

Specialist Diploma in Artificial Intelligence

Al Applications with Deep Learning



Build Your Own Face Recognition Model



Sub topics

- Facial Tasks
- How It Works
- Facial Detection Algorithms
- Face Embedding Algorithms
- Classification Algorithms
- Similarity Measures
- Putting It Altogether

Facial Recognition

FACIAL RECOGNITION TASKS



Tasks

• Facial tasks:

person. Example use cases:

- Facial Verification
 Given an image of two faces, tell if the two faces belong to the same
 - Airport security check using passport and a photograph of your face
 - Logging into bank ATM terminals with card + face as a second factor. https://www.bbc.com/news/business-54266602



Tasks

- Facial tasks:
 - Facial Recognition

Given an image of a face, identify the person. Example use cases:

- Scanning photographs of people in a train station and identifying who they are
- Scanning photographs in photo gallery (in mobile phone) to tag the people in the photo



China Tech: 5 ways China is using Facial Recognition



https://www.youtube.com/watch?v=lb5Nv-l2Z4g

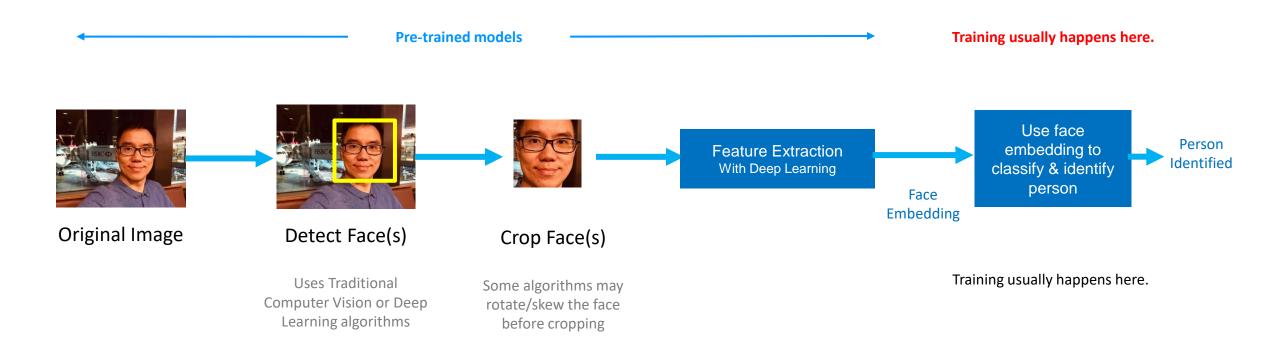
Facial Recognition

HOW IT WORKS



How it Works – Facial Recognition

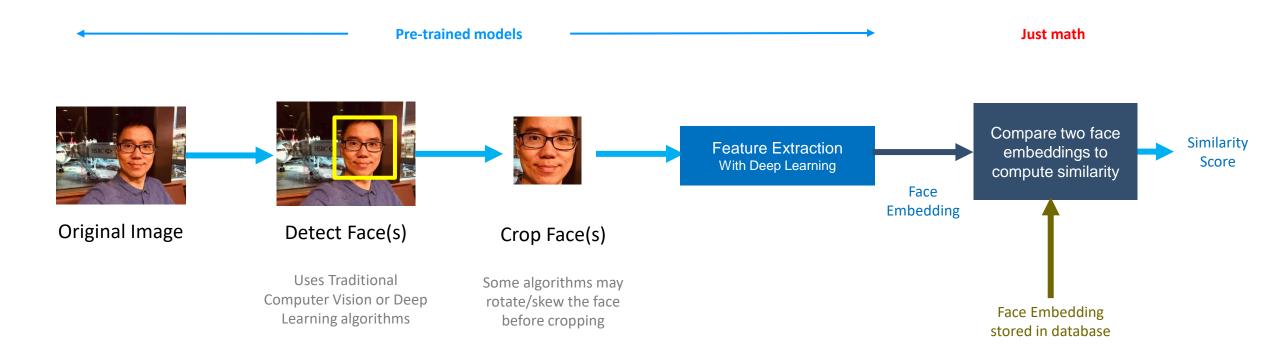
A typical facial recognition works in the following way:





