



Facial Emotion Recognition in Real-Time

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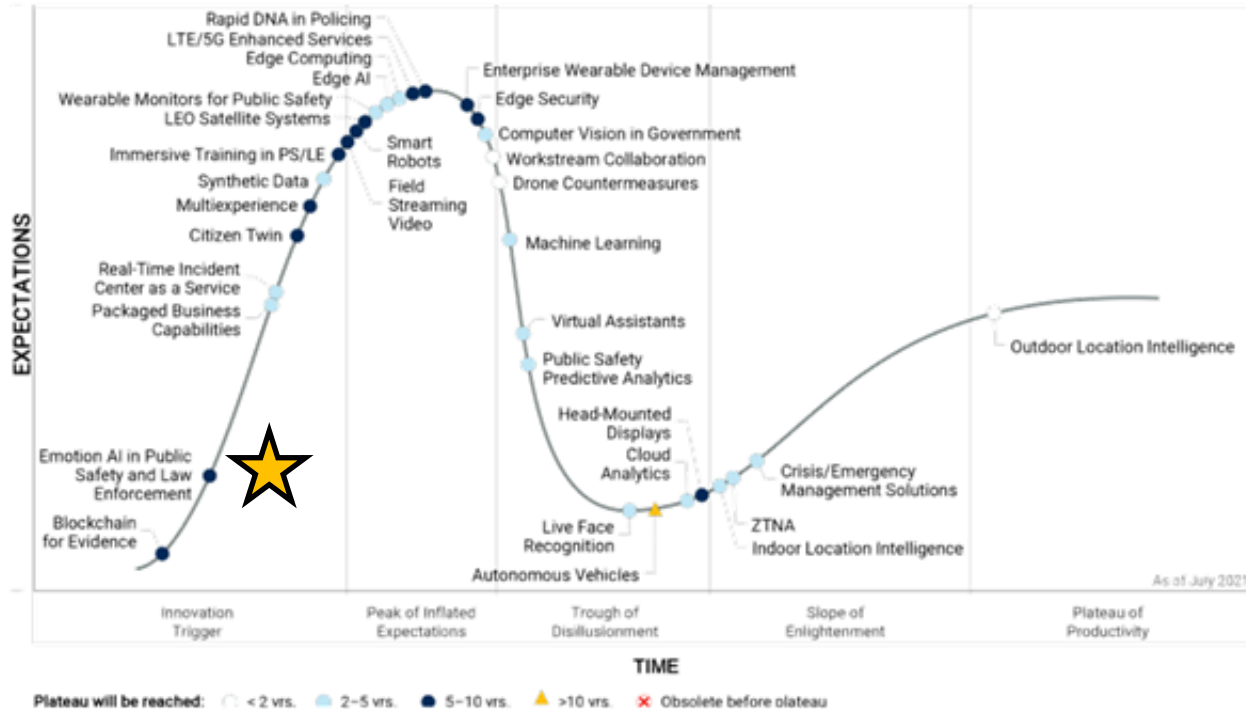


