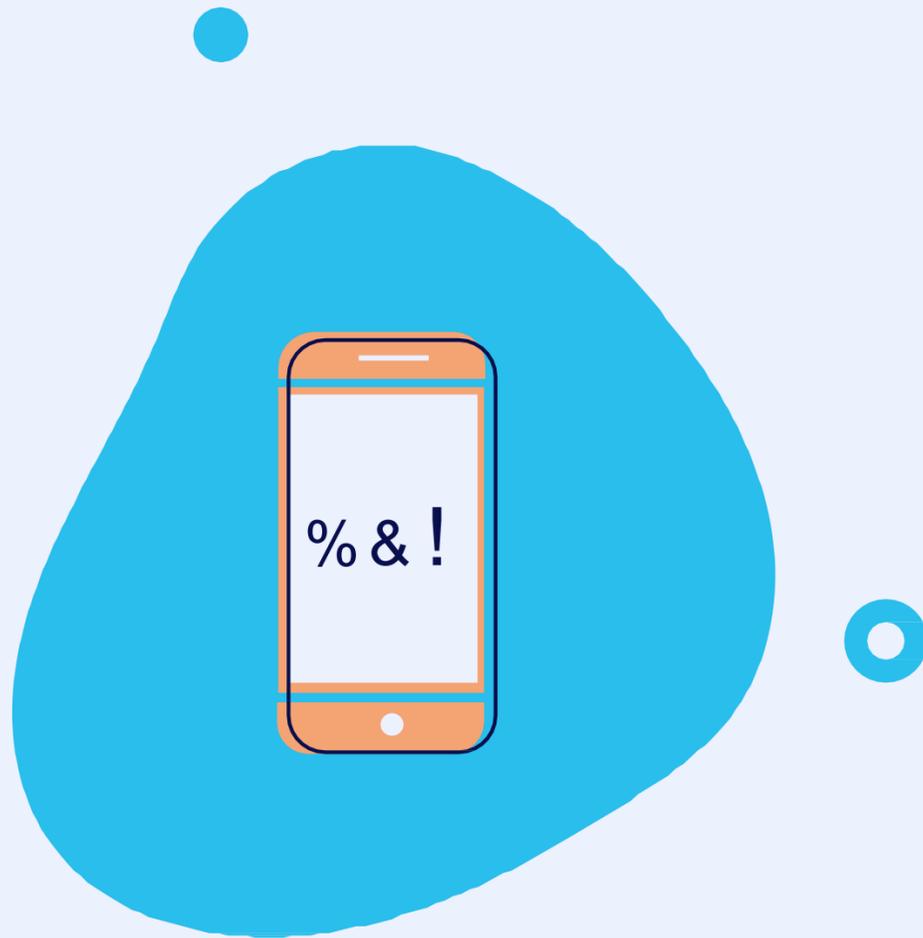


EMAKIA

A system filtering out harassment
on incoming social media data



Corinne David
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"We weren't expecting any of the abuse and harassment and the ways that people have weaponized the platform."

Jack Dorsey
Founder & CEO of Twitter
January 19, 2019



OBJECTIVE:

Create a Safer Environment



Goo Hara



Greta Thunberg



Ilhan Omar



Christine Blasey Ford



Tyler Clementi



Alexandria Ocasio-Cortez



Sulli

→ WOMEN

→ MINORS

→ PERSONS of COLOR

→ DISADVANTAGE GROUPS

→ YOU?