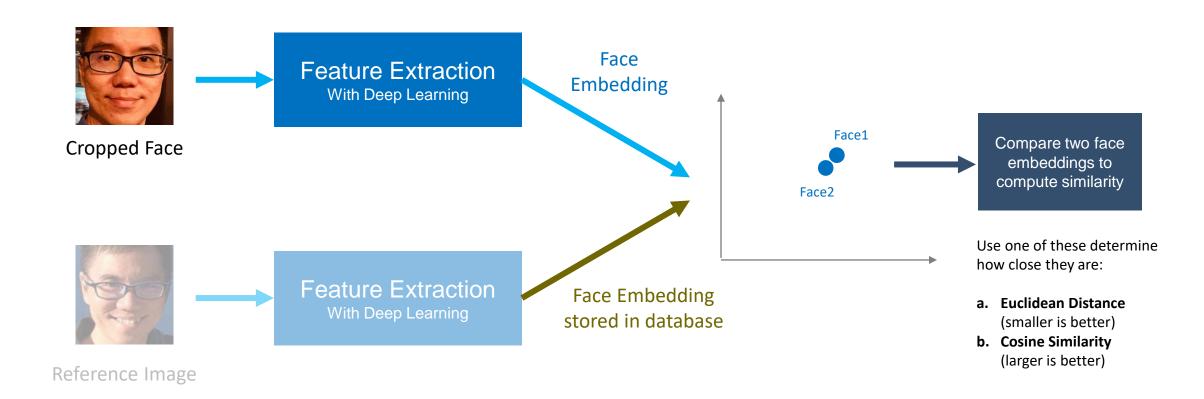
How it Works – Facial Verification

A typical facial recognition works in the following way:





Face Embedding



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Face Embedding

 Like word embeddings, a face embedding is a n-dimensional vector representation of a person's face in an image.

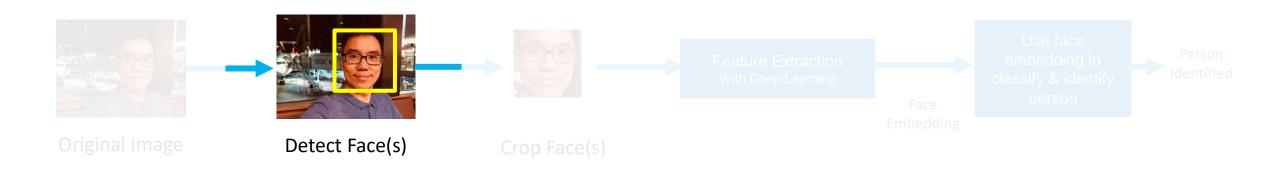
- Imagine the vector as a "point" in n-dimensions so that:
 - 2 photographs of the same person will produce two "points" that are very "near" in the n-dimensional space
 - 2 photographs of a different person will produce two "points" that are very "far" in the n-dimensional space

Facial Recognition

FACE DETECTION ALGORITHMS



Facial Detection



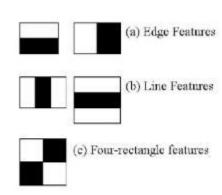
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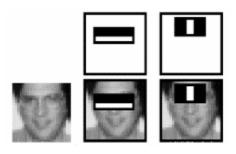
Uses Traditional Computer Vision or Deep Learning algorithms



Haar Cascades

- Haar Cascades is an effective object detection ML algorithm based dark / light areas of an image.
- Haar features are like convolutional filters (light area is -1, dark area is 1) multiplied onto areas of an image
- Uses a cascade of Haar features for classification
- Fastest, least accurate