01

Context & Problem

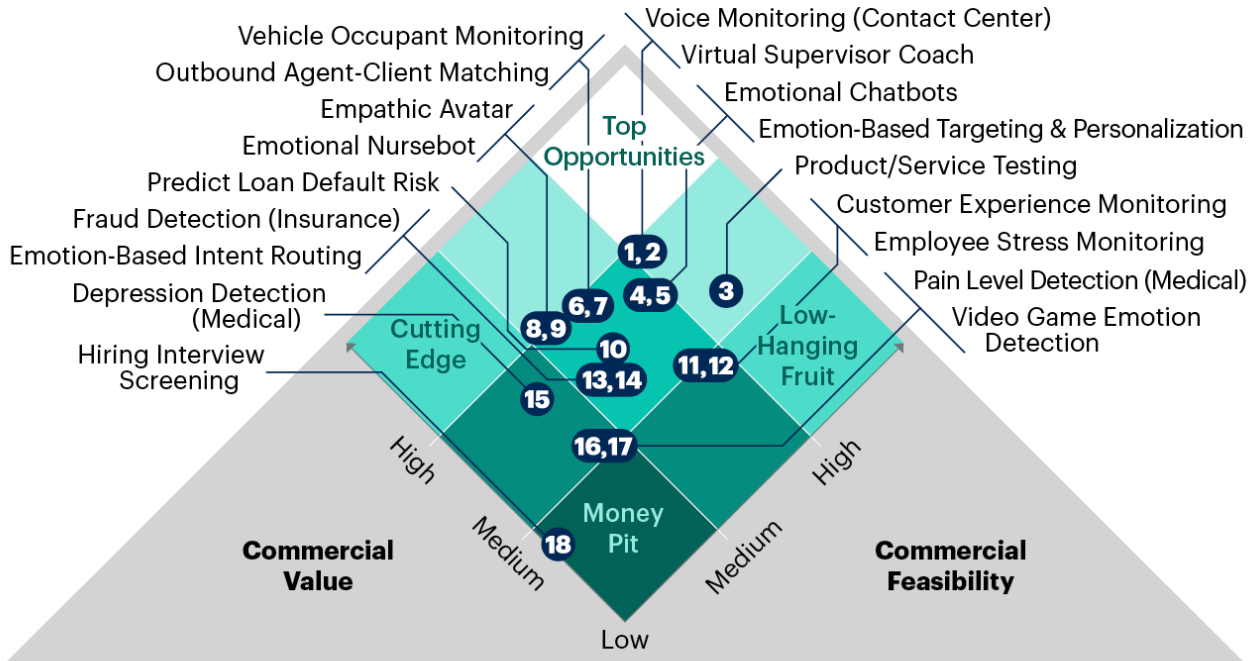
Rising Interest in/Potential of Emotion AI for Law Enforcement

Hype Cycle for Public Safety and Law Enforcement 2021



In a wider context ...

Emotion AI Technology Opportunity Prism



Source: Gartner
741689_C

Why Emotion AI for Law Enforcement?



**Human
Interaction is
Our Core Domain**



**Digital Interaction
is the Present
& Future**



**Keeping Pace
with Digital
Innovation**

Focus on Detecting Emotions Through Facial Expression

Components of Emotion AI



Proposed Solution

- **Test data architectures of pre-trained Convolutional Neural Network models**
 - Baseline model: Custom CNN model
 - Pre-trained models: ResNet-50, MobileNetV2, VGG16, VGGFace
- **Evaluation Criteria:**
 - Accuracy scores on unseen data
 - F1-scores of each emotion class on unseen data



02

Data & Insights

FER-2013 Dataset

- ~ 28k labelled images in train set and ~3.5k in test set
- Posed and unposed headshots in grayscale 48x 48 pixels
- 7 emotion classes: angry, disgust, fear, happy, neutral, sad, surprise
- Created using Google image search of each emotion and synonyms of the emotions
- Well-explored dataset in the data science community

Class Imbalance Especially with Happy and Disgust



Our Dataset is Not Perfect

Occlusions



Unrealistic



Mis-labelling?



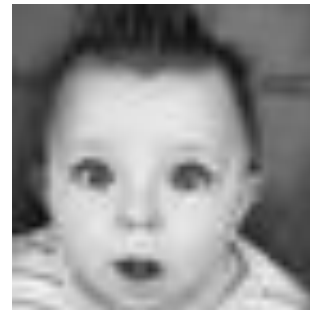
Lighting Variations



Angle Variations

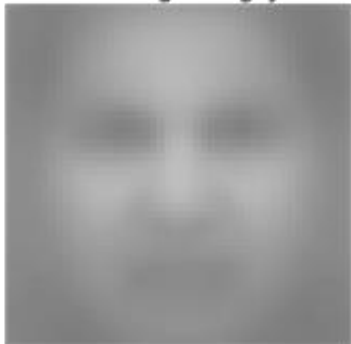


Failed to account for age, gender or cultural differences?



