

Face Shape Classification

using Convolutional Neural Network

DSI-16 Capstone Project

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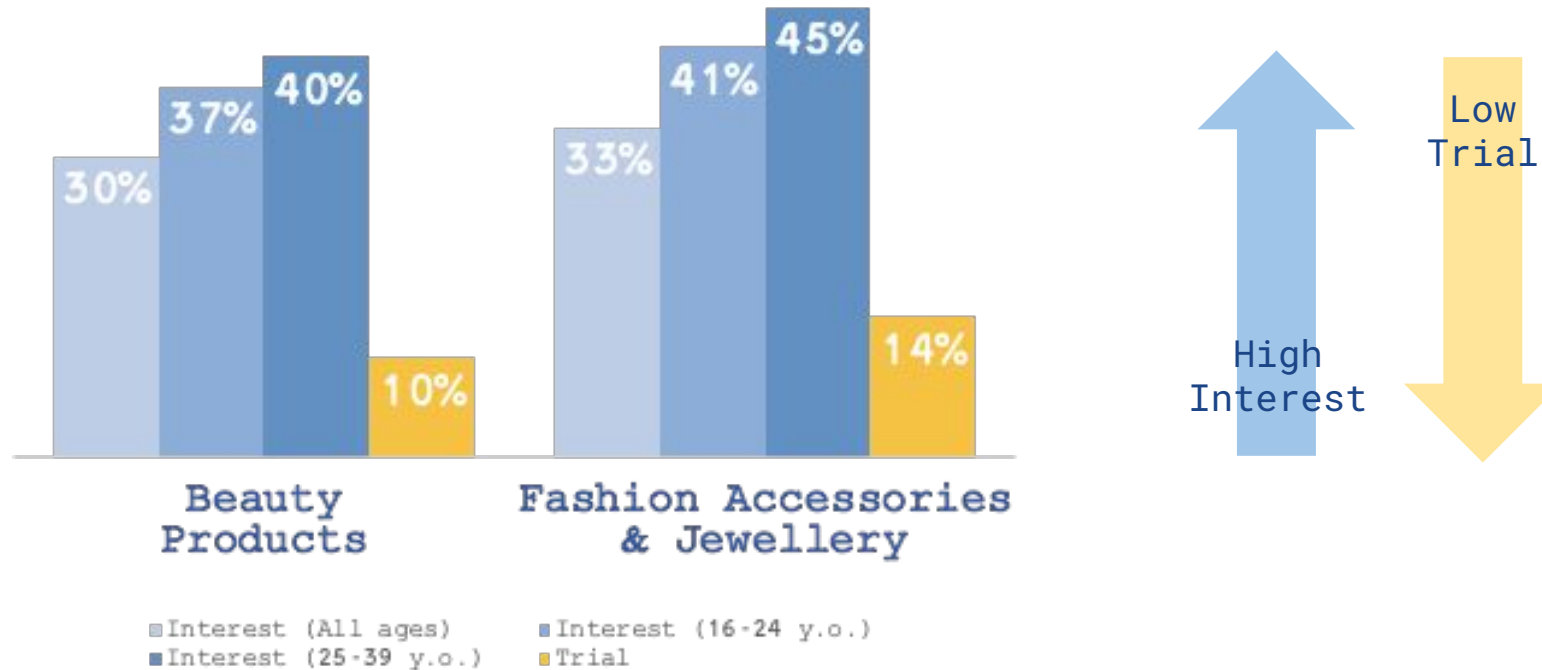
Agenda

- **Problem Statement**
- **Project Approach**
- **Data Exploration & Pre-processing**
- **Modelling & Evaluation**
 - CNN built from scratch
 - CNN with transfer learning (VGG-Face)
- **Conclusions**
- **Next Steps**

Problem Statement:

To enable personalization in beauty & fashion

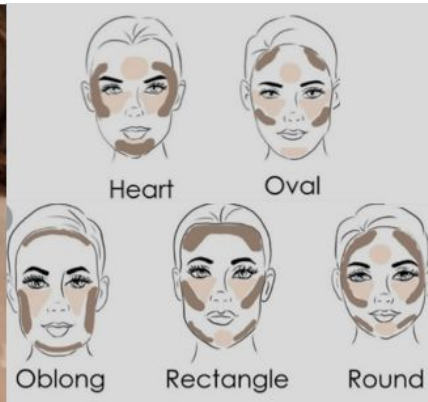
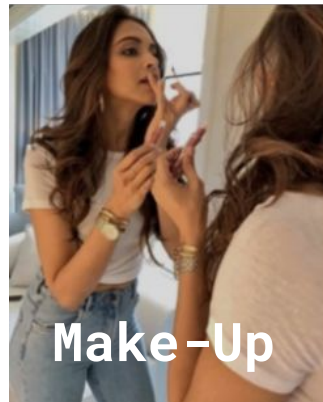
Consumer's Personalization
Interest vs. Trial



Source: The Deloitte Consumer Review: Made-to-order The rise of mass personalization.

<https://www2.deloitte.com/content/dam/Deloitte/ch/Documents/consumer-business/ch-en-consumer-business-made-to-order-consumer-review.pdf>

Develop Personalised Products & Recommendations in Beauty & Fashion



Hair Style

EARRINGS



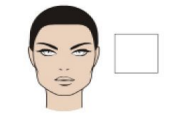
ROUND
Select long and/or angular designs, such as teardrop and dangle earrings. Avoid rounded designs.



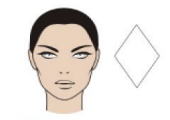
OVAL
Just about any shape will flatter oval. Go for teardrop and studs earrings.



HEART
Balance your features with elongated lines and curves. Dangle, teardrop and chandelier earrings add a lovely contrast.



SQUARE
Rounded designs will soften your features, as will elongated, dangling pieces and hoop earrings. Avoid wide earrings.



DIAMOND
Soft, long and elegant curves will balance your strikingly sharp features. Wear dangle and hoop earrings. Avoid diamond shaped styles.