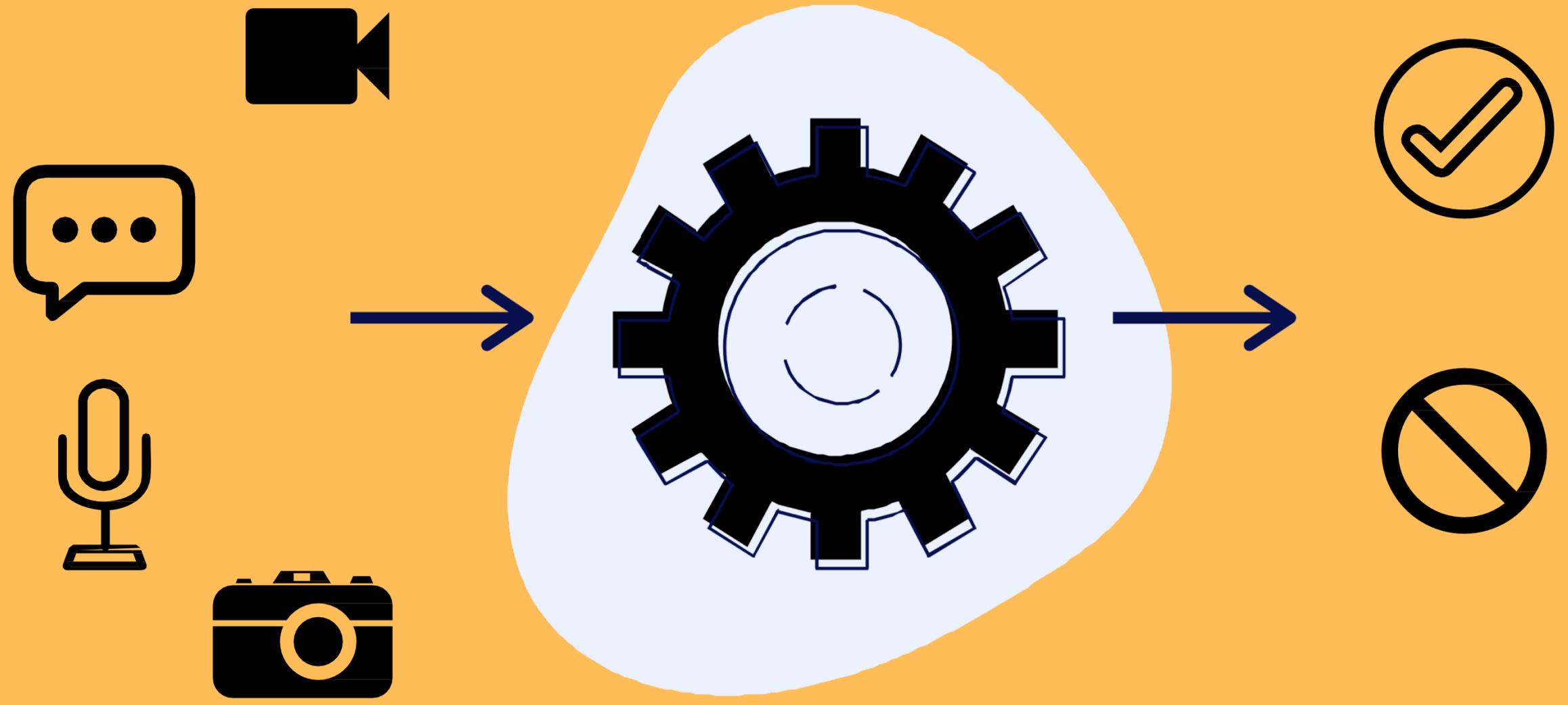


IT CAN BE DONE

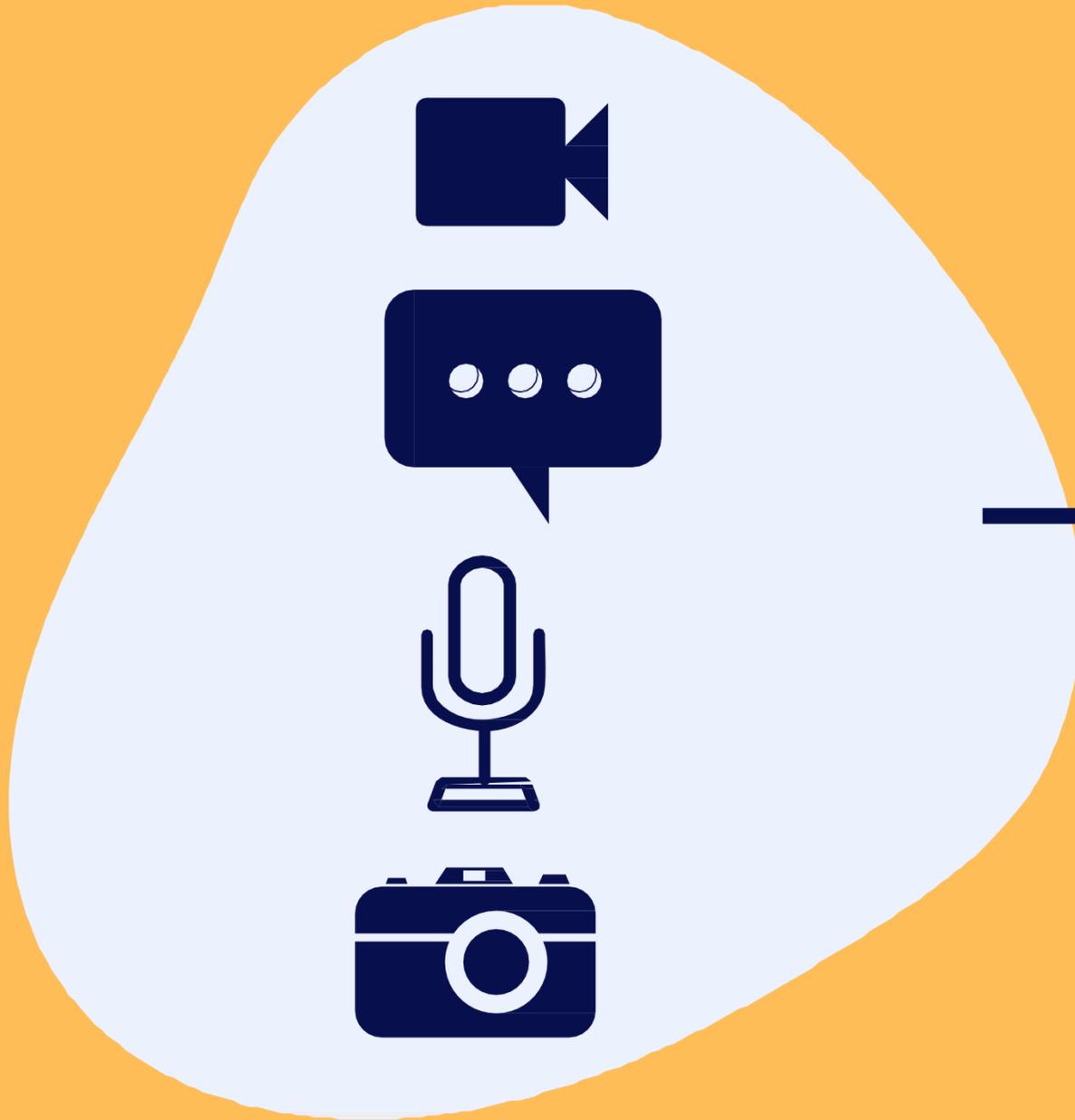
- NYT & Google: Filter incoming data using text classifier to train a model on 16 million comments
- Email Software uses text classifiers to determine whether incoming mail is sent to the inbox folder or the spam folder
- Discussion forums use text classifiers to determine if a comment is appropriate



