

« EMOTION RECOGNITION & GENUINENESS DETECTION »



SAPIENZA
UNIVERSITÀ DI ROMA

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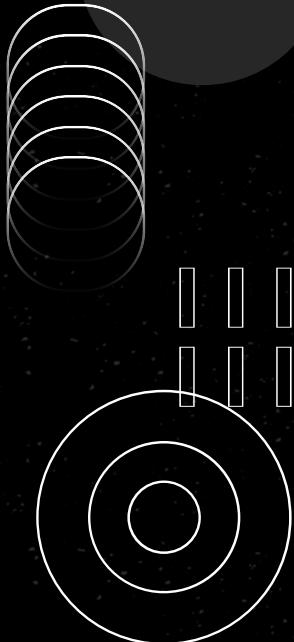
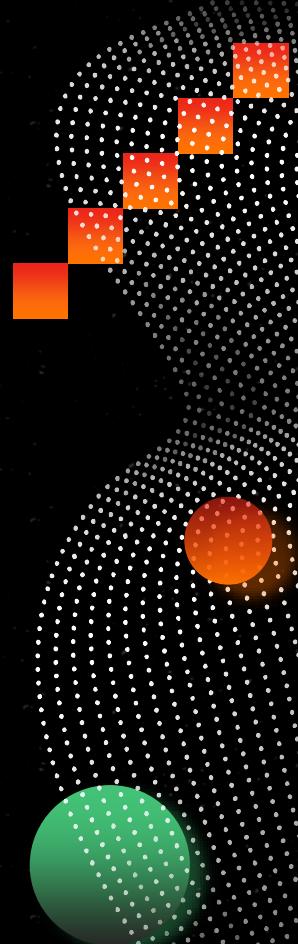


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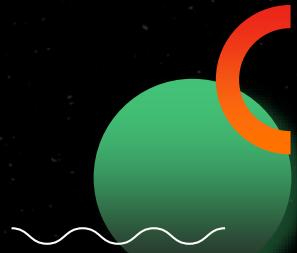


ABOUT THE PROJECT

This project aim to explore two different tasks in the field of facial expression:

- Emotion Recognition: Identifying the emotional state of a person based on their facial expression.
- Genuineness Detection: Determining whether the expressed emotion is genuine or deliberately posed

01



EMOTION RECOGNITION

02

For that task we have built a Convolutional Neural Network (CNN) model to classify facial expressions into one of seven basic emotions: Angry, Disgust, Fear, Happy, Sad, Surprise, and Neutral

The FER-2013 dataset, widely used in the research community for its diversity and complexity, is employed for training and evaluation