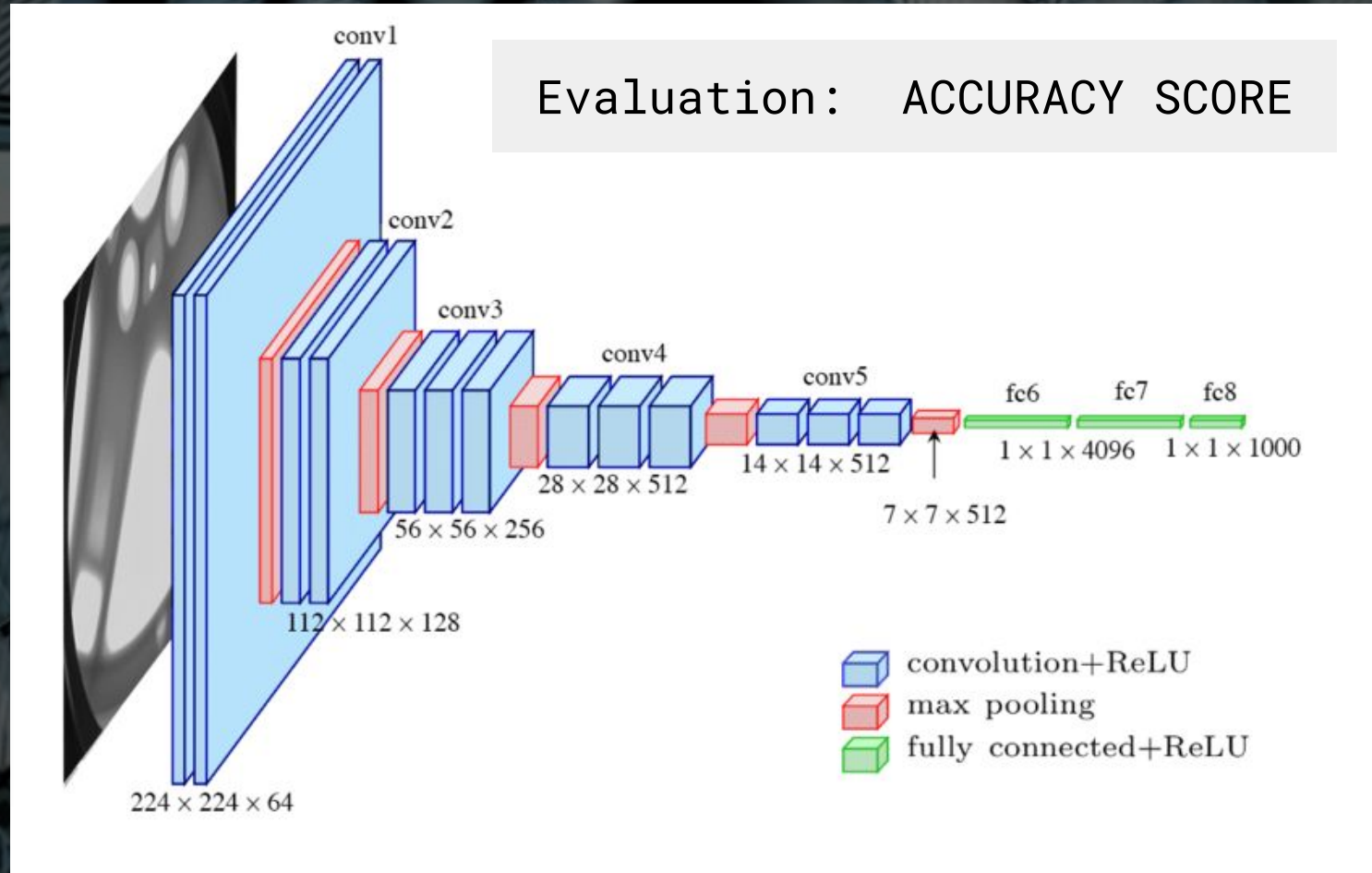
JUSTFAB

THE BEST SUNGLASSES FOR YOUR FACE SHAPE



Deep Learning

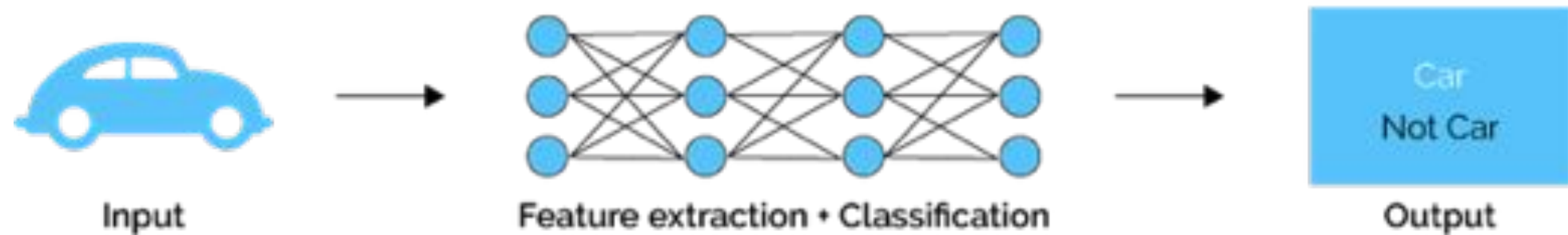
with
Convolutional
Neural Network
(CNN)

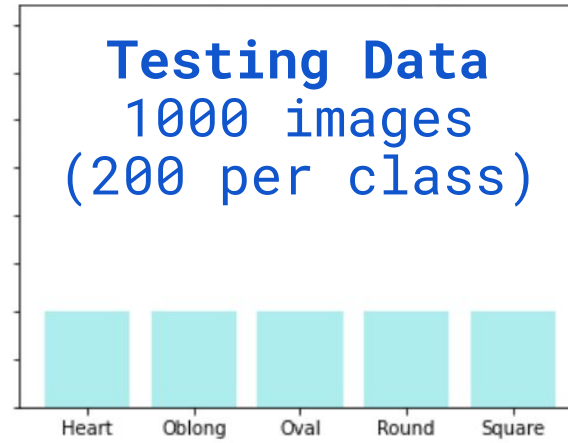
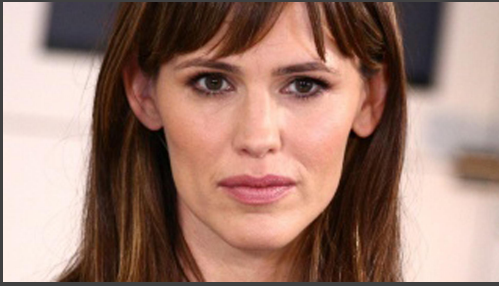


Machine Learning



Deep Learning





Training Data
4000 images
(800 per class)



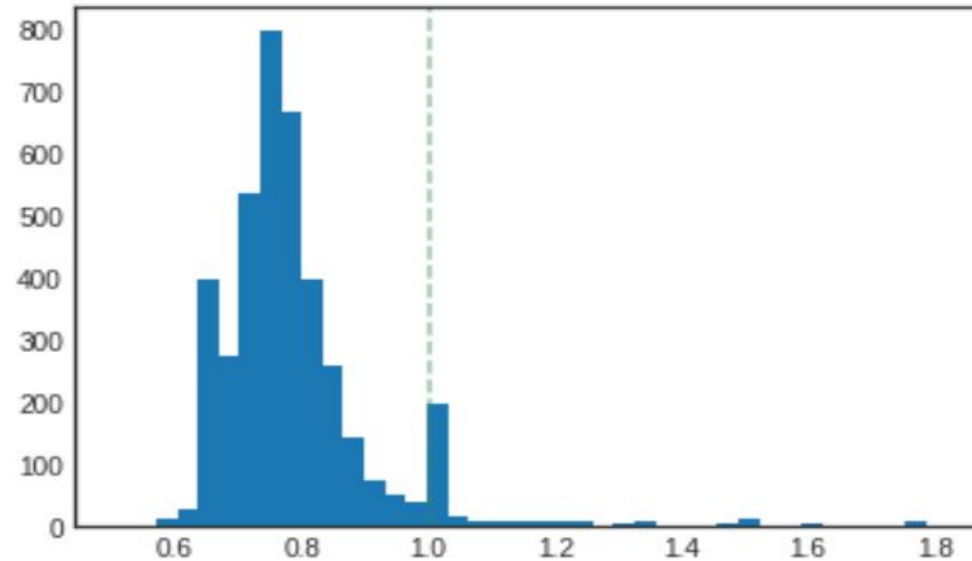
Data Exploratory Analysis

Images are mostly taken as portrait
(aspect ratio < 1)