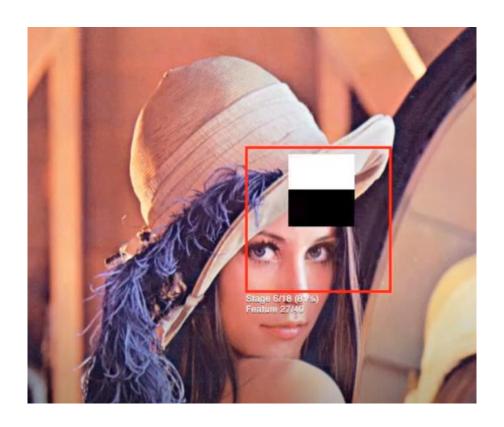




https://en.wikipedia.org/wiki/Viola%E2%80%93Jones_object_detection_framework



Visualizing Haar Cascades



https://www.youtube.com/watch?v=hPCTwxF0qf4



DLib

- Uses Histogram of Oriented Gradients (HoG).
- Divides the image into small square boxes and computes the histogram of edges in different directions.
- Use SVM to classify face / not face.

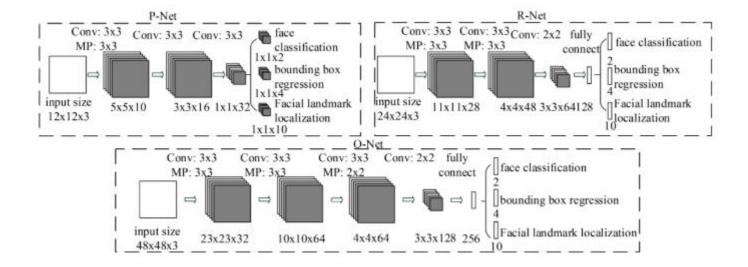


https://sklin93.github.io/hog.html



MTCNN

Multi-Task Cascading Convolution Networks (state-of-the-art)



https://arxiv.org/abs/1604.02878

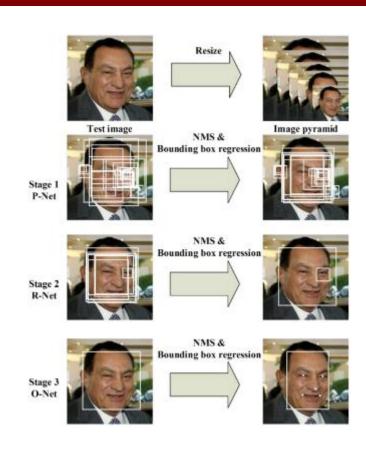
https://towardsdatascience.com/how-does-a-face-detection-program-work-using-neural-networks-

17896df8e6ff



MTCNN

- Utilizes 3-stage cascading CNNs to determine face position and facial landmarks (eyes, nose, mouth)
- Most accurate, slowest



https://arxiv.org/abs/1604.02878

https://towardsdatascience.com/how-does-a-face-detection-program-work-using-neural-networks-17896df8e6ff



Comparison Between Haar and MTCNN

- Based on this author:
 - https://datawow.io/blogs/face-detection-haar-cascade-vs-mtcnn
- Benchmark Dataset: UTK Face (24,111 faces)

	Precision	Recall	Time
Haar Cascade	95.2%	82.6%	25 images / sec (CPU only)
MTCNN	98.0%	89.9%	3 images / sec (CPU only)