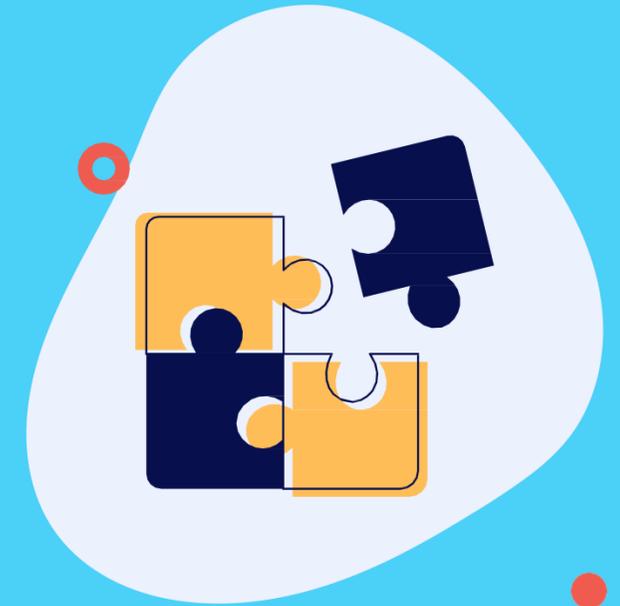
EMAKIA's multimodal solution



A closer look



THE SYSTEM



DATA

- Collect, label and clean data
70 000 English Tweets,
200 000 Italian labeled Tweets