Average Images Distinguish Key Facial Features for Each Emotion

Average angry



Average surprise



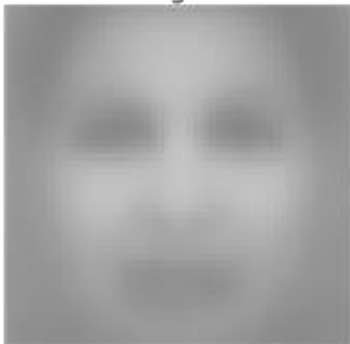
Average happy



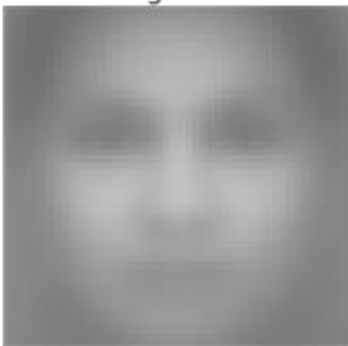
Average disgust



Average fear



Average neutral



Average sad

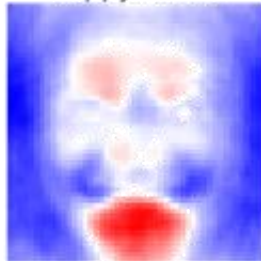


Contrast Images Further Isolate Key Facial Features to Differentiate Emotions

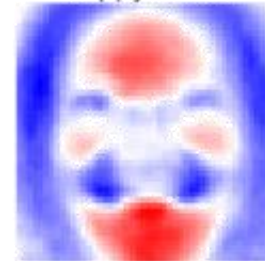
Happy v Angry



Happy v Fear



Happy v Sad



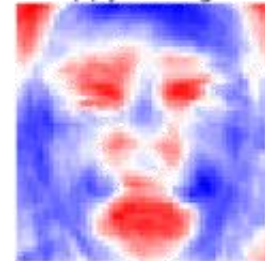
Happy v Surprise



Happy v Neutral



Happy v Disgust

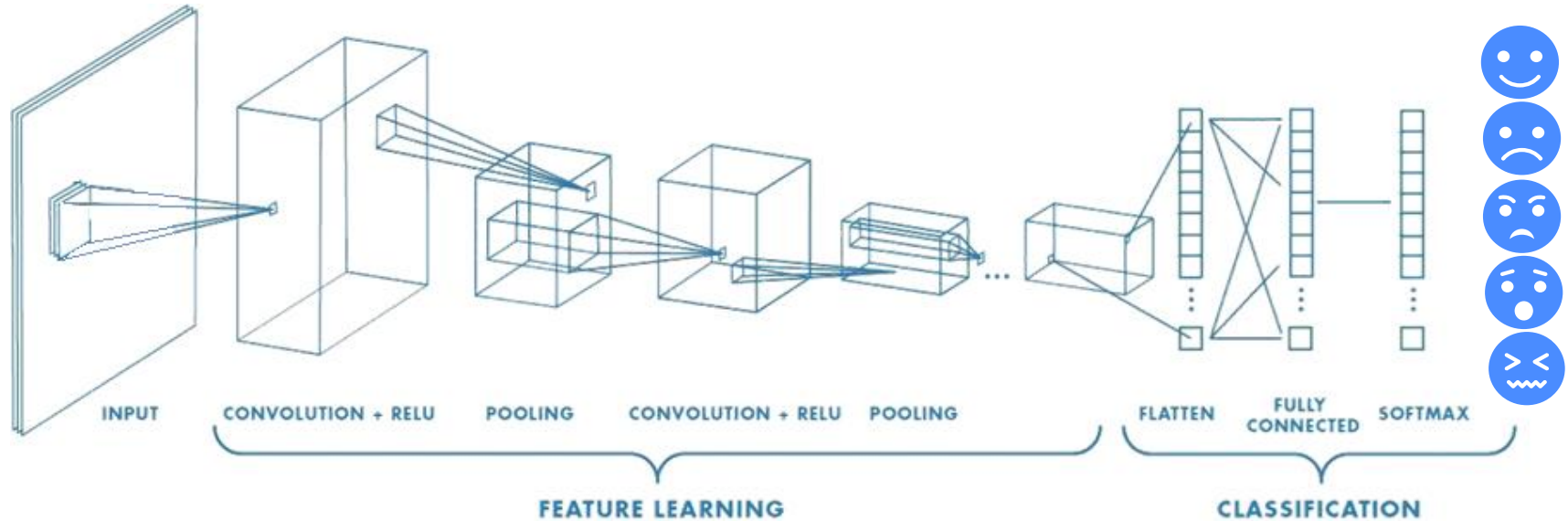




03

**Model
Workflow &
Evaluation**

Structure of Convolutional Neural Network



Source: <https://towardsdatascience.com/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way-3bd2b1164a53>

