

- -> **picking a pre-trained model to reuse**

- -> this is one from google which is called mobile net v2 which is built into tensor flow
- -> **he's defined the base model -> then the image shape which was defined above it**
  - -> the architecture of what we want
  - -> whether we include the classifier or not
  - -> i.e the model which we're importing works for 1,000 different classes of objects -> we're not including the top
  - -> we're just loading the weights for what we want
- -> **then printing out the base model**

```
IMG_SHAPE = (IMG_SIZE, IMG_SIZE, 3)

# Create the base model from the pre-trained model MobileNet V2
base_model = tf.keras.applications.MobileNetV2(input_shape=IMG_SHAPE,
                                                include_top=False,
                                                weights='imagenet')

Downloading data from https://github.com/JonathanCMitchell/mobilenet_v2
9412608/9406464 [=====] - 0s 0us/step

[ ] base_model.summary()
```

- -> aka like the head of the dataset
- -> the architecture and the data
  - -> the architecture is like which rows and columns the matrix will have and the data is what fills it
- -> converting that imported model (to reuse) into a tensor
- -> **freezing the base**
  - -> we are reusing a model and adding our points onto the top of it
  - -> the model we are reusing is the 'base' -> freezing it means it doesn't change
  - -> the only parts we are training are our data which we are adding over the top of it

```
[75] for image, _ in train_batches.take(1):
      pass

      feature_batch = base_model(image)
      print(feature_batch.shape)

(32, 5, 5, 1280)

base_model.trainable = False

[ ] base_model.summary()
```

- -> then printing out another summary of the data
- -> **adding another classifier on top of it**
  - taking the average of all the layers and inputting them into a 1D tensor
  - -> this is the global average pooling
    - moving the points into a 1D tensor
    - -> converting the borrowed model into one dense node

- -> **and then adding a prediction layer**
  - -> the prediction layer creates the final model
  - -> running another summary then returns the number of trainable data
- -> **then combining it with our data to create a new model**

```
[ ] global_average_layer = tf.keras.layers.GlobalAveragePooling2D()
```

Finally we will add the prediction layer that will be a single dense neuron. We

```
prediction_layer = keras.layers.Dense(1)
```

Now we will combine these layers together in a model.

```
[ ] model = tf.keras.Sequential([
    base_model,
    global_average_layer,
    prediction_layer
])
```

```
[ ] model.summary()
```

- -> **then he's ran another summary**
  - -> global average pooling

```
model.summary()
```

```
Model: "sequential_3"
```

Layer (type)	Output Shape	Param #
mobilenetv2_1.00_160 (Model)	(None, 5, 5, 1280)	2257984
global_average_pooling2d_1 (	(None, 1280)	0
<b>dense_5 (Dense)</b>	(None, 1)	1281

```
=====
Total params: 2,259,265
Trainable params: 1,281
Non-trainable params: 2,257,984
=====
```

- -> the only data which is trainable is our data (not the model we've borrowed)
- -> **training the model**
  - -> the learning rate
    - -> he's picked one which is slow
    - -> not making major changes to the model
  - -> then using binary cross entropy for the loss function because there are two input parameters
  - -> evaluating the model currently without training it on our validation data
  - -> it's returning the accuracy of the model as it trains it
  - -> then running the training with the base layer

```
# We can evaluate the model right now to see how it does before training it on our new images
initial_epochs = 3
validation_steps=20

loss0,accuracy0 = model.evaluate(validation_batches, steps = validation_steps)

20/20 [=====] - 15s 737ms/step - loss: 0.7161 - accuracy: 0.5609
```

- -> it takes an hour to train with the base layer -> but has ~90% accuracy once it's been trained
- -> keras can be used to load the model -> rather than tensor flow
- **-> using the model**
  - -> model.predict <- and it's suggesting the different parameters you could have

- **Object detection**

- -> you can use tensor flow to do object detection and recognition
- -> there is an API for object detection
- -> you can use this project using the documentation on its GitHub page from tensor flow
- -> Python also has a facial recognition module
- -> you need enough training data
  - -> then to create a model which uses mobile nets
  - -> mobile nets is the google training model which he's reused in this example

- **Next**

- -> recurrent neural networks