```
In [1]: from google.colab import files
       files.upload()
       Choose Files | No file chosen
                                        Upload widget is only available when the cell has been
       executed in the current browser session. Please rerun this cell to enable.
       Saving kaggle.json to kaggle.json
       Out[1]:
       7d6879"}'}
In [2]: !mkdir ~/.kaggle
        !cp kaggle.json ~/.kaggle/
        !chmod 600 ~/.kaggle/kaggle.json
        !cd ~/.kaggle/
In [3]: !kaggle competitions download -c dogs-vs-cats
       Downloading dogs-vs-cats.zip to /content
       100% 810M/812M [00:21<00:00, 42.9MB/s]
       100% 812M/812M [00:21<00:00, 40.1MB/s]
In [4]: !unzip -qq dogs-vs-cats.zip
        !unzip -qq train.zip
```

Copying Images to Train, Test and Validation Folders

```
import os, shutil, pathlib
In [5]:
        original_dir = pathlib.Path("train")
        new_base_dir = pathlib.Path("cats_vs_dogs_small")
        def make_subset(subset_name, start_index, end_index):
            for category in ("cat", "dog"):
                dir = new_base_dir / subset_name / category
                os.makedirs(dir)
                fnames = [f"{category}.{i}.jpg" for i in range(start_index, end_index)]
                for fname in fnames:
                     shutil.copyfile(src=original_dir / fname,
                                     dst=dir / fname)
        make_subset("validation", start_index=0, end_index=500)
        make_subset("test", start_index=500, end_index=1000)
        make_subset("train", start_index=1000, end_index=2000)
```

Data Preprocessing

```
In [6]: from tensorflow.keras.utils import image_dataset from directory
        train_dataset = image_dataset_from_directory(
             new_base_dir / "train",
             image size=(180, 180),
             batch_size=32)
        validation_dataset = image_dataset_from_directory(
            new_base_dir / "validation",
             image_size=(180, 180),
            batch size=32)
        test_dataset = image_dataset_from_directory(
            new_base_dir / "test",
```

```
image_size=(180, 180),
batch_size=32)

Found 2000 files belonging to 2 classes.
Found 1000 files belonging to 2 classes.
Found 1000 files belonging to 2 classes.
Model Building
```

```
In [7]: from tensorflow import keras
        from tensorflow.keras import layers
        from keras import regularizers
        inputs = keras.Input(shape=(180, 180, 3))
        x = layers.Rescaling(1./255)(inputs)
        x = layers.Conv2D(filters=32,kernel_size=3,activation="relu")(x)
        x = layers.MaxPooling2D(pool_size=2)(x)
        x = layers.Conv2D(filters=64, kernel_size=3, activation="relu")(x)
        x = layers.MaxPooling2D(pool_size=2)(x)
        x = layers.Conv2D(filters=128,kernel_size=3, activation="relu")(x)
        x = layers.MaxPooling2D(pool_size=2)(x)
        x = layers.Conv2D(filters=256,kernel_size=3, activation="relu")(x)
        x = layers.MaxPooling2D(pool_size=2)(x)
        x = layers.Conv2D(filters=256, kernel_size=3, activation="relu")(x)
        x = layers.Flatten()(x)
        outputs = layers.Dense(1, activation="sigmoid")(x)
        model = keras.Model(inputs=inputs, outputs=outputs)
        model.summary()
```

Model: "model"