TRAIN MODEL

- Apple Core ML Classifiers
- Google AutoML Classifiers

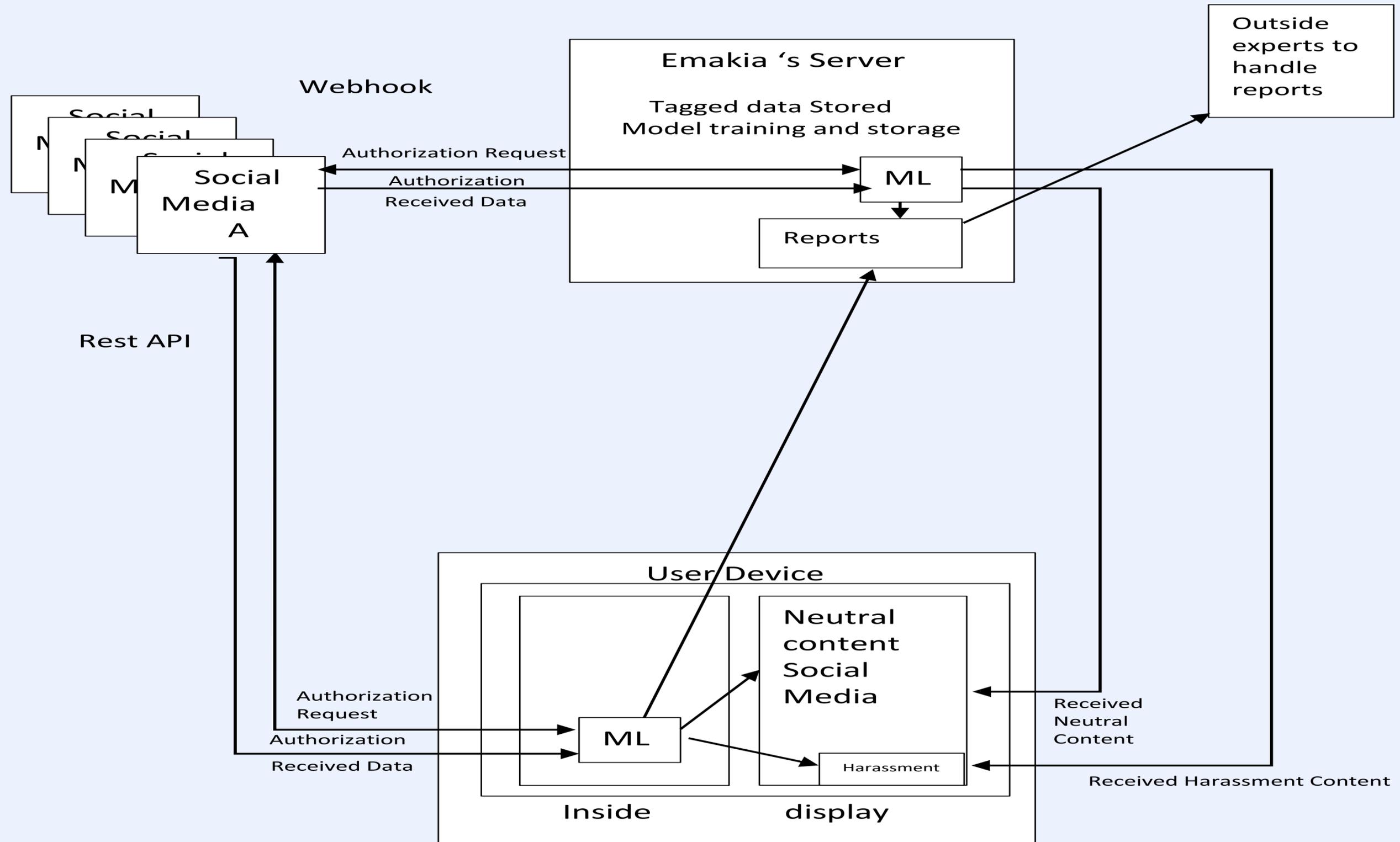
ENAËLLE

- Mobile Application running on iPhone or Android
- Detect harassment on real-time data with the model

REPORTS

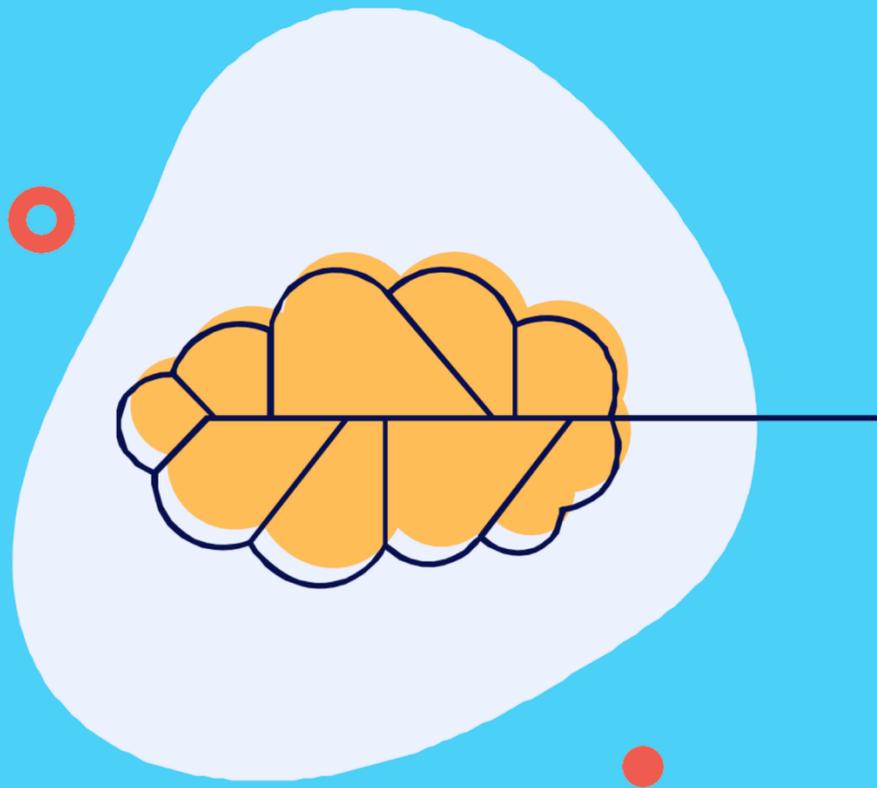
- Harassment received
- History of the harassing senders, friends and followers



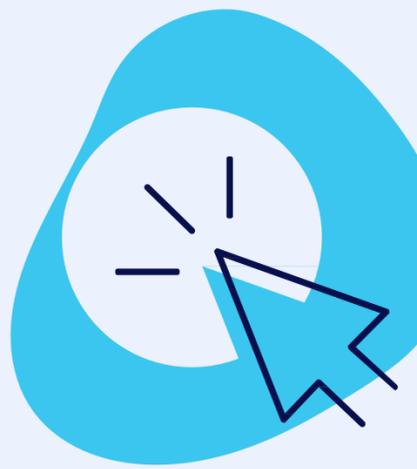


What's running on the server?

- Collect, label and store data
- Validation of the data
- Train ML classifiers for each language with the labeled data
- Programs to search social media data for harassment data
- Process text, image, audio, and video
- Create reports of the detected harassment, fake news, deepfake



ENAËLLE



→ Register an email of a buddy/mentor; Emakia sends reports to the user and the buddy system