Baseline Model

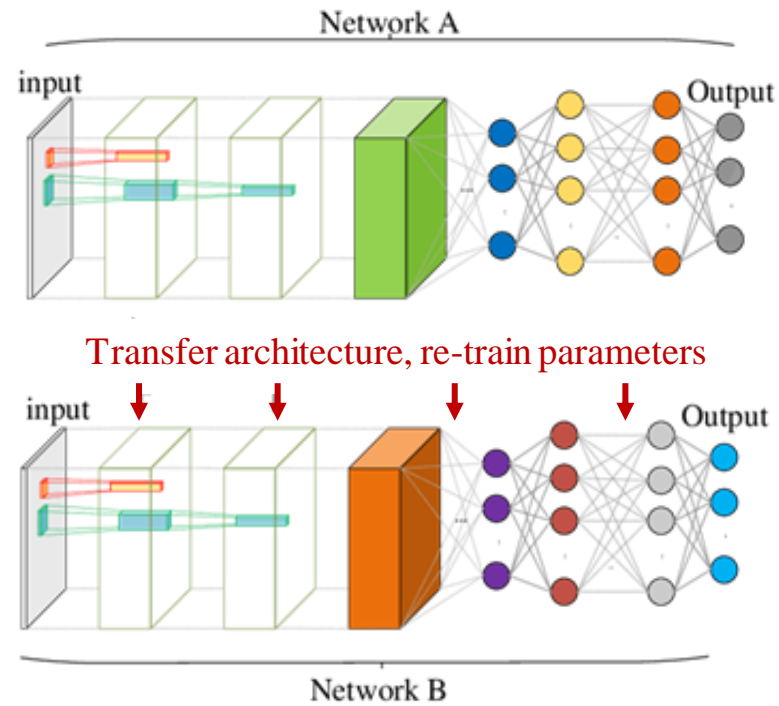
- 4 Convolution w Max-Pooling Layers
- 2 Fully Connected Layers

Layer (type)	Output Shape	Param #
conv_layer_1 (Conv2D)	(None, 46, 46, 128)	3584
batch_normalization (Batch Normalization)	(None, 46, 46, 128)	512
pool_layer_1 (MaxPooling2D)	(None, 23, 23, 128)	0
dropout (Dropout)	(None, 23, 23, 128)	0
conv_layer_2 (Conv2D)	(None, 21, 21, 256)	295168
batch_normalization_1 (Batch Normalization)	(None, 21, 21, 256)	1024
pool_layer_2 (MaxPooling2D)	(None, 10, 10, 256)	0
dropout_1 (Dropout)	(None, 10, 10, 256)	0
conv_layer_3 (Conv2D)	(None, 8, 8, 512)	1180160
batch_normalization_2 (Batch Normalization)	(None, 8, 8, 512)	2048
pool_layer_3 (MaxPooling2D)	(None, 4, 4, 512)	0
dropout_2 (Dropout)	(None, 4, 4, 512)	0
conv_layer_4 (Conv2D)	(None, 2, 2, 512)	2359808
batch_normalization_3 (Batch Normalization)	(None, 2, 2, 512)	2048
pool_layer_4 (MaxPooling2D)	(None, 1, 1, 512)	0
dropout_3 (Dropout)	(None, 1, 1, 512)	0
flatten (Flatten)	(None, 512)	0
dense (Dense)	(None, 512)	262656
dropout_4 (Dropout)	(None, 512)	0
dense_1 (Dense)	(None, 256)	131328
dropout_5 (Dropout)	(None, 256)	0
dense_2 (Dense)	(None, 5)	1285
=====		
Total params: 4,239,621		
Trainable params: 4,236,805		
Non-trainable params: 2,816		

