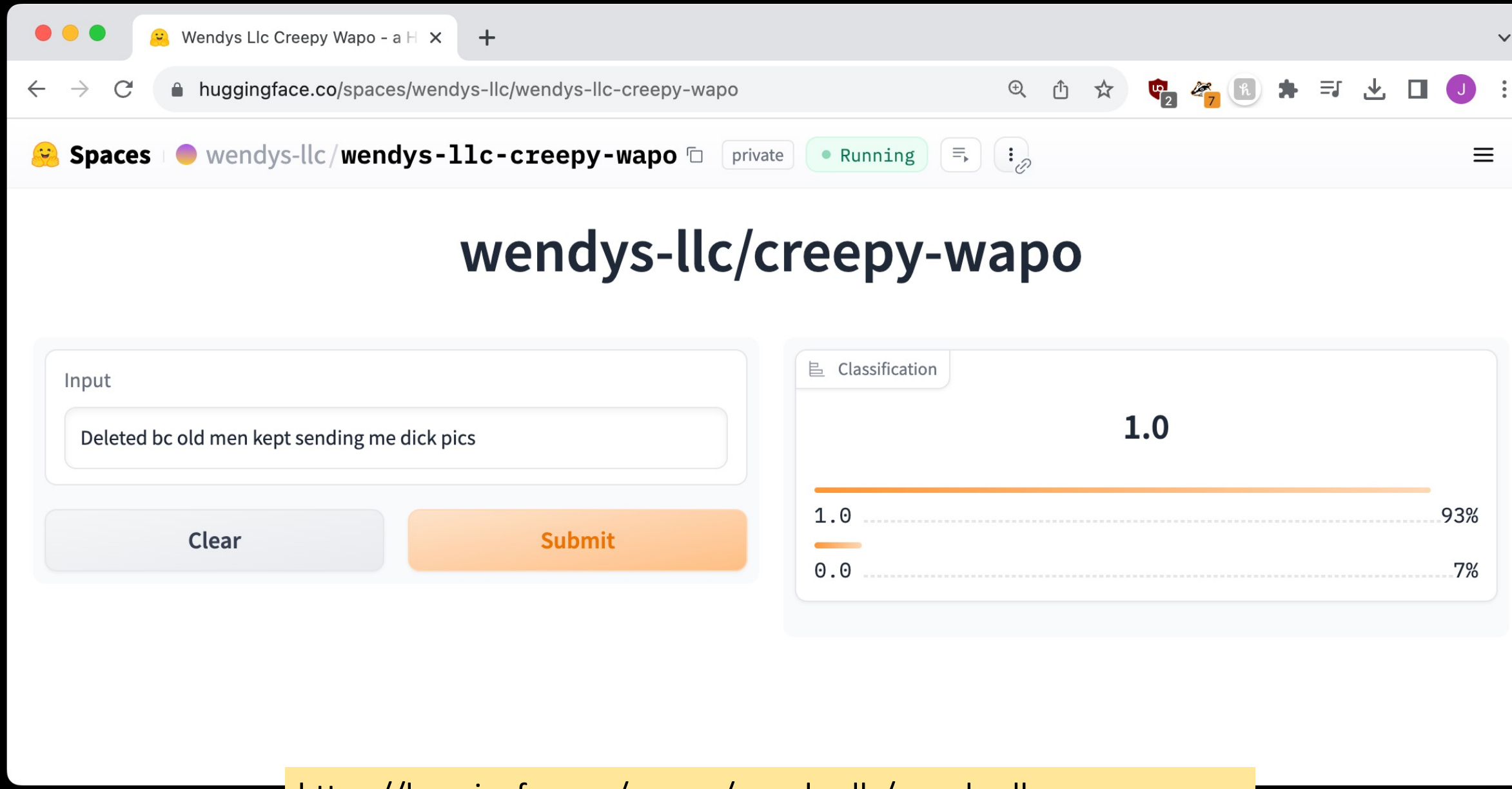
Ideation and experimentation

Builds trust within newsroom

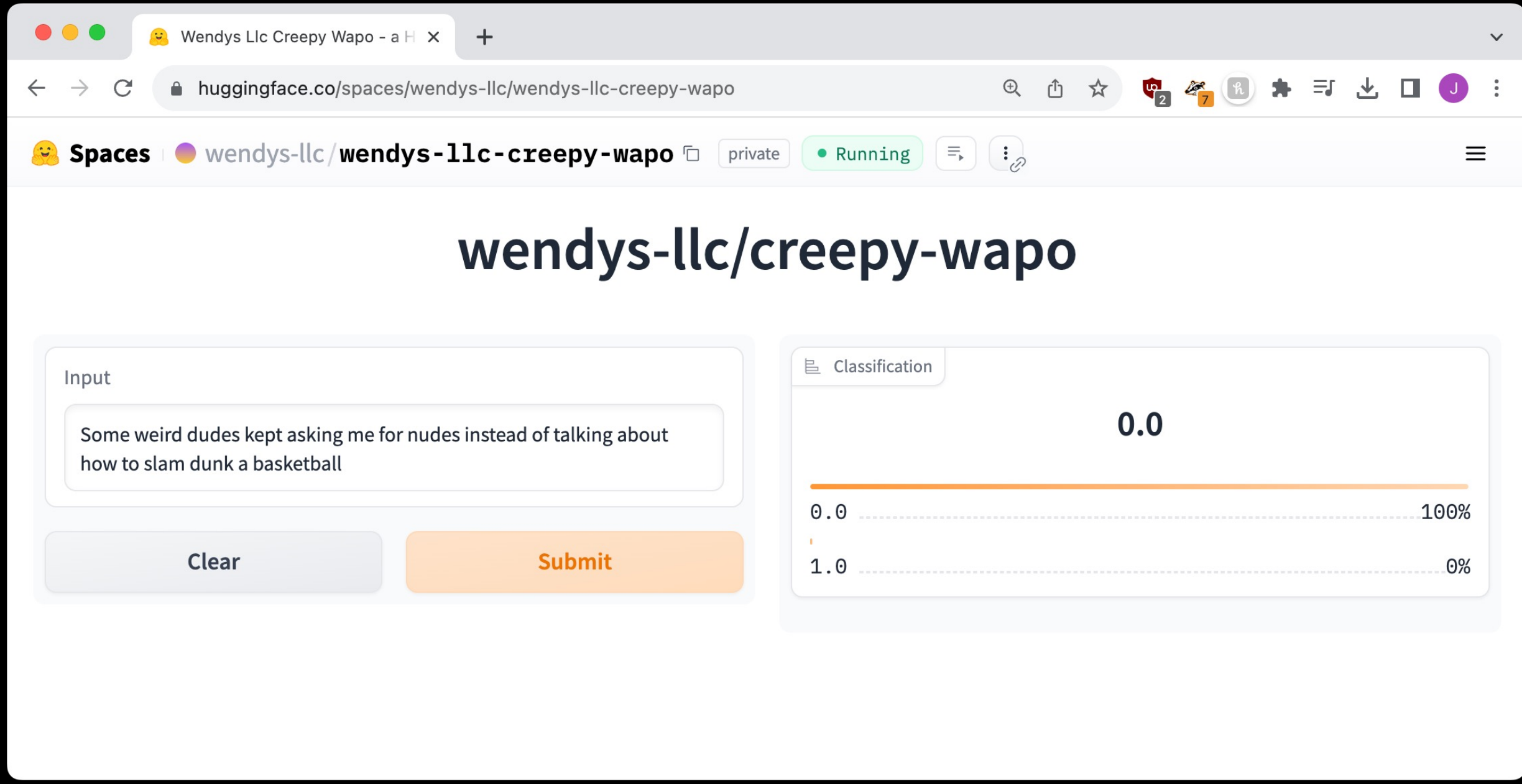
Easy "red teaming" from non-tech folks


Peak of Inflated Expectations
Expectations
Slope of Enlightenment
Plateau of Productivity
Innovation Trigger
Trough of Disillusionment
Time