Neutral content of the social media platforms is displayed



→ Portal to access the different social media platforms

Harassing content is accessible with a Tab bar

→ Provide the same functionality as the social media platform



Text Classification

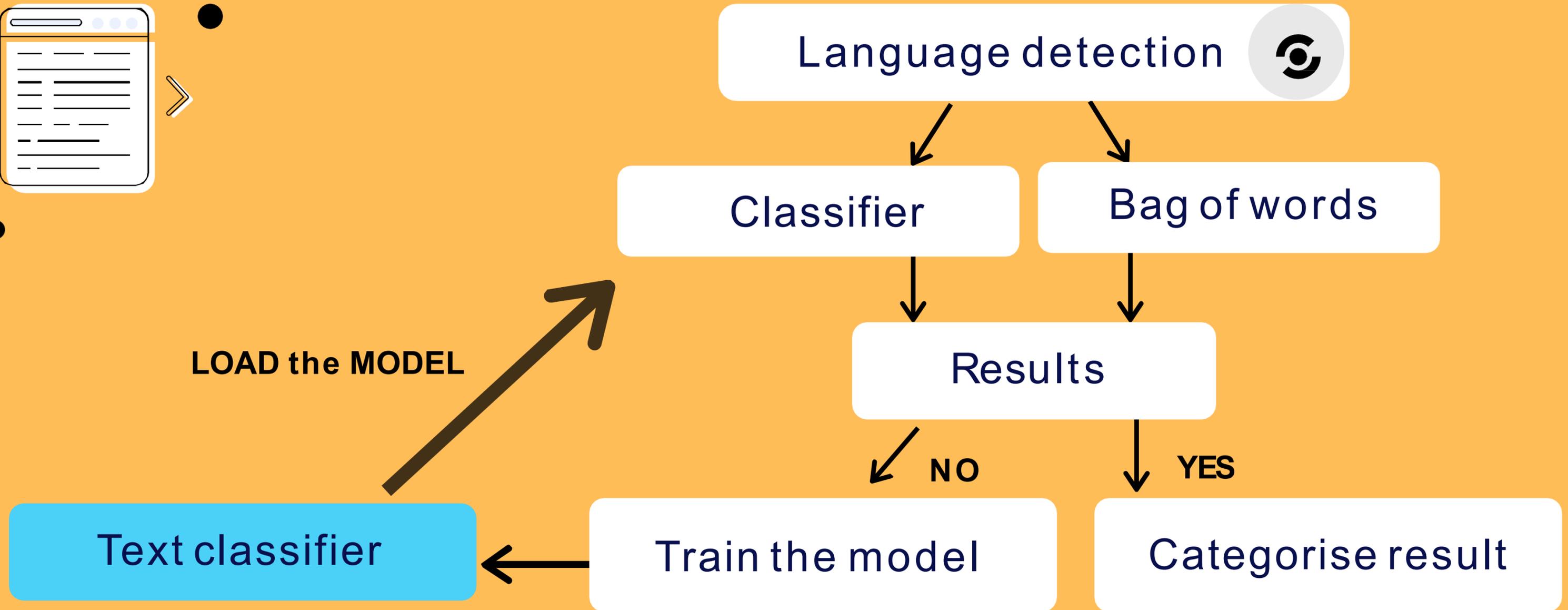
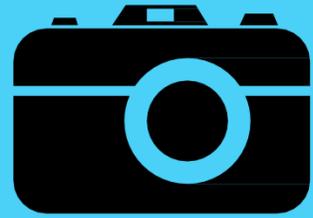


Image Classification



OCR

NO Text

WITH Text

Image Classifier

Text Classifier

1. APPLE:

Core ML Vision

Apple Vision Framework

2. GOOGLE:

Auto ML Vision

Vision API

HOW TO FEED THE DATA?