Comparison Between Haar, HoG, MTCNN



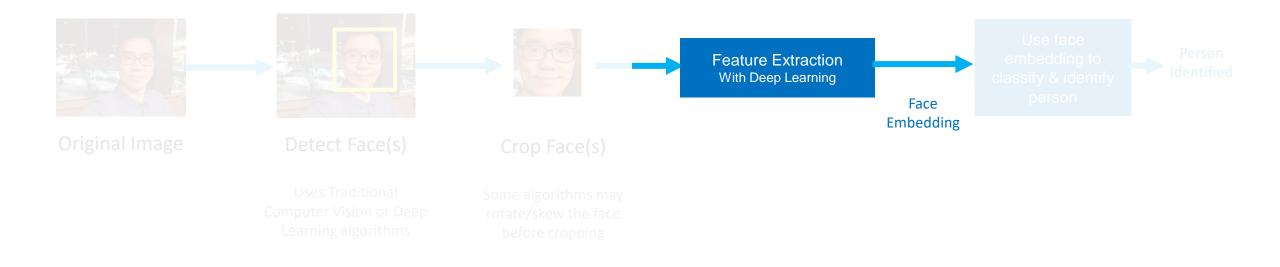
https://www.youtube.com/watch?v=GZ2p2hj2H5k

Facial Recognition

FACE EMBEDDING NETWORKS



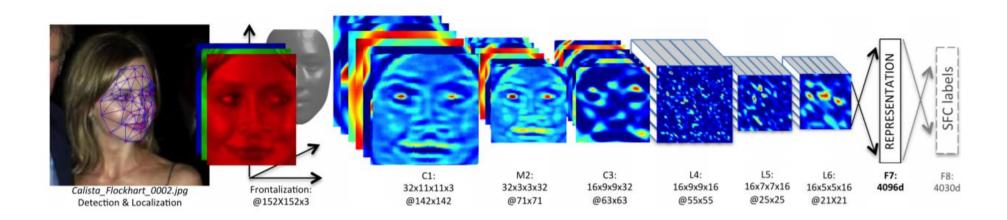
Facial Detection





DeepFace

- Published by Facebook in 2014
- Improves accuracy by using 3D facial modelling to 'align' face before recognition





FaceNet

- Published by Google 2015,
- FaceNet uses a Deep Convolutional Neural Network

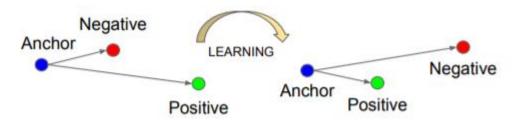


- The inputs are batches of cropped faces
- The "Deep Architecture" is Inception / Zeilur & Fergus architecture
- During training, FaceNet uses Triplet Loss
- During normal use, FaceNet produces the Face Embedding of an image of a person's face



FaceNet Training with Triplet Loss

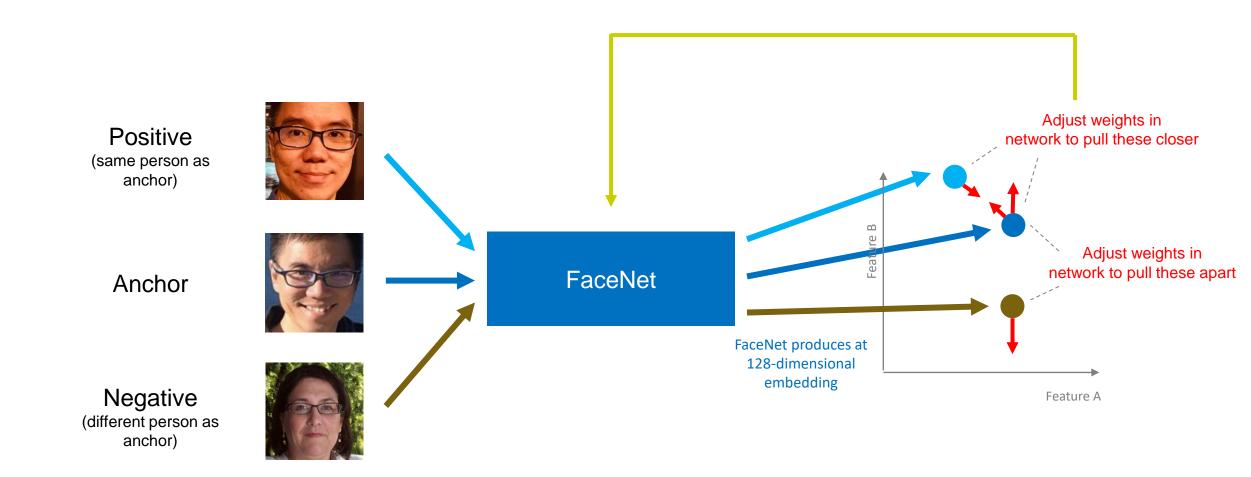
• Triplet Loss:



- For each training, randomly pick 3 faces:
 - 1 anchor
 - 1 positive example (another image of the same person in anchor image)
 - 1 negative example (another image of the different person in anchor image)
- Objective of Triplet Loss:
 - Maximize distance between (anchor to negative example)
 - Minimize distance between (anchor to positive example)



FaceNet



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Comparison Between DeepFace and FaceNet

- Read comparison:
 - https://projectsflix.com/machine-learning/deeplearning/deepface-and-facenet-for-face-recognition/
 - Based on the papers provided in the previous slides

	Accuracy on LFW Dataset	Accuracy on YTF Dataset	Embedding
DeepFace	97.2%	91.4%	4096-dimensions
FaceNet	99.6%	95.1%	128-dimensions



Other Materials

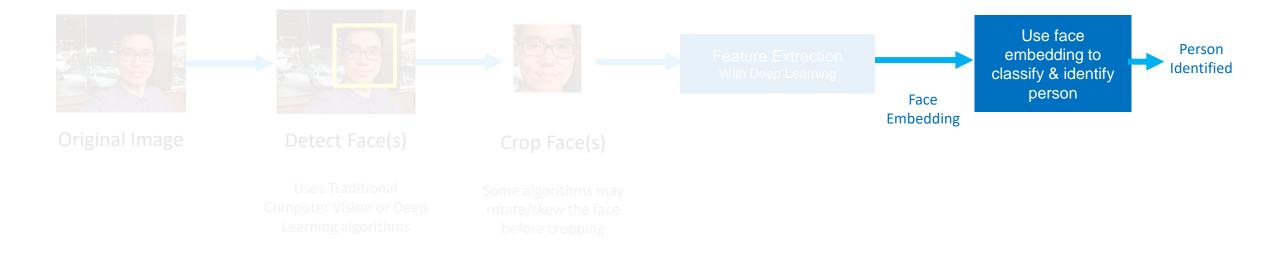
- Presentation of FaceNet's Triplet Loss by Andrew Ng:
 - https://www.youtube.com/watch?v=d2XB5-tuCWU

Facial Recognition

CLASSIFICATION ALGORITHMS



Facial Detection



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Classification

- A classification algorithm is required when you want to perform:
 - Facial Recognition
- The output from the deep learning network is a set of n-dimensional face embedding.

 Since it's a fixed set of output of n numbers, you can use any classification algorithm available in Scikit-Learn

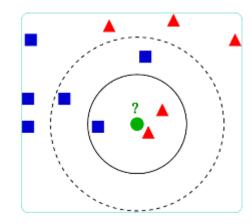


K-NN Classification

- *k*-Nearest Neighbours:
 - Look for all indexed face embeddings in your database
 - Find the k nearest indexed face embeddings (points)
 matching the new one that you've just obtained from your image

NOTE: Set k = 1 to classify based on nearest point

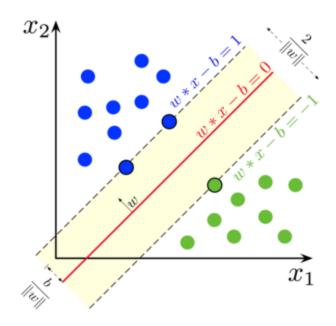
- One-shot
 One photo per face; no training happens
- More data = slower inference





SVM Classification

- Support Vector Machine:
 - Divides the face embedding space into one or more linear / radial hyperplanes
- Good with small number of data
- Training speed slow with more data
- Faster inference compared to kNN