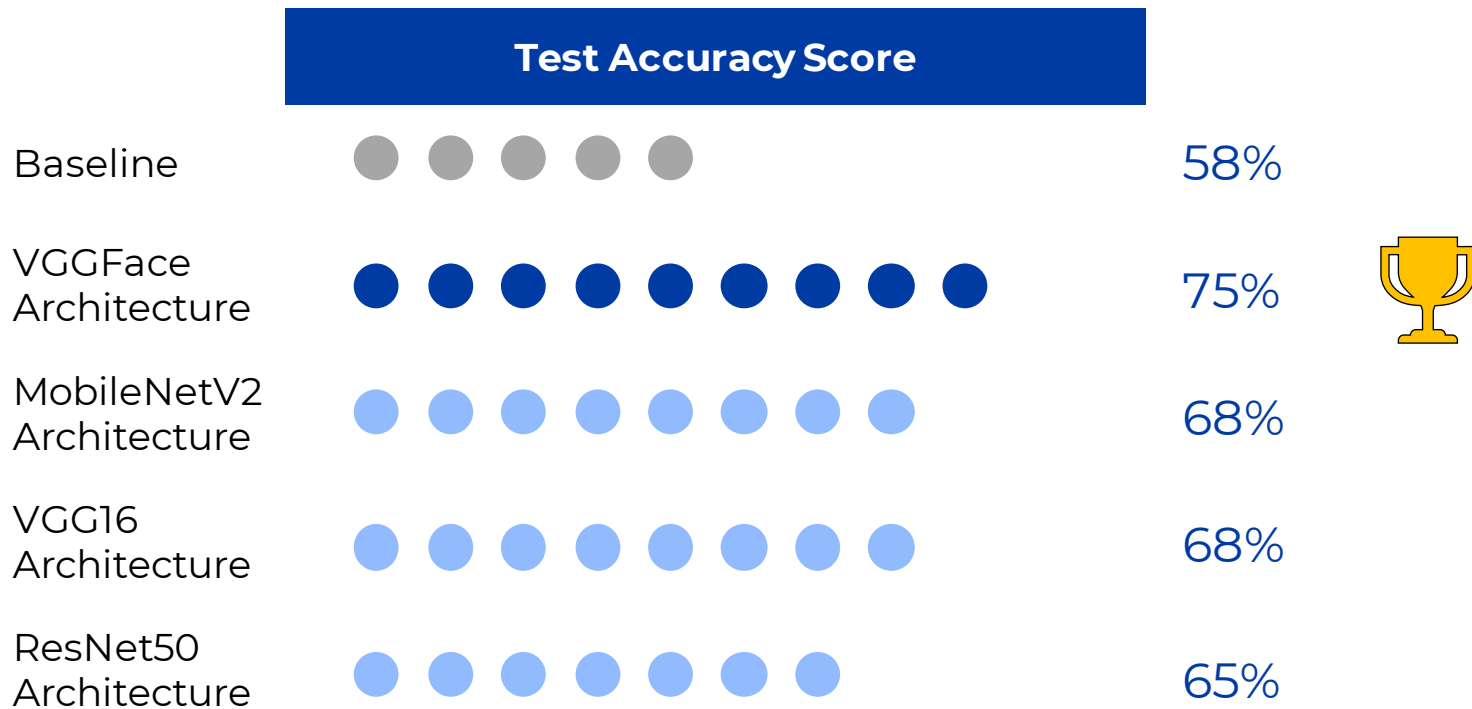
If you know how to ride a bicycle, it will be easier to learn to ride a bike