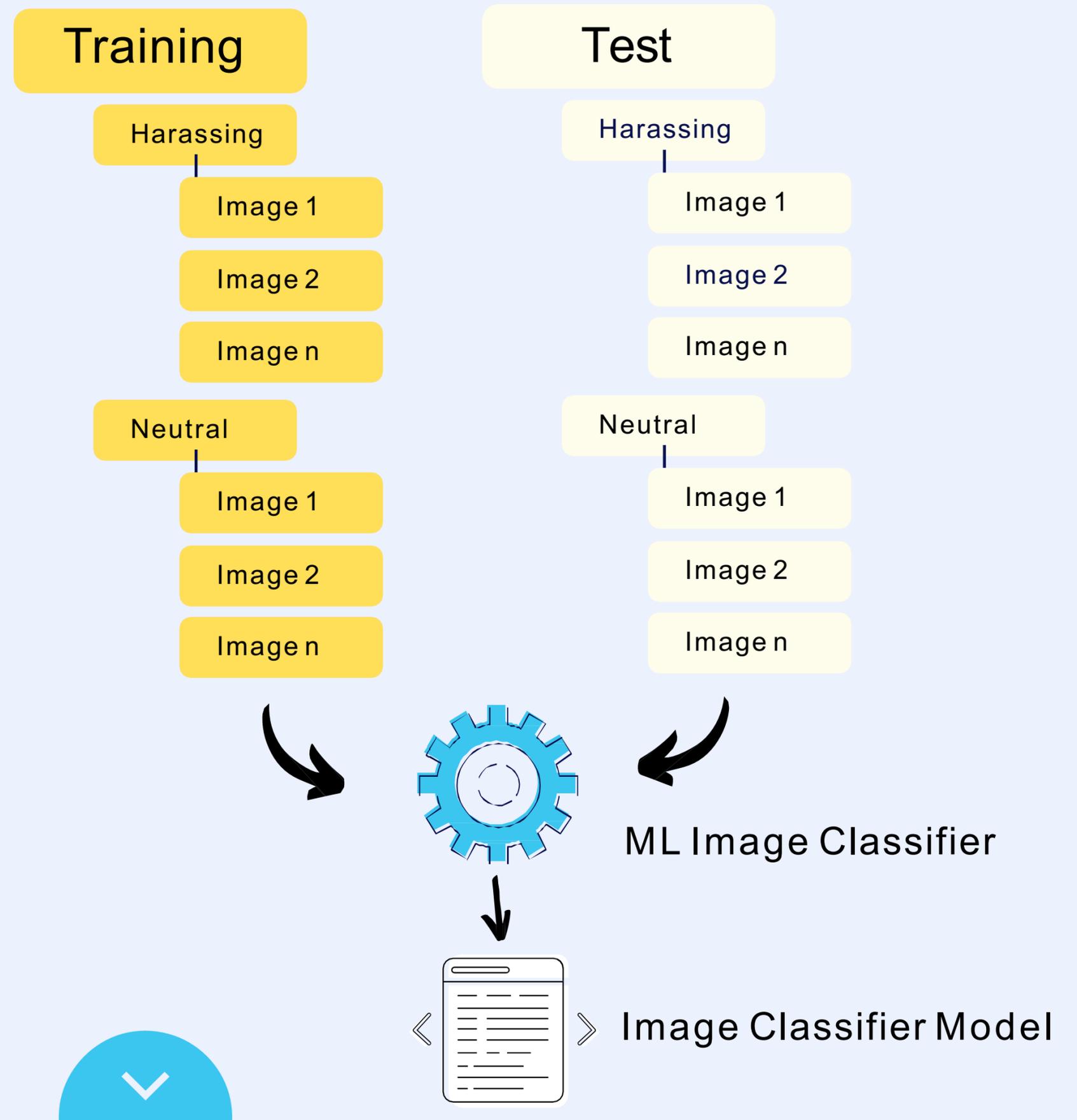


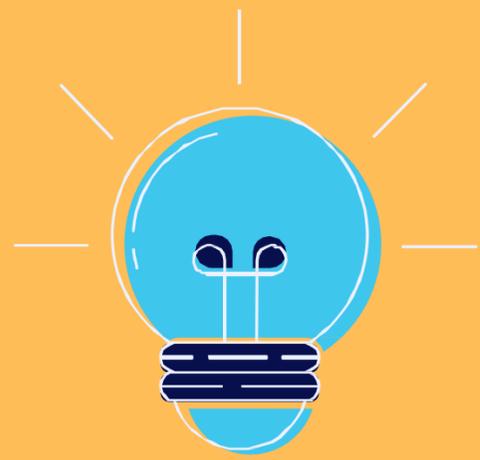
Core ML Audio



Core ML Video Apple Vision
AutoML Video

Audio & Video classification





Other applications

- **FAKE NEWS
DETECTION**

- Reference source with fact checking websites
- Labeled data - fake or true
- Train the model with labeled data and features

- **VIDEO CLASSIFIER
FOR DEEPPFAKE VIDEO**

- Detect artifacts on the video
- Detect Face Swap
- Detect Lip Sync on the audio file of the video



THIRD PARTY DATA



- NAACL_SRW_2016.csv (Waseem et al., 2016) “Hateful Symbols or Hateful People? Predictive Features for Hate Speech Detection on Twitter”
- BullyingV3.0.zip (Xu, 2012) “Learning from bullying traces in social media”
- Twitter-hate-speech-classifier-DFE-a845520.csv from data.world created on Nov. 21, 2016 by @crowdfowerdata
- Labeled_data.csv (Waseem et al., 2016) “Understanding Abuse: A Typology of Abusive Language Detection Subtasks”
- OnlineHarassmentDataset.csv (Golbeck et al., 2017). “A Large Labeled Corpus for Online Harassment Research”

OUR DATA

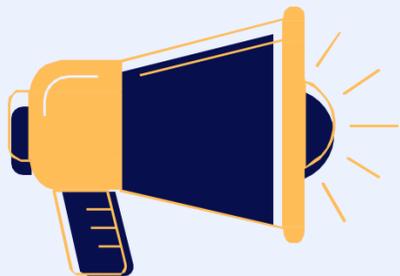
- A Program Collecting Harassing Data uses the standard search Twitter API to obtain tweets with specific harassment terms unknown to the model
- Using the bag-of-words adaptive filter and retraining the text classifier with content yet unknown to the model
- Italian data gathered by Claudia Zaghi



Bag of words resources

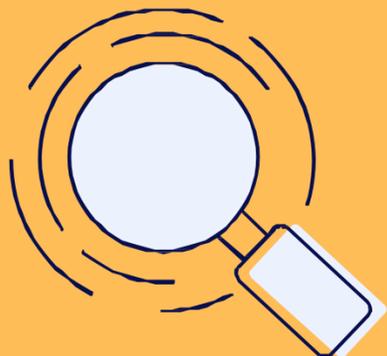
→ Hatebase_dict.csv provided by Hatebase, an online database of hate speech
“Automatic Detection of Cyberbullying on Social Media” by Love Engman Master's Thesis in Computing Science

→ “Automatic Detection of Cyberbullying on Social Media” by Love Engman Master's Thesis in Computing Science



Validation of Labeled Data & Bag-of- Words

- Verify that the data are labeled correctly
- Evaluate if a sentence is harassing by comparing its content against the three bag-of-words sets
- Neutral label might have some terms from the moderate set or the double meaning set
- Harassing sentences usually contain several abusive words and moderate words
- Bag-of-words updated with new terms and the label is corrected if needed