MODEL ARCHITECTURE

3 convolutional blocks + 2 fully connected layers

CONVOLUTIONAL LAYER



BATCH NORMALIZATION



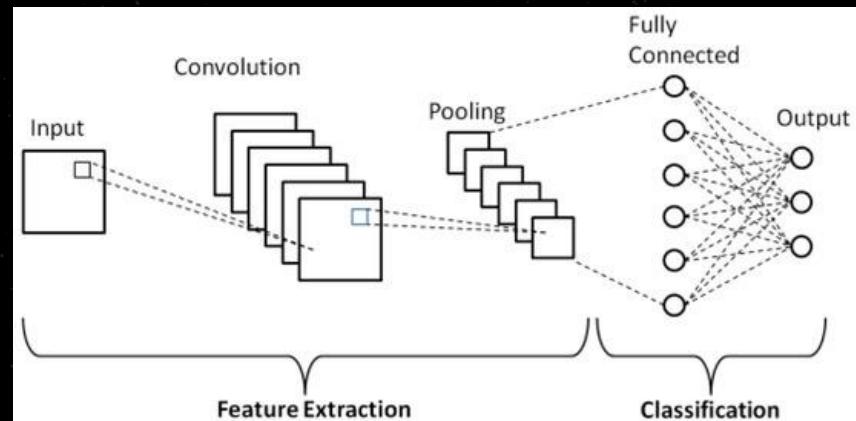
ACTIVATION FUNCTION



MAX POOLING



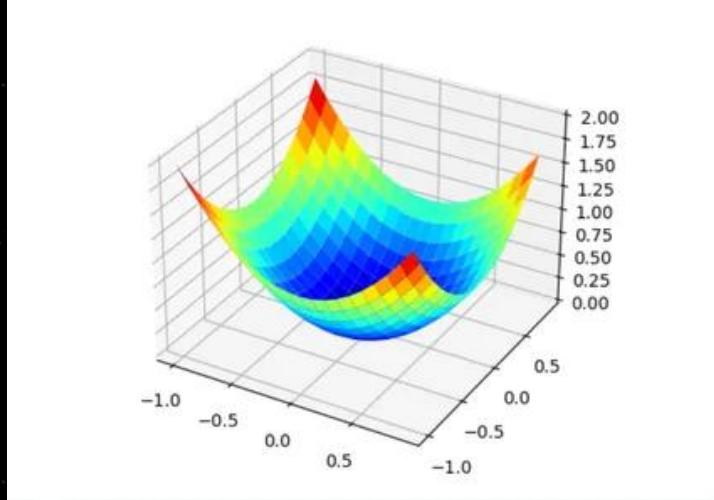
DROPOUT



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MODEL TRAINING AND TESTING



CROSS ENTROPY LOSS

Applies softmax and compute negative log-likelihood, ideal for multiclass classification

ADAM

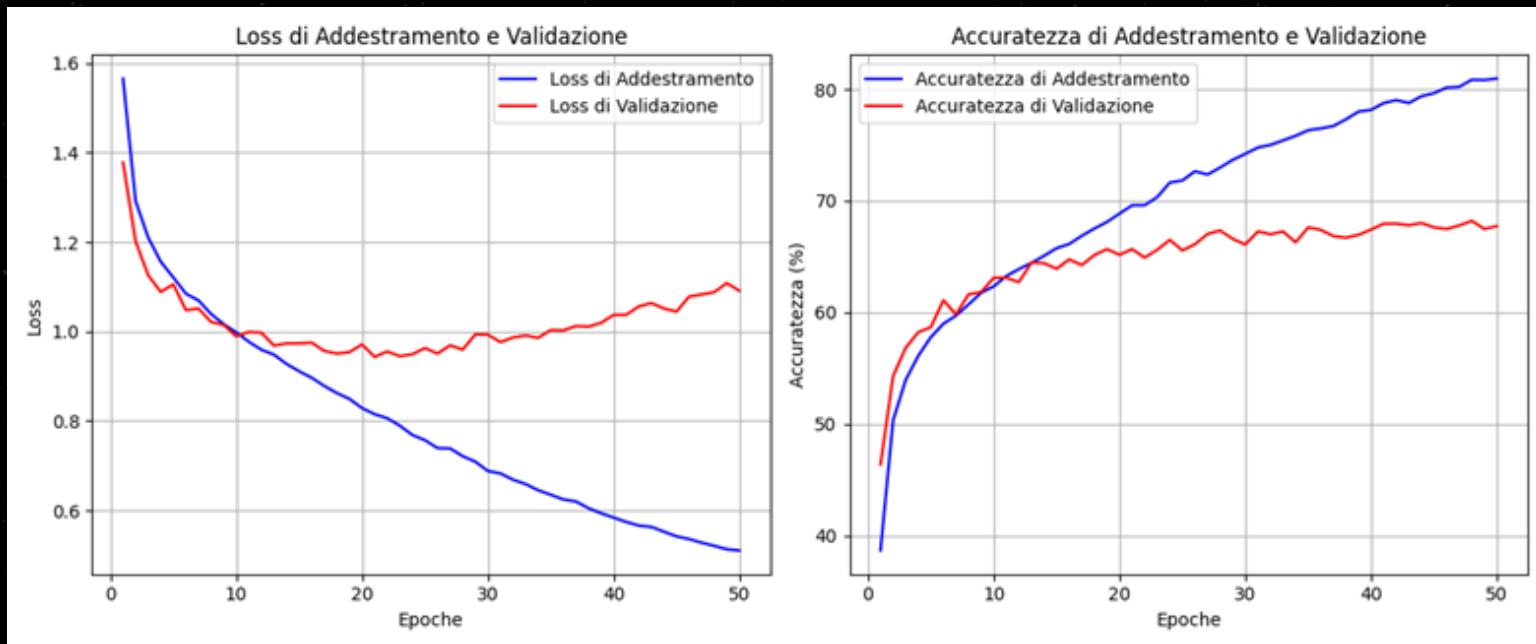
Standard optimizer, used with a 0.001 learning rate

DATASET

Is splitted into train_dataset, val_dataset and test_dataset



ACCURACY AND LOSS PLOTS



03

GENUINNESS DETECTION

How can we detect whether a facial expression is genuine?