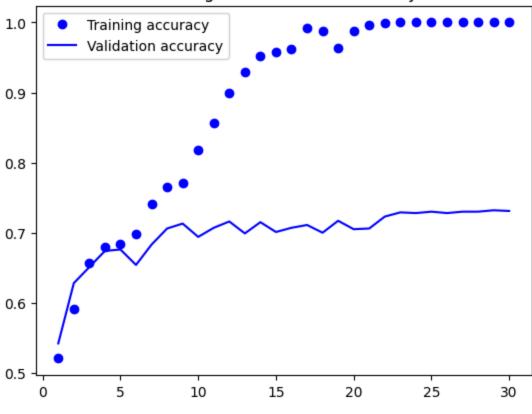
```
Layer (type)
                                   Output Shape
                                                           Param #
        ------
         input_1 (InputLayer)
                                   [(None, 180, 180, 3)]
         rescaling (Rescaling)
                                   (None, 180, 180, 3)
         conv2d (Conv2D)
                                   (None, 178, 178, 32)
                                                           896
         max pooling2d (MaxPooling2 (None, 89, 89, 32)
         D)
         conv2d_1 (Conv2D)
                                   (None, 87, 87, 64)
                                                           18496
         max_pooling2d_1 (MaxPoolin (None, 43, 43, 64)
         g2D)
         conv2d_2 (Conv2D)
                                   (None, 41, 41, 128)
                                                           73856
         max pooling2d 2 (MaxPoolin (None, 20, 20, 128)
         g2D)
         conv2d_3 (Conv2D)
                                   (None, 18, 18, 256)
                                                           295168
         max_pooling2d_3 (MaxPoolin (None, 9, 9, 256)
         g2D)
         conv2d_4 (Conv2D)
                                   (None, 7, 7, 256)
                                                           590080
         flatten (Flatten)
                                   (None, 12544)
         dense (Dense)
                                                           12545
                                   (None, 1)
        ______
        Total params: 991041 (3.78 MB)
        Trainable params: 991041 (3.78 MB)
        Non-trainable params: 0 (0.00 Byte)
        from keras.optimizers import Adam
In [8]:
        model.compile(loss="binary_crossentropy",
                    optimizer="adam",
                     metrics=["accuracy"])
In [9]:
        for data_batch, labels_batch in train_dataset:
            print("data batch shape:", data_batch.shape)
            print("labels batch shape:", labels_batch.shape)
        data batch shape: (32, 180, 180, 3)
        labels batch shape: (32,)
In [10]:
        callbacks = [
            keras.callbacks.ModelCheckpoint(
                filepath="convnet_from_scratch.keras",
                save_best_only=True,
                monitor="val_loss")
        history = model.fit(
```

train\_dataset,
epochs=20,
validation\_data=validation\_dataset,
callbacks=callbacks)