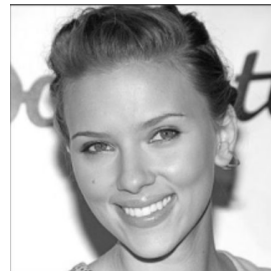
Portrait



Distribution of images by aspect_ratio
Portrait < 1 : Square $= 1$: Landscape > 1



Square = 1



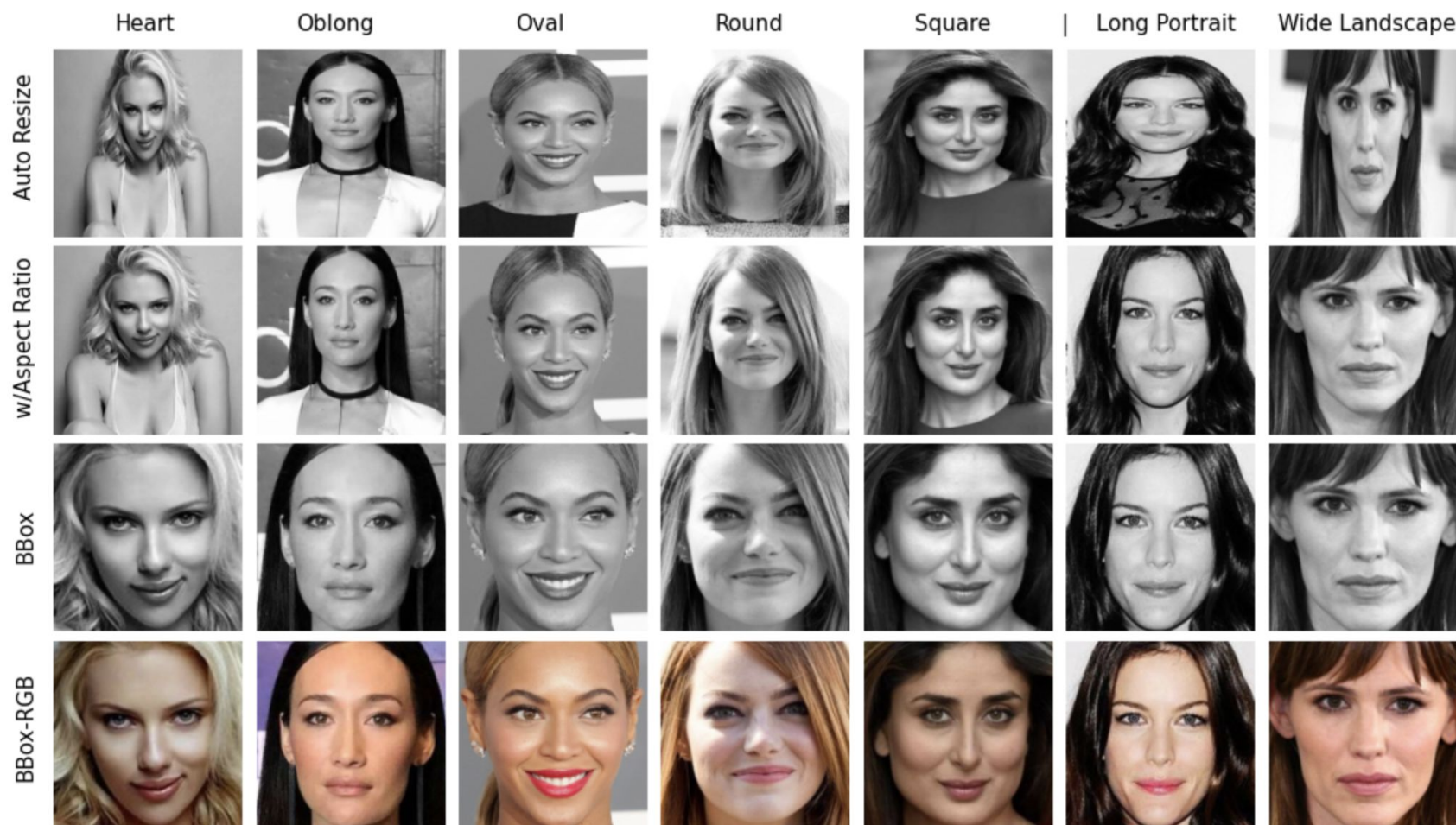
Landscape



Baseline Accuracy = 20%

Image Preprocessing

Modelling



Training
Accuracy

Validation
Accuracy

73.90%

42.70%

72.50%

47.30%

76.73%

68.60%

94.17%

71.20%

Baseline Accuracy = 20%

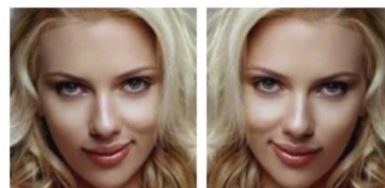
Image Preprocessing

Modelling

Training
Accuracy