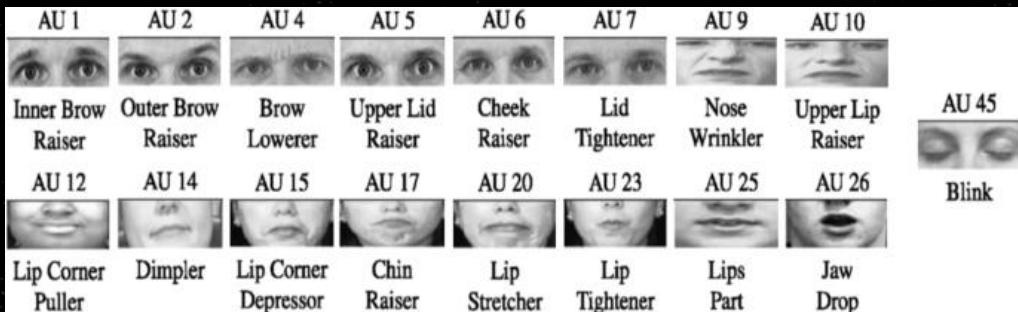
...By observing involuntary facial microexpressions.



ACTION UNITS

Action Units are **movements of facial muscles**, as defined in the **Facial Action Coding System (FACS)**, each AU corresponds to a specific muscle contraction.

Some AUs (like AU6 in a genuine smile) are **hard to fake**, making them useful for detecting true emotions.

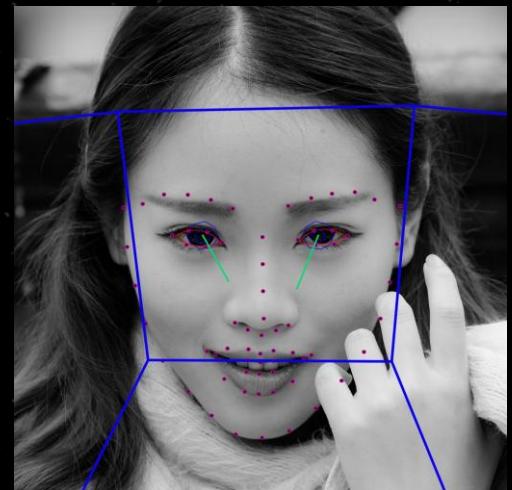
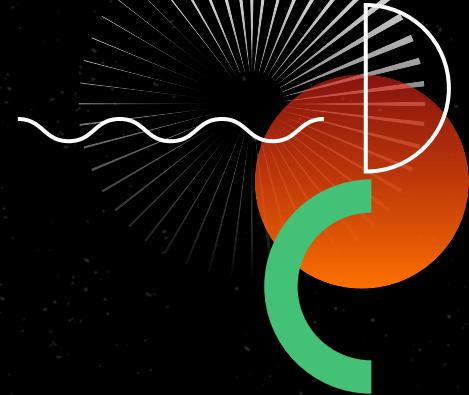


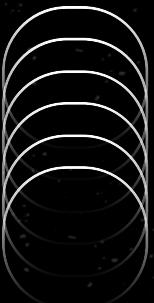
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OPEN FACE

OpenFace is an open-source facial behavior analysis toolkit. Although it offers multiple functionalities, we used it solely to extract Action Units (AUs) from images.

It detects both the presence and intensity of each AU, however, we retained only the intensity values for our analysis.





BUILD THE DATASET

No suitable public dataset was available, so we created a custom one

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STEP 01

Identify a dataset with posed facial expressions



STEP 02

Identify a dataset with genuine facial expressions



STEP 04

Combine the extracted data and assign the corresponding labels



STEP 03

Use OpenFace to extract facial Action Units



THE MODEL (MLP)