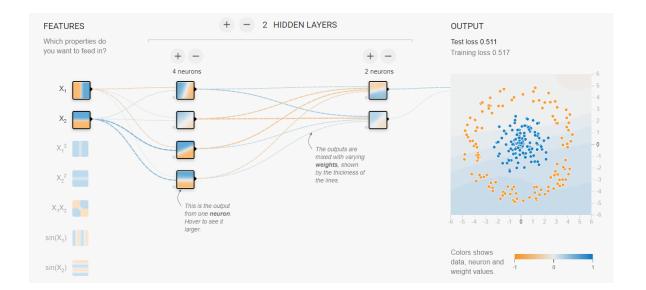


https://en.wikipedia.org/wiki/Support-vector_machine



Neural Networks

- Artificial Neural Networks:
 - Using neurons to classify face embeddings
- May require more than 1 photo per face for training

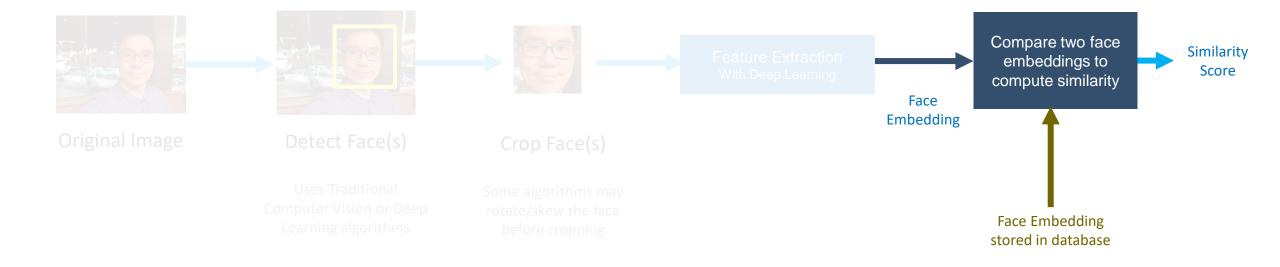


Facial Recognition

SIMILARITY MEASURES



How it Works – Facial Verification





Similarity Measure

- A similarity measure is required when you want to perform:
 - Facial verification
- Useful if you have a reference image of a person, and the live image of another person, and you want to compare to see if both are the same person.

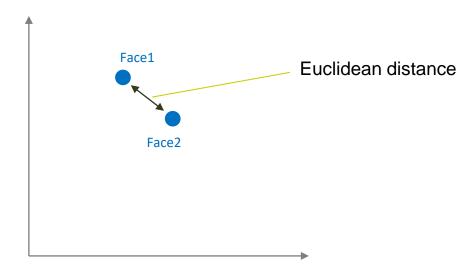
 If the similarity of the face embedding of both images are within a threshold (decided by you) => both faces belong to same person



Euclidean Distance

- Distance between 2 points, p and q
- Computed with this formula: $d(\mathbf{p}, \mathbf{q}) = \sqrt{\sum_{i=1}^{n} (q_i p_i)^2}$

• Smaller value => more similar

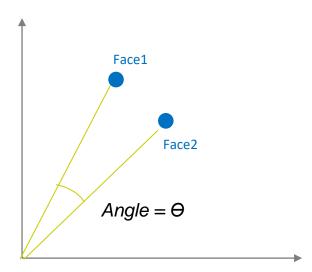




Cosine Similarity

- The cosine of the angle between 2 points, A and B from (0, 0)
- Computed with this formula: $\sin(\theta) = \frac{\mathbf{A} \cdot \mathbf{B}}{\|\mathbf{A}\| \|\mathbf{B}\|} = \frac{\sum_{i=1}^{n} A_i B_i}{\sqrt{\sum_{i=1}^{n} A_i^2} \sqrt{\sum_{i=1}^{n} B_i^2}}$