Validation
Accuracy

Flipping



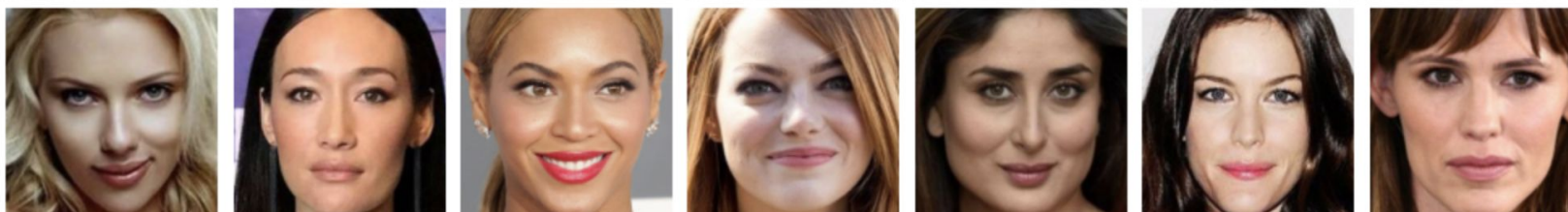
Rotating



80.20%

76.90%

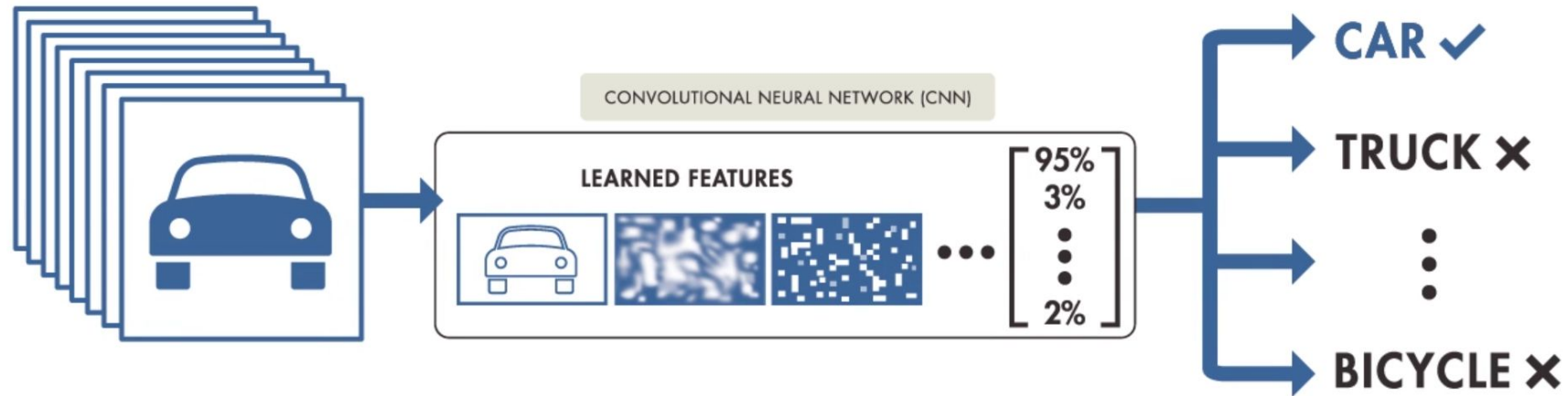
BBox-RGB



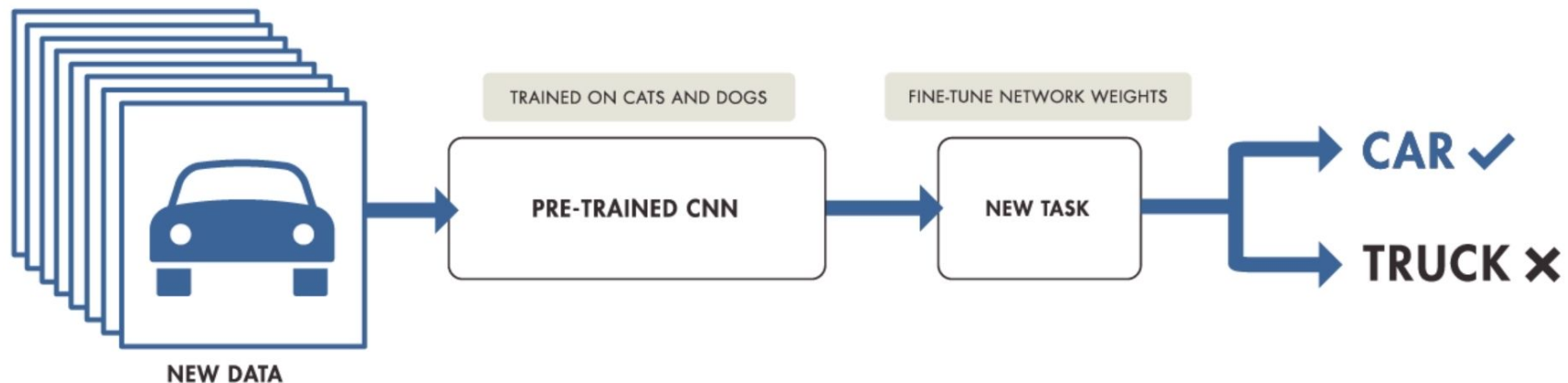
94.17%

71.20%

TRAINING FROM SCRATCH



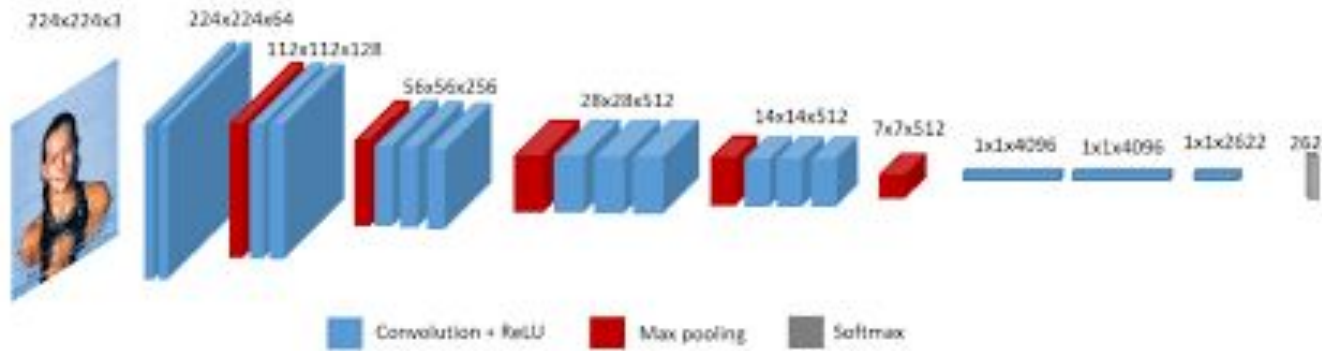
TRANSFER LEARNING



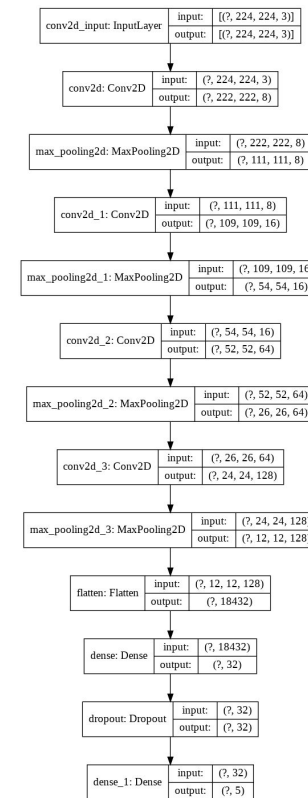
Transfer Learning VGG-FACE

| Architecture

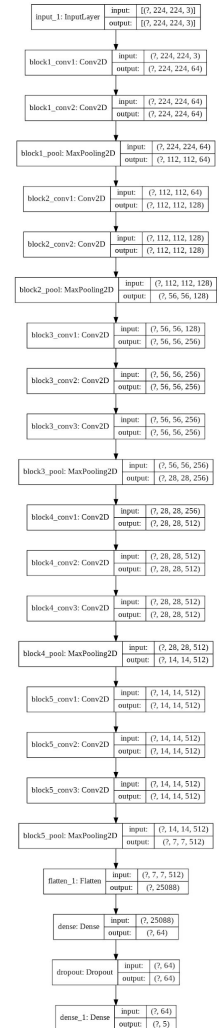
VGG-Face pre-trained weights
(trained on 2.6 Million images)