2 HIDDEN LAYER

Followed by: Batch Normalization, Leaky ReLU, Dropout

BCE WITH LOGITS LOSS

Sigmoid activation + binary cross-entropy loss

STEP LR

Gradually decrease the learning rate over time to improve convergence

1 OUTPUT LAYER

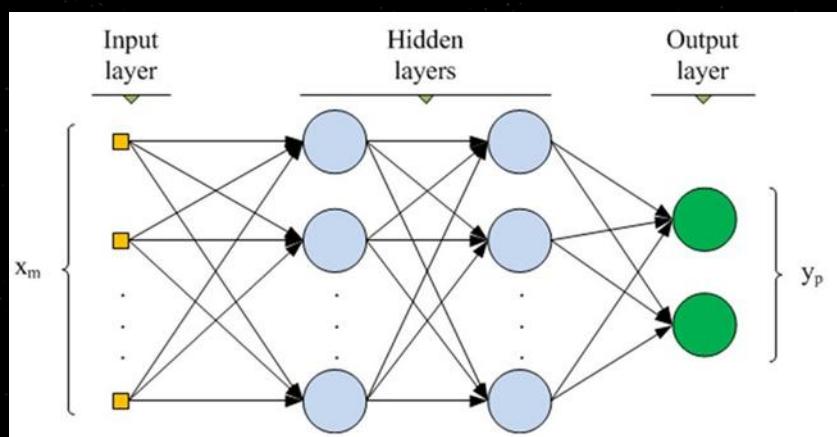
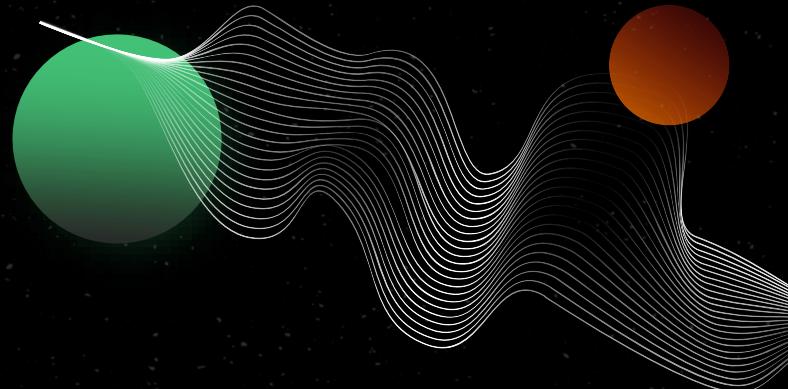
The output is a logit for binary classification

ADAM OPTIMIZER

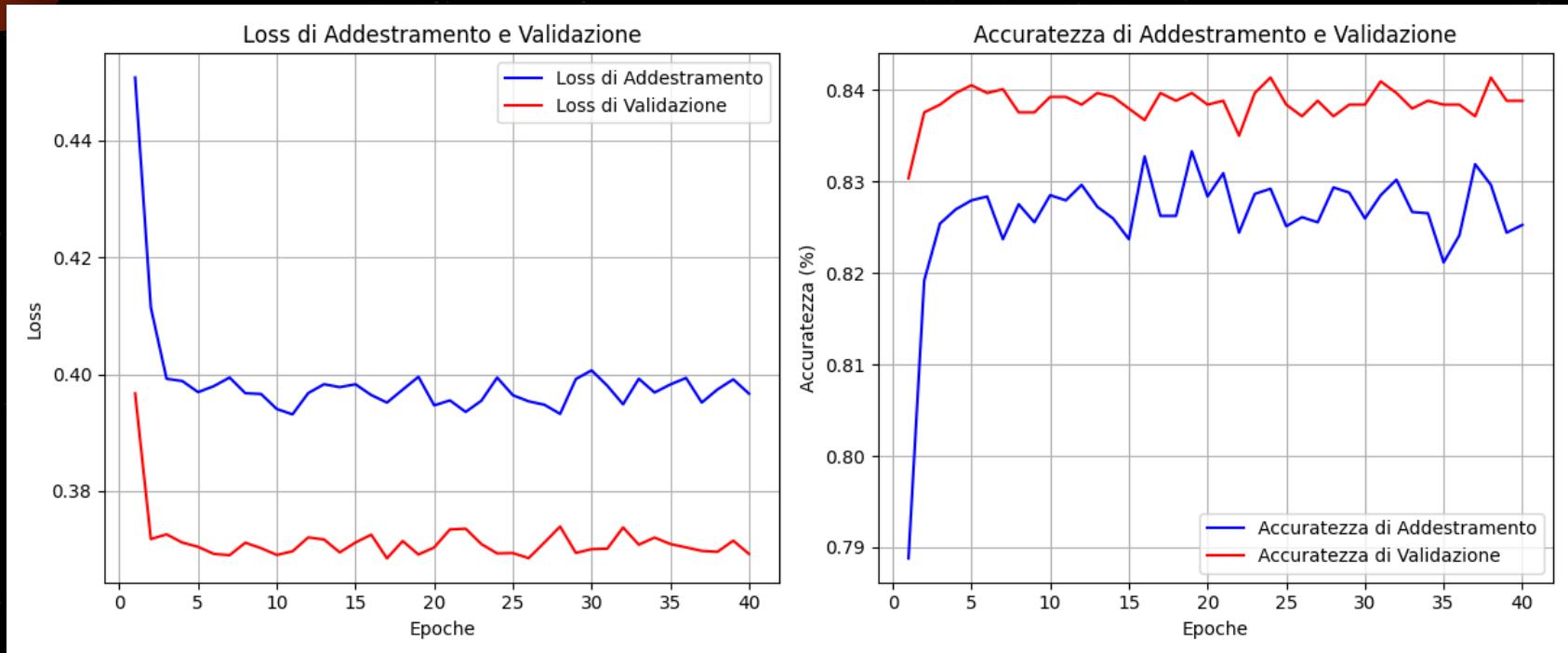
With 0.0005 learning rate, plus a L2 penalty