Core ML Model creation

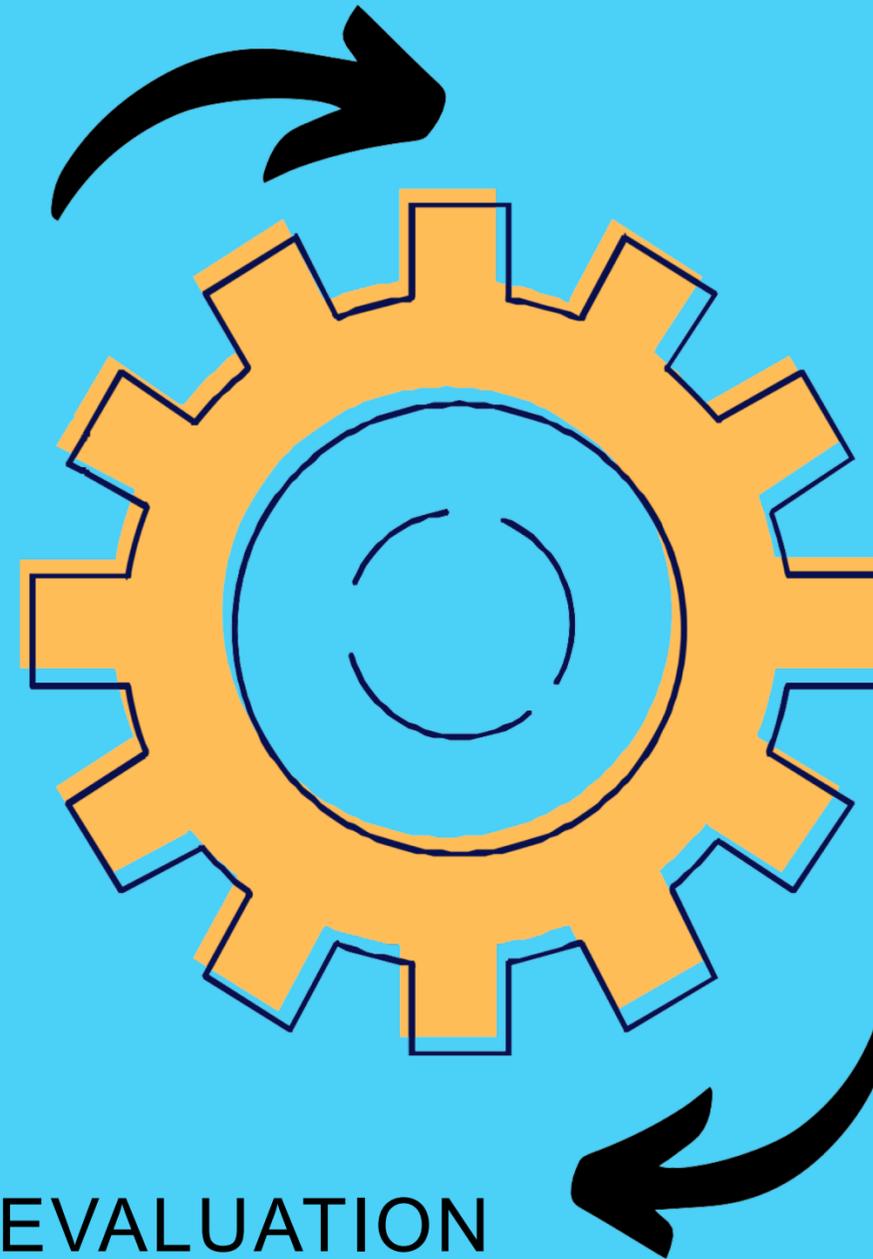
1. DATA Labeled Tweets

63,000 English
128,000 Italian

2. TRAINING

25 to 29 iterations

3. EVALUATION

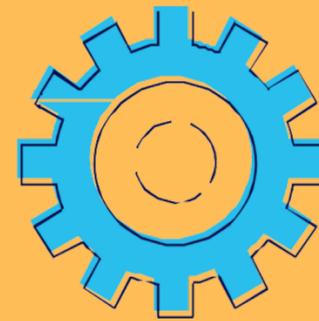




accuracy on training set:

99.34% English

98.12 % Italian



Text Classifier Model

Test set



accuracy on test set:

90.21% English

88.56 % Italian

Text Classifier Training & Testing

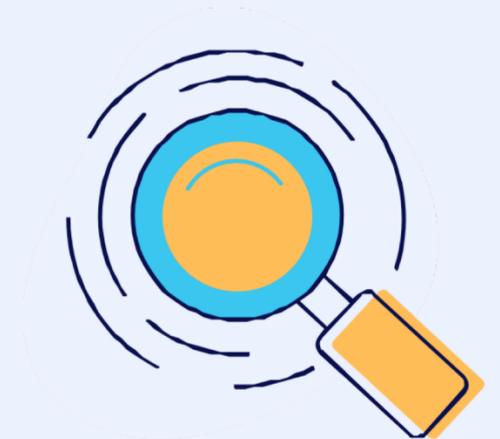
Automatic Feature extraction

entity recognition, script identification, tokenization, lemmatization,
parts-of-speech tagging, and language identification

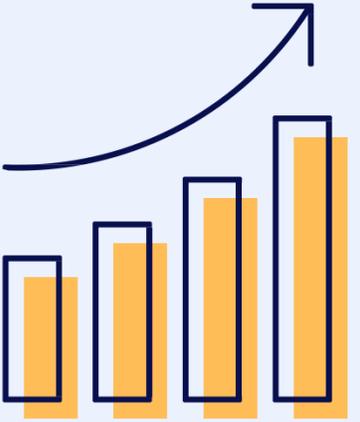


Model definition with Max Entropy

Choice among Logistic Regression, Nearest Neighbor Classifier,
Support Vector Machines, Boosted Decision Trees, Random Forests.