From Scratch



VGG-16



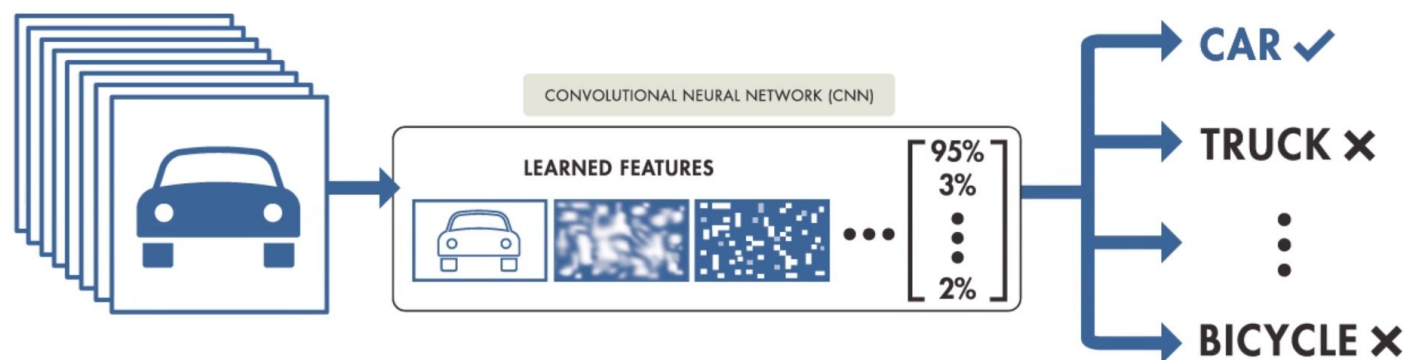
Baseline Accuracy = 20%

Modelling

Training
Accuracy

Validation
Accuracy

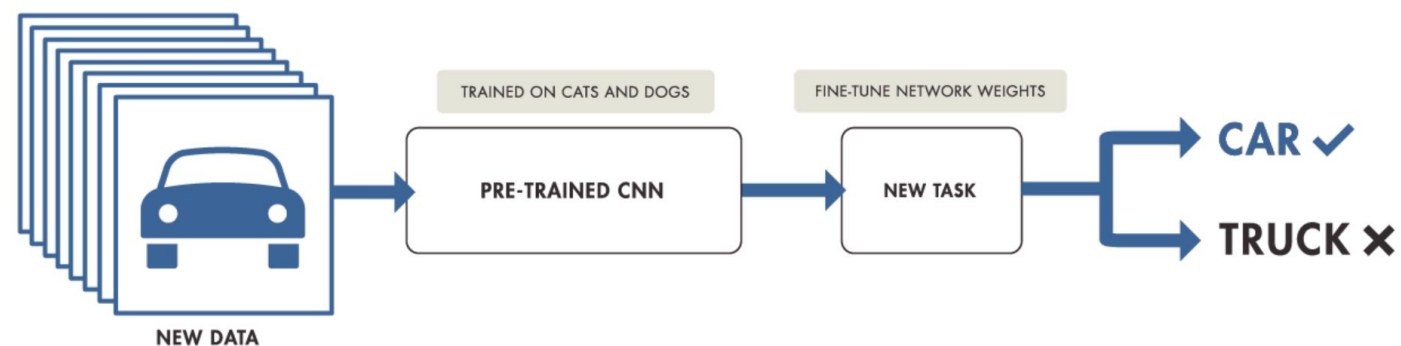
TRAINING FROM SCRATCH



80.20%

76.90%

TRANSFER LEARNING

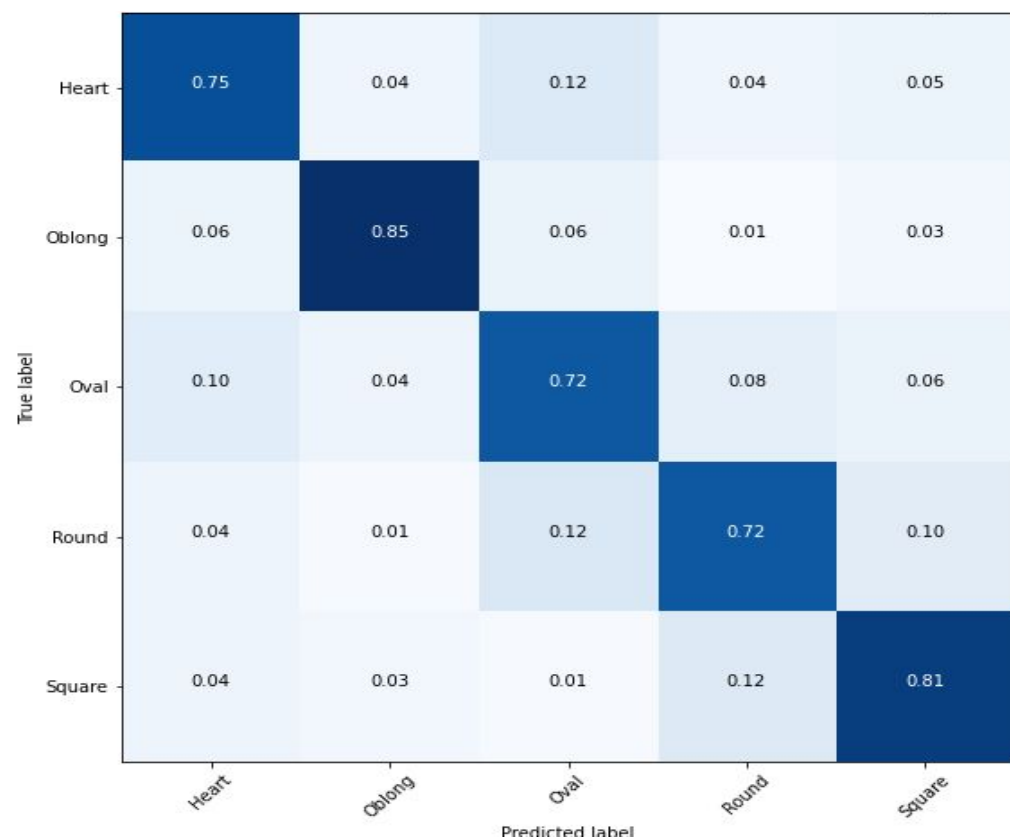