CROSS VALIDATION

Used 4-fold validation to improve validation robustness



ACCURACY AND LOSS PLOTS



MODEL BIAS

Although the model achieved high accuracy, its performance is not consistently reliable due to discrepancies between the two datasets used.



Despite both being 224 x 224 grayscale images noticeable differences in style are evident between them

- Left image from posed
- Right image from genuine

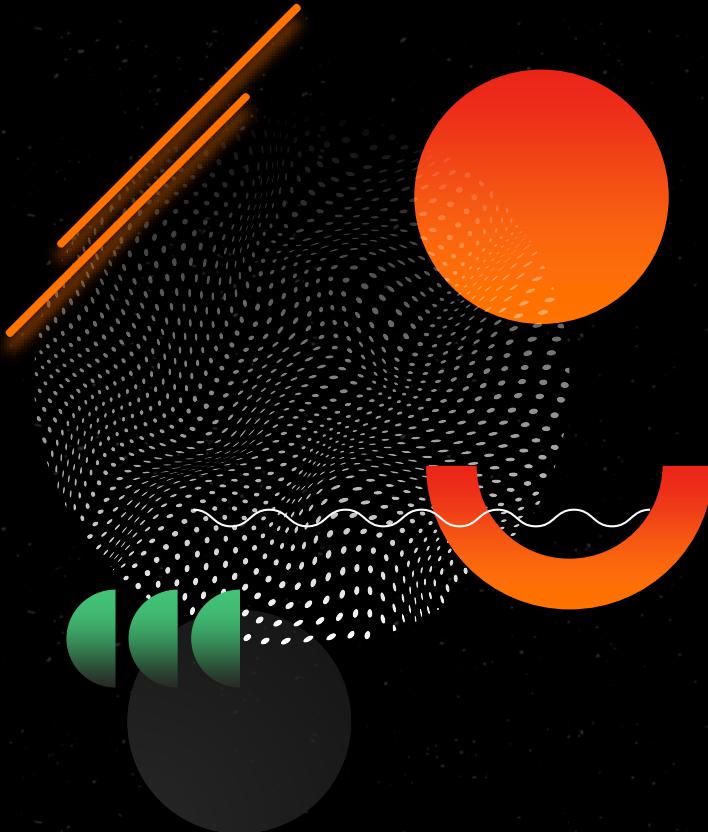


RESULTS 04

Despite some challenges along the way, we developed two different models—a CNN and an MLP—each tailored to its specific task.

When tested in real scenarios, we found that while the models aren't perfect and do make occasional mistakes, they perform well in most cases.

Overall, we are proud of the work we have accomplished



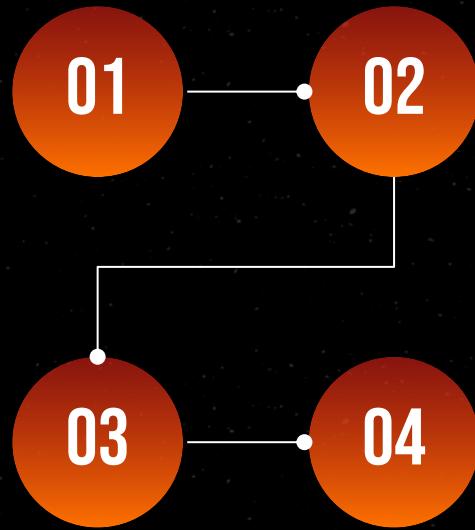
FINAL WORKFLOW

RUN THE MAIN.PY

This will load the models and activate the webcam for real-time inference

GENUINENESS DETECTION

To obtain the genuineness press «A», AUs will then be extracted and fed to the model