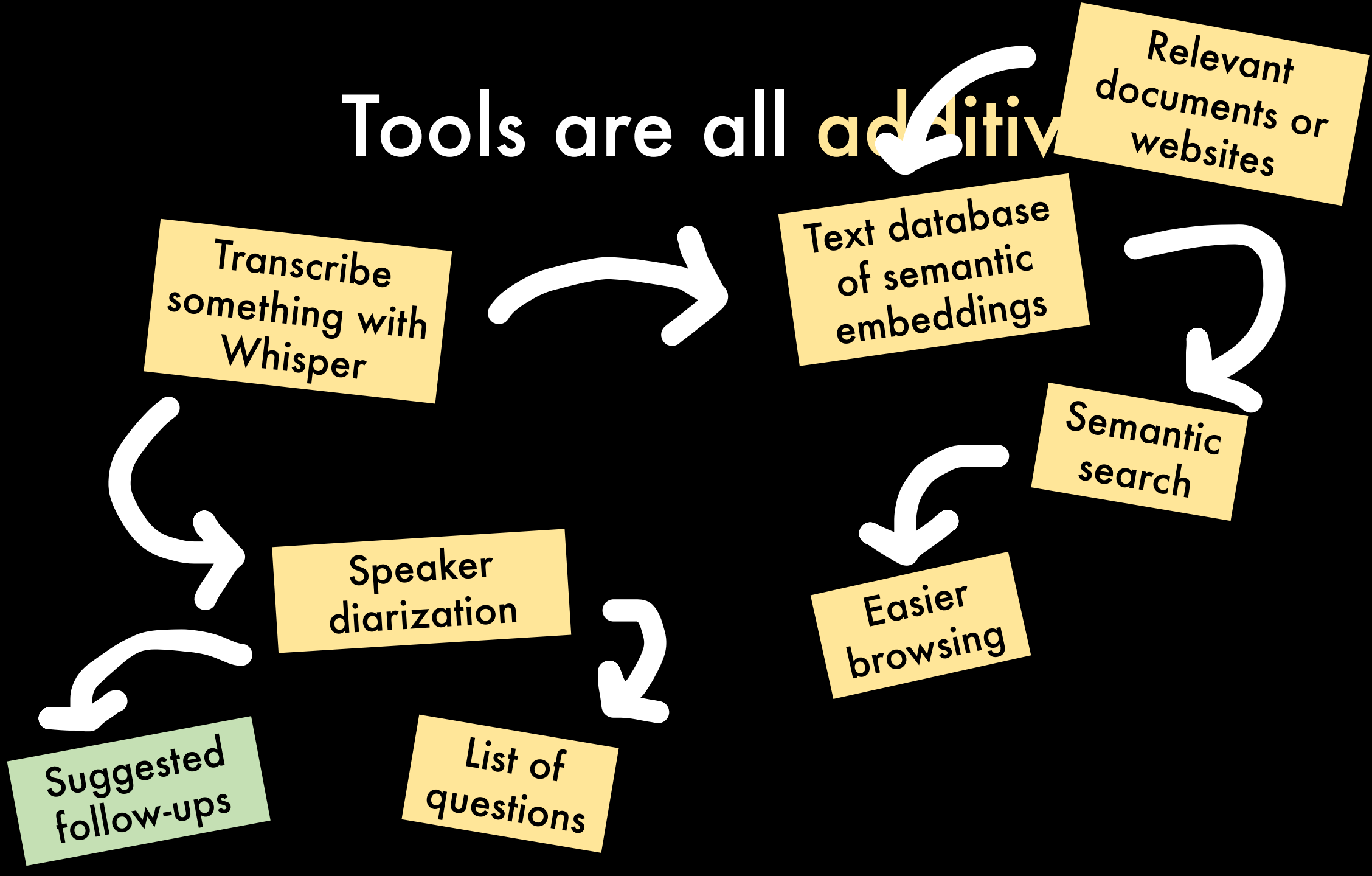

Peak of Inflated
ectations
Expectations
Innovation
Trigger
Trough of
Disillusionment
Plateau of
Productivity
Time



https://huggingface.co/spaces/wendys-llc/wendys-llc-creepy-wapo


Wendys Llc Creepy Wapo - a H
huggingface.co/spaces/wendys-llc/wendys-llc-creepy-wapo
Spaces
wendys-llc/wendys-llc-creepy-wapo
private
Running
wendys-llc/creepy-wapo
Input
Some weird dudes kept asking me for nudes instead of talking about how to slam dunk a basketball
Clear
Submit
Classification
0.0
0.0
100%
1.0
0%

Tools are all additive



Testing + tracking infrastructure

Data overview

The data is exclusively from summaries of management reports in financial reports in the period 20 January to 11 April, 2023. In this period 92 articles were given an accepted summary. 54 of the articles were given an acceptable summary in the first attempt (59%), while the remaining 38 articles took several attempts. A total of 193 summaries were generated, of which 101 summaries were rejected. We didn't give up on any articles.

The rejected 101 summaries fell into the following categories [number of articles] (number of summaries):

- Missing important information [14] (28)
- Incorrect summary [14] (23)
- AI evaluates content [9] (17)
  (e.g. the model states that the company has had a successful year)
- Poor language [9] (12)
- Irrelevant information [4] (6)

https://generative-ai-newsroom.com/summaries-in-danish-with-openai-cbb814a119f2

Model tasks mentioned today

Categorizing pieces of text
Extracting people/places/things from text
Text embeddings (conceptual document similarity)
Detecting objects in images (pools, faces)
Detecting stuff in images (vegetation, streets)
Classifying images (illegal amber mines, gender)
Transcription
Speaker detection
Text-to-audio
Video + text = new video
Video -> audio -> text

Your homework

List out all the content and data you have available. How could you describe it, add metadata, connect it, or republish it using the tools we talked about?

bit.ly/ona23-beyond-chatgpt

AI tools from before and beyond ChatGPT

Jonathan Soma
js4571@columbia.edu · @dangerscarf
aifaq.wtf · normalai.org · investigate.ai
Knight Chair in Data Journalism, Columbia University