96.47%

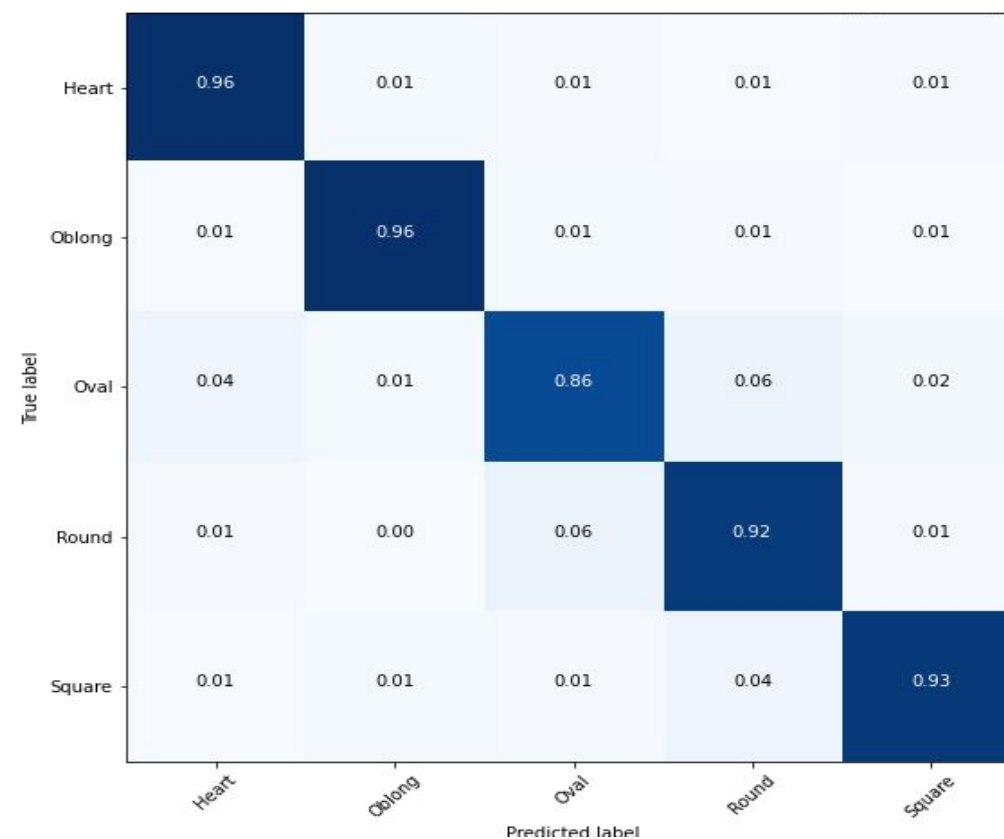
92.70%

Confusion Matrix

TRAINING FROM SCRATCH



TRANSFER LEARNING

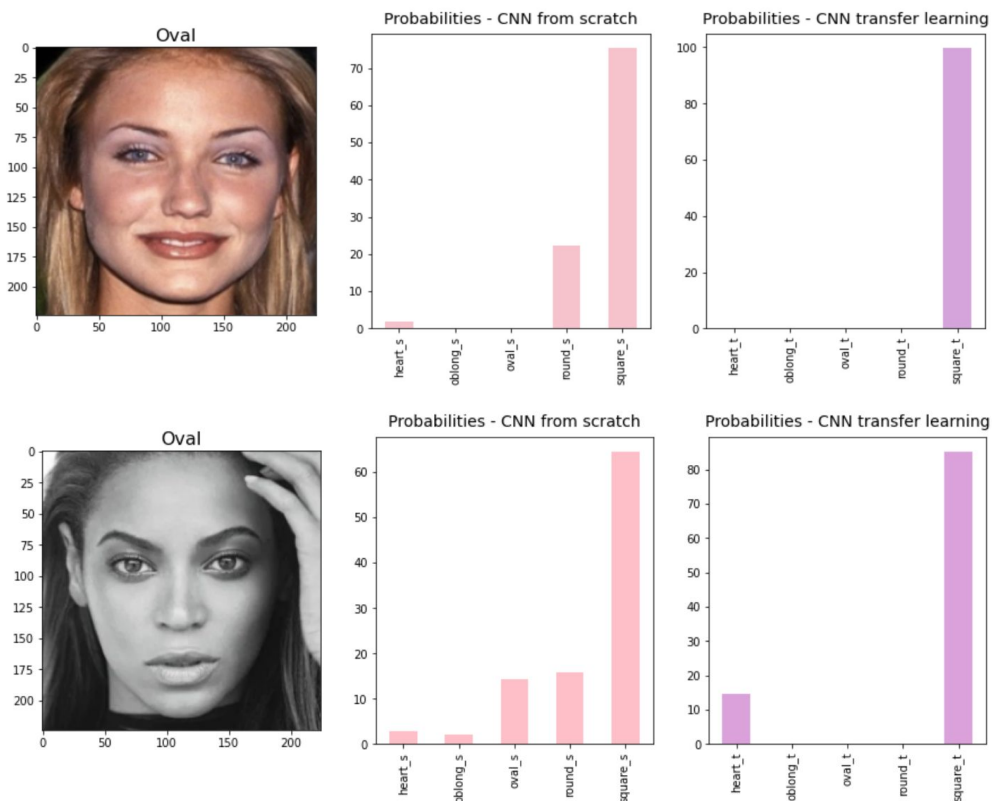


“Difficult” images misclassified by both models

Oval misclassified as Heart

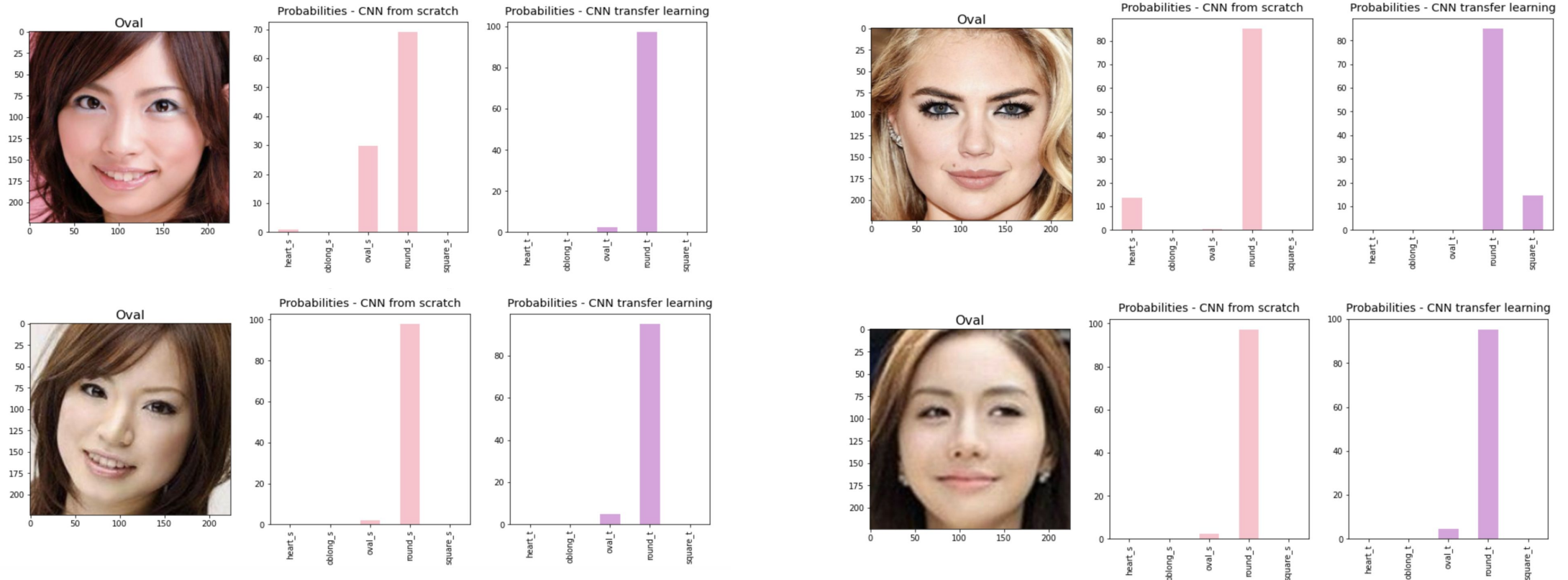