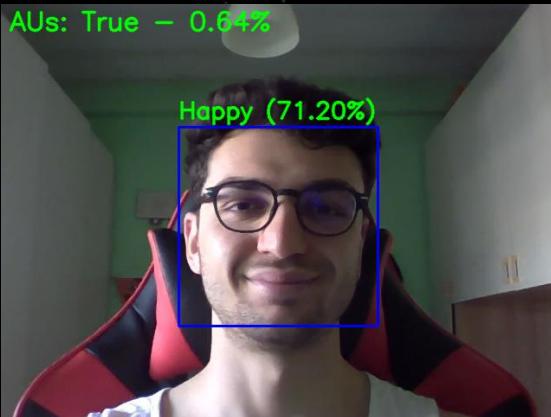
EMOTION RECOGNITION

When a face is detected, the CNN model recognizes the emotion in real time

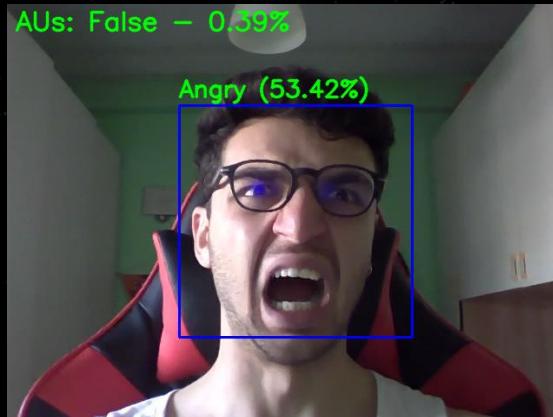
FINAL RESULT

A freeze frame will show the emotion label and a true / false value indicating genuineness

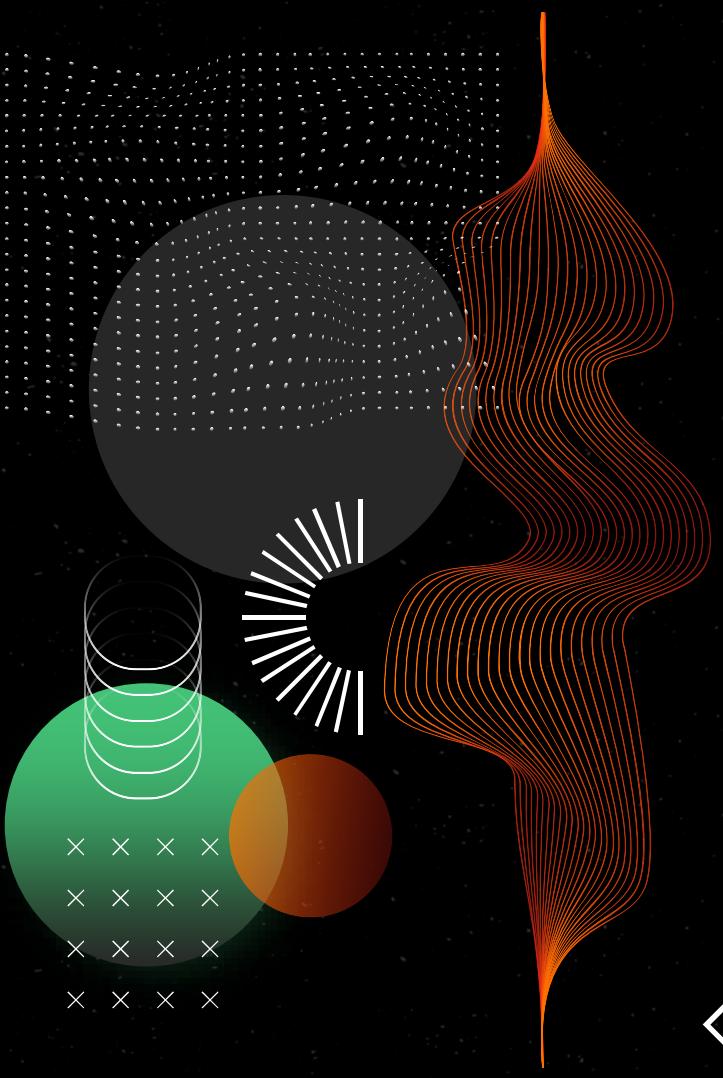
SHOW THE RESULTS



Why do you smile?



Take it easy bro..



Do you have any question
for us?



Thanks so much for listening!

Hope you enjoyed the presentation, and good luck when it's your turn!

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