Text Classifier results



English Data

True\Predict	Harassment	Neutral
Harassment	87 %	13 %
Neutral	2 %	98 %

Auto ML

True\Predict	Harassment	Neutral
Harassment	85.5 %	14.5%
Neutral	8 %	92 %

Core ML

Italian Data

True\Predict	Harassment	Neutral
Harassment	88 %	12 %
Neutral	5 %	95 %

Auto ML

True\Predict	Harassment	Neutral
Harassment	86.5 %	13.5 %
Neutral	9.5 %	90.5 %

Core ML



Testing on Real-Time Data

- Enaëlle
- Testing application
 - Collect tweets from a list of users
 - with a high incidence of harassing terms
 - with a different style of writing
 - The application filters and evaluates tweets from the list
with classifiers



Improve the Accuracy



Core ML Text

Auto ML Text

Same results?

YES

NO

Label the content

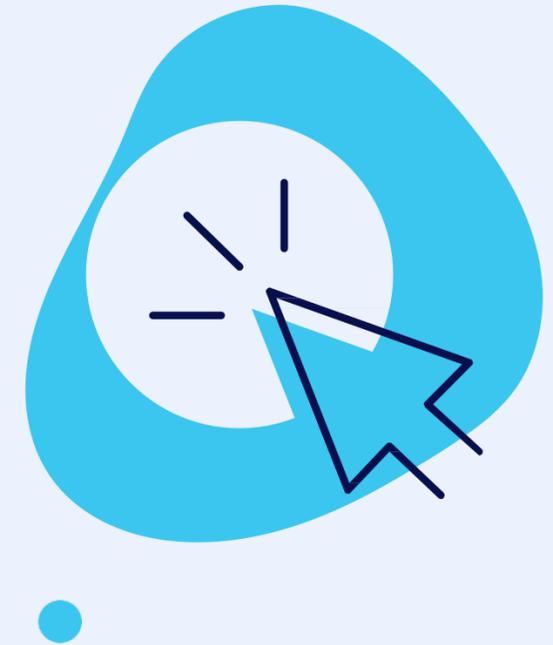
Label the content

- BoW
- History of the senders
- Followers history
- Results of the models

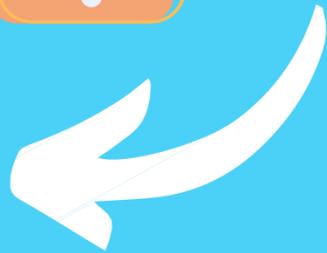


REPORT

- The report analyzes
 - How the harassment spreads
 - What the harassment topics are
 - What are the emotional sentiment the abusive content carries
 - Geographic location of the senders
- The report is sent to the appropriate responder to resolve the issue
- Request the social media platform to remove the harassing content
- A follow-up checks done to see what action was taken
- Reports are stored
- The report and any follow-ups are sent to the users and their buddy-mentors
- The sender is automatically blocked



OUTGOING DATA



- The ML classifier is not applied to the outgoing data
- When the user composes a tweet, it is sent with no alteration



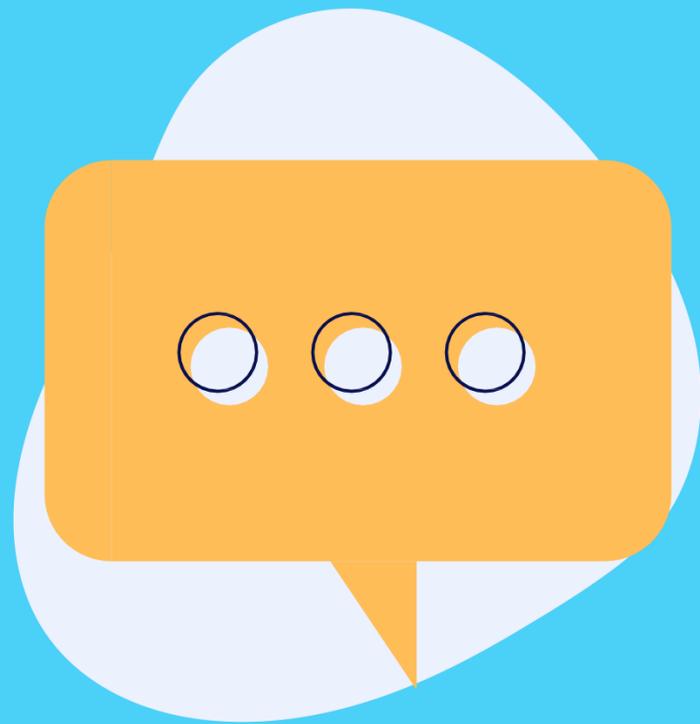
CUSTOM ML MODELS



- Customized ML classifier models are applied to incoming data on mobile devices and on the Emakia server
- The classifiers are tuned to the user's definition of what is harassing or not
- This process is done by retraining the classifier with content specific to the user preferences



Questions?



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