4 Pre-Trained Architectures

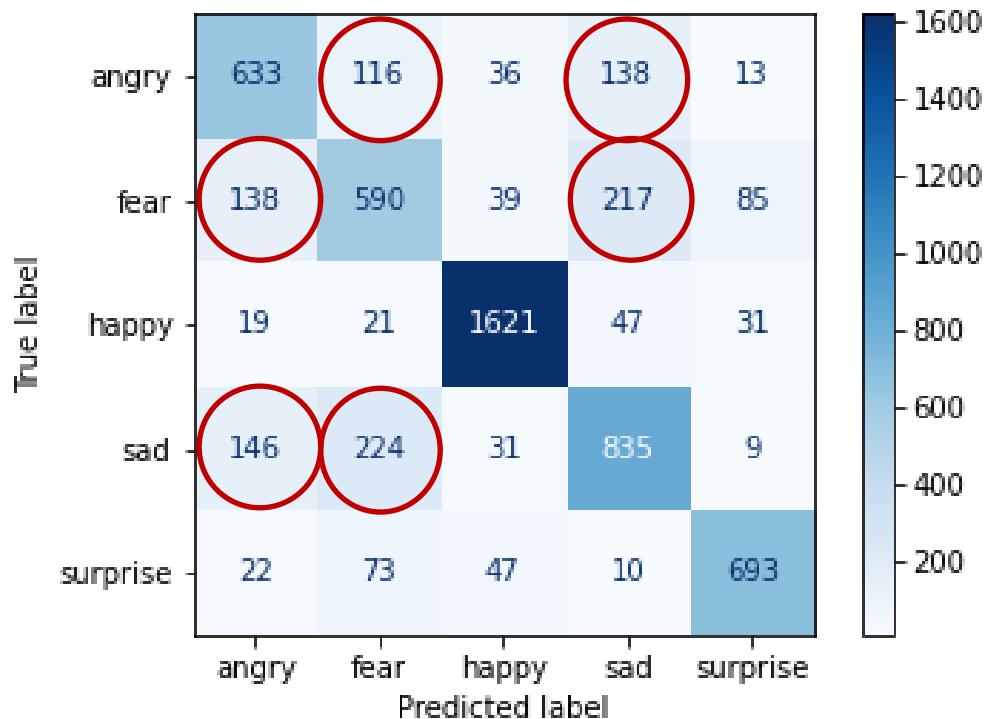
- ResNet-50
- MobileNetV2
- VGG16
- VGGFace



Model Evaluation



Our model is very good at classifying happy, but tend to confuse angry, fear and sadness



Comparison of F1-Scores

	VGGFace	MobileNetV2	VGG16	ResNet-50
angry	0.67	0.57	0.54	0.52
fear	0.56	0.40	0.51	0.47
happy	0.92	0.89	0.87	0.82
sad	0.67	0.62	0.61	0.56
surprise	0.83	0.77	0.78	0.77

Actual Label: Sad



Predictions

VGGFace

Sad

VG166

Happy

MobileNetV2

Angry

ResNet-50

Angry

Actual Label: Angry



Predictions

VGGFace

Angry

VG166

Angry

MobileNetV2

Sad

ResNet-50

Sad

Actual Label: Surprise



Predictions

VGGFace	Fear
VG166	Fear
MobileNetV2	Surprise
ResNet-50	Surprise

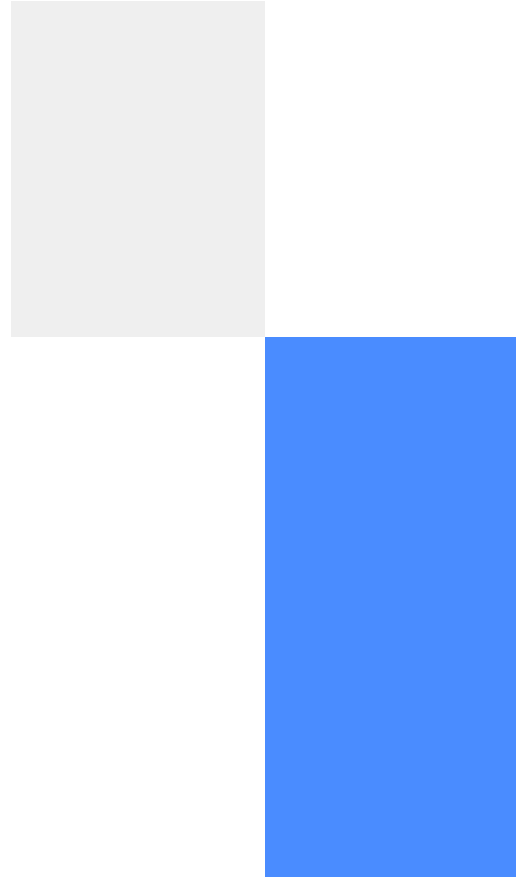
Demo:

Model Deployment

Using OpenCV

04

Conclusion & Next Steps



Conclusion



**Quality of labelled
data matters (a lot!)**



**Cultural differences
& bias matter too**



**Tech adoption will
be extra tough for a
traditional sector**

Next Steps



1. To explore emotion recognition with facial landmark detection using Computer Vision libraries e.g. dlib

- Possibly, combining with Paul Ekman's theory on micro expressions



2. To acquire 'culturally relevant' datasets and re-run modelling steps



3. To explore prospects of emotion detection using audio and text analysis too

Thank you

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