

```
from tensorflow.keras.preprocessing.image
import ImageDataGenerator
```

Rescale to normalize data

```
train_datagen = ImageDataGenerator(rescale=1./255)

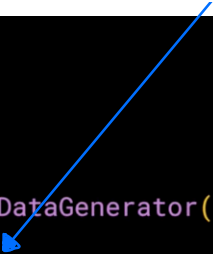
train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(300, 300),
    batch_size=128,
    class_mode='binary')
```

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    train_dir,
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    batch_size=128,
    class_mode='binary')
```

Always point it at the directory that contains sub-directories that contain your images. The names of the sub-directories will be the labels for your images that are contained within them. So make sure that the directory you're pointing to is the correct one.

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    class_mode='binary')
```



Resize images as it is loaded

```
train_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(300, 300),
    batch_size=128,
    class_mode='binary')
```

```

train_datagen = ImageDataGenerator(rescale=1./255)

train_generator = train_datagen.flow_from_directory(
    train_dir,
    target_size=(300, 300),
    batch_size=128,
    class_mode='binary')

```

```

test_datagen = ImageDataGenerator(rescale=1./255)

validation_generator = test_datagen.flow_from_directory(
    validation_dir,
    target_size=(300, 300),
    batch_size=32,
    class_mode='binary')

```

The validation generator should be exactly the same except of course it points at a different directory, the one containing the sub-directories containing the test images.

```

model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(16, (3,3), activation='relu',
                           input_shape=(300, 300, 3)),
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Conv2D(32, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(512, activation='relu'),
    tf.keras.layers.Dense(1, activation='sigmoid')
])

```

Start at 298 by 298 and then have that etc., etc. until by the end, we're at a 35 by 35 image.

```

model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(16, (3,3), activation='relu',
                           input_shape=(300, 300, 3)),
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Conv2D(32, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(512, activation='relu'),
    tf.keras.layers.Dense(1, activation='sigmoid')
])

```

Color

```

model = tf.keras.models.Sequential([
    tf.keras.layers.Conv2D(16, (3,3), activation='relu',
                           input_shape=(150, 150, 3)),
    tf.keras.layers.MaxPooling2D(2, 2),
    tf.keras.layers.Conv2D(32, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(512, activation='relu'),
    tf.keras.layers.Dense(1, activation='sigmoid')
])

```

Layer (type)	Output Shape	Param #
conv2d_5 (Conv2D)	(None, 298, 298, 16)	448
max_pooling2d_5 (MaxPooling2D)	(None, 149, 149, 16)	0
conv2d_6 (Conv2D)	(None, 147, 147, 32)	4640
max_pooling2d_6 (MaxPooling2D)	(None, 73, 73, 32)	0
conv2d_7 (Conv2D)	(None, 71, 71, 64)	18496
max_pooling2d_7 (MaxPooling2D)	(None, 35, 35, 64)	0
flatten_1 (Flatten)	(None, 78400)	0
dense_2 (Dense)	(None, 512)	40141312
dense_3 (Dense)	(None, 1)	513


Total params: 40,165,409  
 Trainable params: 40,165,409  
 Non-trainable params: 0

```
from tensorflow.keras.optimizers import RMSprop

model.compile(loss='binary_crossentropy',
              optimizer=RMSprop(lr=0.001),
              metrics=['acc'])
```

```
history = model.fit_generator(
    train_generator,
    steps_per_epoch=8,
    epochs=15,
    validation_data=validation_generator,
    validation_steps=8,
    verbose=2)
```

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history = model.fit_generator(
    train_generator,
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    epochs=15,
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```



Stream images from training directory

```
history = model.fit_generator(  
    train_generator,  
    steps_per_epoch=8,  
    epochs=15,  
    validation_data=validation_generator,  
    validation_steps=8,  
    verbose=2)
```

There are 1,024 images in the training directory, so we're loading them in 128 at a time. So in order to load them all, we need to do 8 batches.

```
history = model.fit_generator(  
    train_generator,  
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```

It had 256 images, and we wanted to handle them in batches of 32, so we will do 8 steps.

```

history = model.fit_generator(
    train_generator,
    steps_per_epoch=8,
    epochs=15,
    validation_data=validation_generator,
    validation_steps=8,
    verbose=2)

```

```

import numpy as np
from google.colab import files
from keras.preprocessing import image

uploaded = files.upload()

for fn in uploaded.keys():

    # predicting images
    path = '/content/' + fn
    img = image.load_img(path, target_size=(300, 300))
    x = image.img_to_array(img)
    x = np.expand_dims(x, axis=0)

    images = np.vstack([x])
    classes = model.predict(images, batch_size=10)
    print(classes[0])
    if classes[0]>0.5:
        print(fn + " is a human")
    else:
        print(fn + " is a horse")

```

So these parts are specific to Colab, they are what gives you the button that you can press to pick one or more images to upload. The image paths then get loaded into this list called uploaded.



```
import numpy as np
from google.colab import files
from keras.preprocessing import image
```

```
uploaded = files.upload()
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for fn in uploaded.keys():
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    # predicting images
    path = '/content/' + fn
    img = image.load_img(path, target_size=(300, 300))
    x = image.img_to_array(img)
    x = np.expand_dims(x, axis=0)
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```
    images = np.vstack([x])
    classes = model.predict(images, batch_size=10)
    print(classes[0])
    if classes[0]>0.5:
        print(fn + " is a human")
    else:
        print(fn + " is a horse")
```

The loop then iterates through all of the images in that collection. And you can load an image and prepare it to input into the model with this code. Take note to ensure that the dimensions match the input dimensions that you specified when designing the model.

```
import numpy as np
from google.colab import files
from keras.preprocessing import image
```

```
uploaded = files.upload()
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for fn in uploaded.keys():
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```
    # predicting images
    path = '/content/' + fn
    img = image.load_img(path, target_size=(300, 300))
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```
    images = np.vstack([x])
    classes = model.predict(images, batch_size=10)
    print(classes[0])
    if classes[0]>0.5:
        print(fn + " is a human")
    else:
        print(fn + " is a horse")
```

You can then call `model.predict`, passing it the details, and it will return an array of classes. In the case of binary classification, this will only contain one item with a value close to 0 for one class and close to 1 for the other.