Oval misclassified as Square

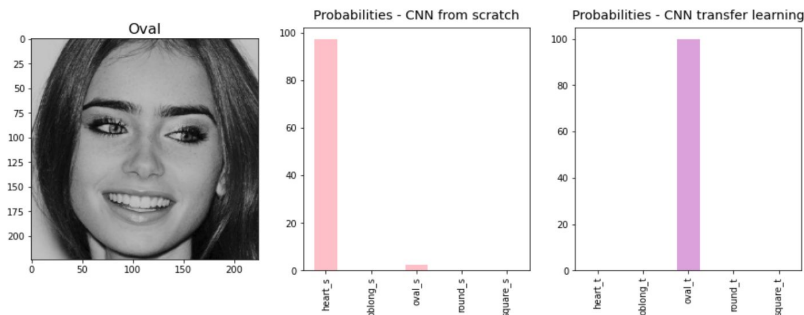
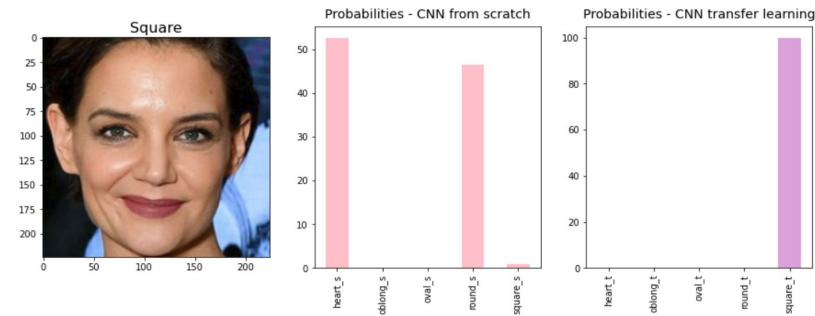
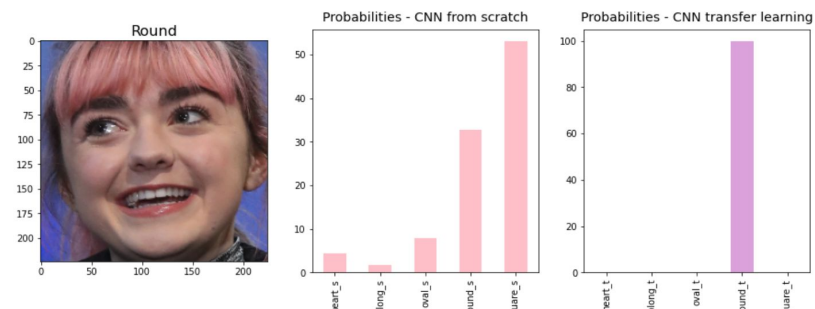
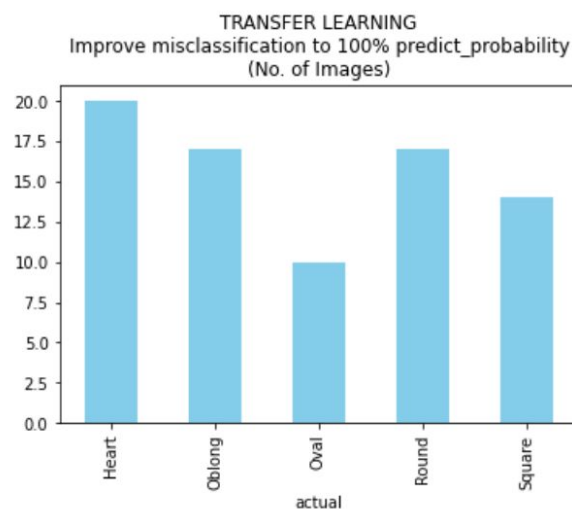
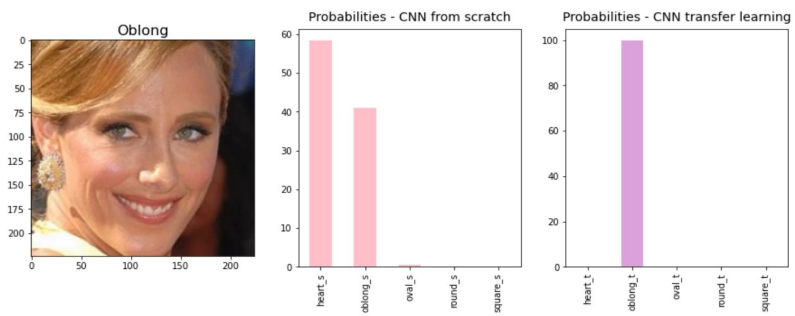
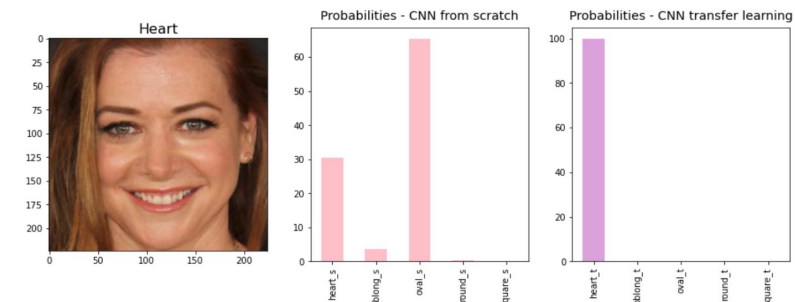