• Larger value => more similar



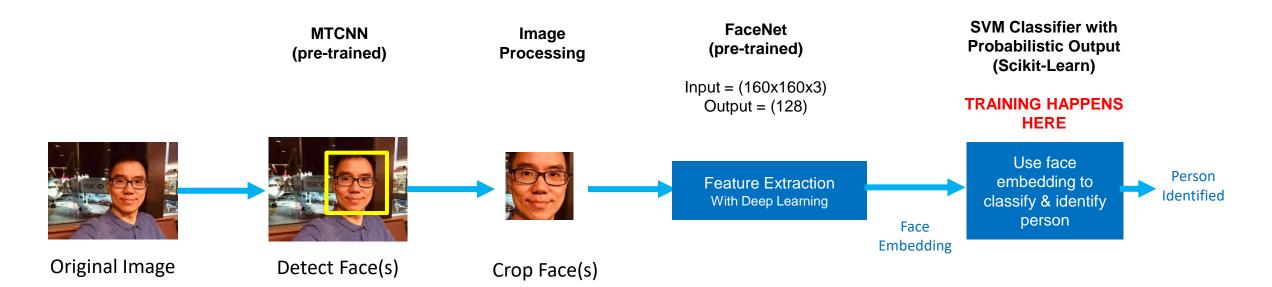
Facial Recognition

PUTTING IT ALTOGETHER



Putting It Together – Facial Recognition

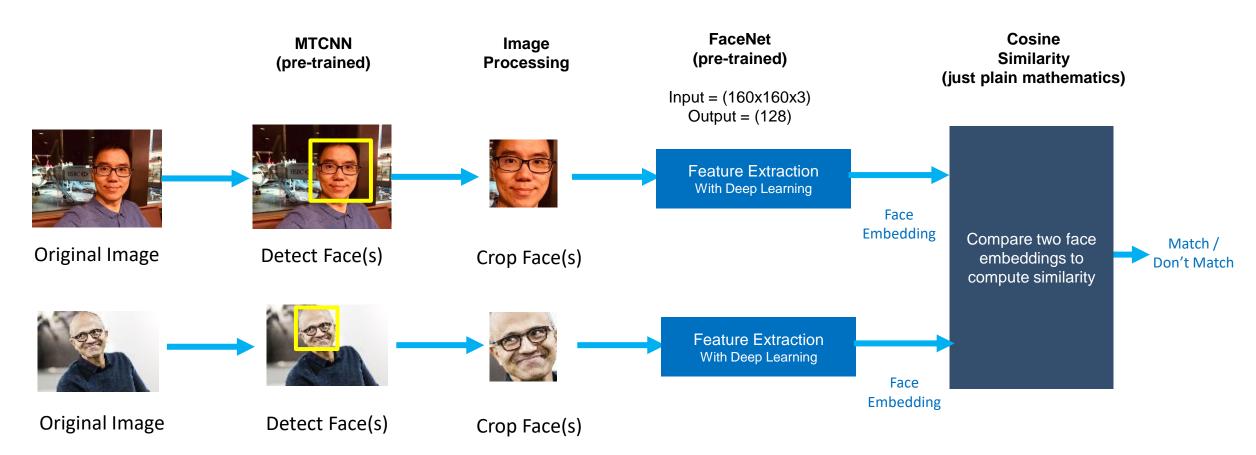
 To put together a Facial Recognition engine, we must decide on the libraries at each step. For our practical, this is what we will use:





Putting It Together – Facial Verification

• To put together a Facial Verification engine, this is what we will use:





Putting It Together

- The libraries we use:
 - Pre-Trained MTCNN for Facial Detection https://github.com/ipazc/mtcnn
 - OpenCV for image loading, resizing, and other operations
 - Pre-Trained FaceNethttps://github.com/nyoki-mtl/keras-facenet
 - Scikit-Learn for SVM Classifier
 - Numpy for mathematical computations



Summary

- Facial Tasks
- How It Works
- Facial Detection Algorithms
- Face Embedding Algorithms
- Classification Algorithms
- Similarity Measures
- Putting It Altogether