Mostly “Asian” Oval more mistaken as Round

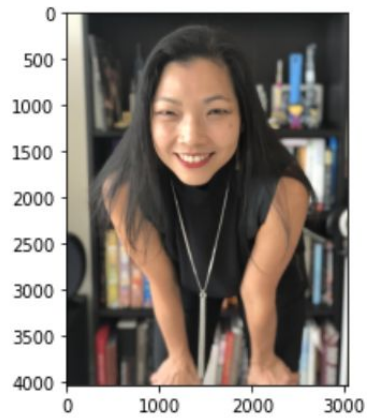
Oval misclassified as Round



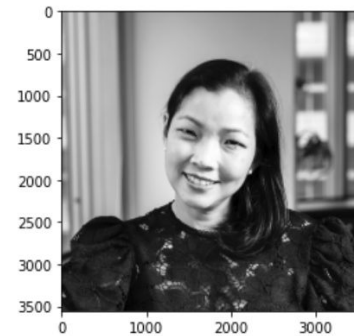
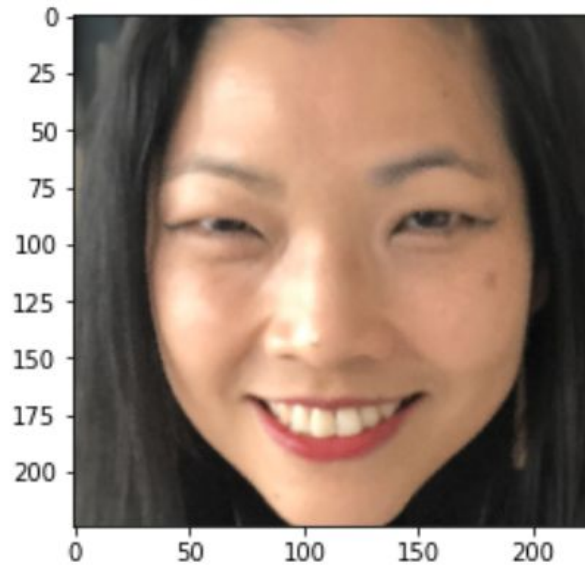
Improvement from Transfer Learning: 100% probability of the predicted class



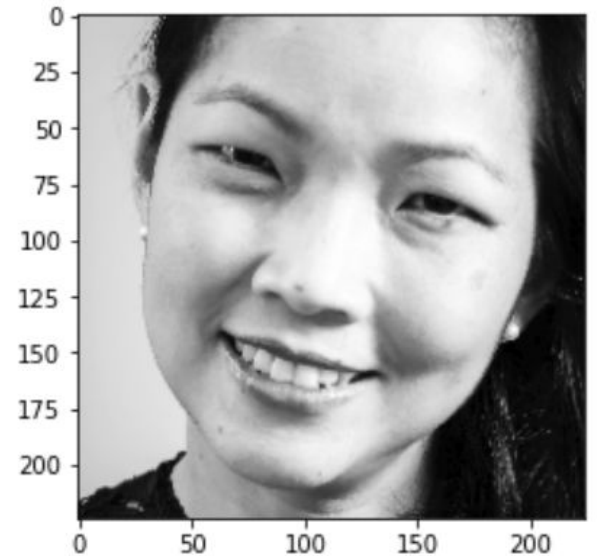
Predictions



Your face shape is Heart
Probability 94.93



Your face shape is Round
Probability 99.99



Conclusion

- The model predicted the 5 face shapes well with 92.7% accuracy.
- Key drivers are:
 - Face Detection (Bounding Box)
 - Image Augmentation with flip & rotation
 - Pretrained weights from VGG-Face

Limitations

&

Way Forward

- Trained on adult female faces
- Lower accuracy on OVAL faces
- Predictions depend on input image (angle, pose, cropping)

Extend training to:

- Male Face Shapes
- Different ages and races
- With/Without Glasses

Consistent input source (i.e. guide/bounding box in app)

1-EDA.ipynb

Capstone Project: Face Shape Classification - Part 1

This notebook consists of the following processes:

- Define the Problem Statement
- Gather Data
- Explore Data
- Prepare and save the first set of data for modelling

Problem Statement

1-EDA.ipynb

2-IMAGE_PREPROCESSING.ipynb

Capstone Project: Face Shape Classification - Part 2

This notebook further Explore Data and Image Preprocessing prior to modelling. This includes the following:

1. Image cropping & resizing by maintaining aspect ratio.
2. Face Detection with Bounding Box
3. Explore different image colors & filters
4. Prepare data and export files for modelling

2-IMAGE_PREPROCESSING....

3-CNN_MODELLING-updated.ipynb

Capstone Project: Face Shape Classification - Part 3

This notebook explores Convolutional Neural Network (CNN) Models on different image datasets based on the following image preprocessing:

- Model 1: Auto resized to 224 x 224
- Model 2: Images are cropped & resized by maintaining aspect ratio
- Model 3: Face Detection with Bounding Box in Grayscale

3-CNN_MODELLING-update...

4-IMAGE_AUGMENTATION_updated.ipynb

Capstone Project: Face Shape Classification - Part 4

This notebook explores image augmentation, including:

- Exploratory Data Analysis to identify the right variables (flip, degrees of rotation)
- Preprocessing & Modelling

```
from google.colab import drive
drive.mount('/content/drive')
```

4-IMAGE_AUGMENTATION_...

5-TRANSFER_LEARNING_VGGFACE.ipynb

Capstone Project: Face Shape Classification - Part 5

This notebook explores transfer learning from a pre-trained Oxford VGGFace model.

As the VGGFace model was built on older versions of Keras (v2.2.4) and Tensorflow (v1.14.0), hence we will only use the weights from the model and apply to VGG-16 architecture.

More details on VGGFace can be found on keras-vggface-project by Refik Can Malli.

5-TRANSFER_LEARNING_V...

6-MODEL_EVALUATION.ipynb

Capstone Project: Face Shape Classification - Part 6

This notebook evaluates the 2 best CNN models (built from scratch vs. VGG-Face). Analysis on confusion matrix, misclassification, and prediction probabilities.

```
from google.colab import drive
drive.mount('/content/drive')
```

Import Libraries

6-MODEL_EVALUATION.ipynb

7-PREDICT_AND_DEPLOY.ipynb

Capstone Project: Face Shape Classification - Part 7

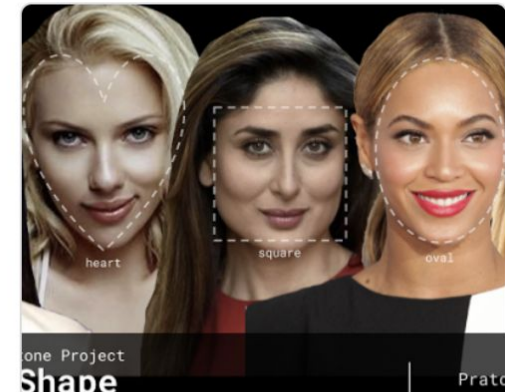
This notebook loads the model to predict on new data, and saves the model for web interface.

```
from google.colab import drive
drive.mount('/content/drive')
```

Import Libraries

```
!pip install mtcnn
```

7-PREDICT_AND_DEPLOY.ip...



Capstone Presentation_Prat...

https://github.com/Pratch-yani/Face_Shape_Classification
<https://drive.google.com/drive/folders/1r5cwyD55d33jSVSIJuZ2IguWynQ26Dq2?usp=sharing>

Thank you :)