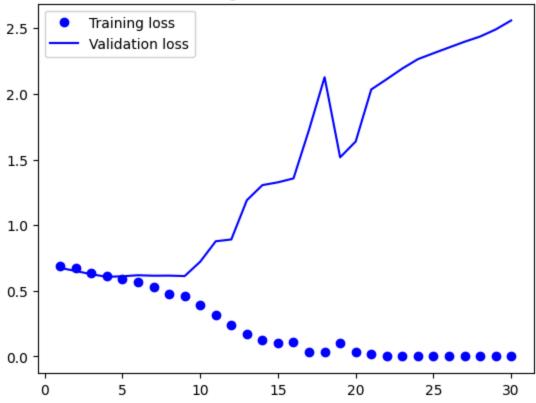
```
Epoch 1/30
63/63 [============= - 7s 34ms/step - loss: 0.6906 - accuracy: 0.52
05 - val_loss: 0.6742 - val_accuracy: 0.5420
Epoch 2/30
05 - val_loss: 0.6512 - val_accuracy: 0.6280
Epoch 3/30
63/63 [=================== ] - 1s 23ms/step - loss: 0.6331 - accuracy: 0.65
65 - val_loss: 0.6264 - val_accuracy: 0.6510
63/63 [============ - 2s 23ms/step - loss: 0.6097 - accuracy: 0.67
90 - val_loss: 0.6065 - val_accuracy: 0.6740
Epoch 5/30
63/63 [============== ] - 2s 23ms/step - loss: 0.5914 - accuracy: 0.68
45 - val loss: 0.6108 - val accuracy: 0.6760
Epoch 6/30
63/63 [============== - - 1s 22ms/step - loss: 0.5667 - accuracy: 0.69
80 - val_loss: 0.6188 - val_accuracy: 0.6540
Epoch 7/30
63/63 [============ - 1s 22ms/step - loss: 0.5297 - accuracy: 0.74
15 - val loss: 0.6151 - val accuracy: 0.6830
Epoch 8/30
63/63 [============== - - 1s 22ms/step - loss: 0.4780 - accuracy: 0.76
55 - val_loss: 0.6159 - val_accuracy: 0.7060
Epoch 9/30
63/63 [============ - 1s 21ms/step - loss: 0.4626 - accuracy: 0.77
15 - val loss: 0.6124 - val accuracy: 0.7130
Epoch 10/30
63/63 [============] - 1s 21ms/step - loss: 0.3951 - accuracy: 0.81
85 - val_loss: 0.7217 - val_accuracy: 0.6940
Epoch 11/30
63/63 [================== ] - 1s 21ms/step - loss: 0.3181 - accuracy: 0.85
60 - val_loss: 0.8775 - val_accuracy: 0.7070
Epoch 12/30
63/63 [============== - 1s 21ms/step - loss: 0.2431 - accuracy: 0.89
90 - val_loss: 0.8911 - val_accuracy: 0.7160
Epoch 13/30
95 - val_loss: 1.1898 - val_accuracy: 0.6990
Epoch 14/30
15 - val_loss: 1.3050 - val_accuracy: 0.7150
Epoch 15/30
63/63 [======================= ] - 1s 21ms/step - loss: 0.1052 - accuracy: 0.95
85 - val_loss: 1.3263 - val_accuracy: 0.7010
Epoch 16/30
63/63 [============= - - 1s 22ms/step - loss: 0.1075 - accuracy: 0.96
15 - val_loss: 1.3561 - val_accuracy: 0.7070
Epoch 17/30
63/63 [============] - 1s 22ms/step - loss: 0.0360 - accuracy: 0.99
20 - val_loss: 1.7265 - val_accuracy: 0.7110
Epoch 18/30
63/63 [============ - 1s 22ms/step - loss: 0.0325 - accuracy: 0.98
80 - val_loss: 2.1266 - val_accuracy: 0.7000
63/63 [============== - - 1s 22ms/step - loss: 0.0997 - accuracy: 0.96
40 - val_loss: 1.5165 - val_accuracy: 0.7170
Epoch 20/30
63/63 [=======================] - 1s 22ms/step - loss: 0.0373 - accuracy: 0.98
75 - val_loss: 1.6372 - val_accuracy: 0.7050
```

```
Epoch 21/30
        60 - val_loss: 2.0332 - val_accuracy: 0.7060
        Epoch 22/30
        63/63 [================== ] - 1s 21ms/step - loss: 0.0055 - accuracy: 0.99
       95 - val_loss: 2.1120 - val_accuracy: 0.7230
       Epoch 23/30
        63/63 [=======================] - 1s 21ms/step - loss: 0.0020 - accuracy: 1.00
       00 - val_loss: 2.1930 - val_accuracy: 0.7290
        Epoch 24/30
        63/63 [================== ] - 1s 21ms/step - loss: 6.9427e-04 - accuracy:
        1.0000 - val_loss: 2.2640 - val_accuracy: 0.7280
        Epoch 25/30
       63/63 [============] - 1s 21ms/step - loss: 4.0110e-04 - accuracy:
        1.0000 - val loss: 2.3091 - val accuracy: 0.7300
        Epoch 26/30
        1.0000 - val_loss: 2.3529 - val_accuracy: 0.7280
        Epoch 27/30
       63/63 [============ ] - 1s 21ms/step - loss: 2.5498e-04 - accuracy:
       1.0000 - val loss: 2.3967 - val accuracy: 0.7300
        Epoch 28/30
        63/63 [=================] - 1s 22ms/step - loss: 2.1239e-04 - accuracy:
        1.0000 - val_loss: 2.4364 - val_accuracy: 0.7300
        Epoch 29/30
        63/63 [================== ] - 1s 22ms/step - loss: 1.7774e-04 - accuracy:
        1.0000 - val loss: 2.4901 - val accuracy: 0.7320
        Epoch 30/30
        1.0000 - val loss: 2.5587 - val accuracy: 0.7310
In [11]: import matplotlib.pyplot as plt
        accuracy = history.history["accuracy"]
        val accuracy = history.history["val accuracy"]
        loss = history.history["loss"]
        val_loss = history.history["val_loss"]
        epochs = range(1, len(accuracy) + 1)
        plt.plot(epochs, accuracy, "bo", label="Training accuracy")
        plt.plot(epochs, val_accuracy, "b", label="Validation accuracy")
        plt.title("Training and validation accuracy")
        plt.legend()
        plt.figure()
        plt.plot(epochs, loss, "bo", label="Training loss")
        plt.plot(epochs, val_loss, "b", label="Validation loss")
        plt.title("Training and validation loss")
        plt.legend()
        plt.show()
```

## Training and validation accuracy



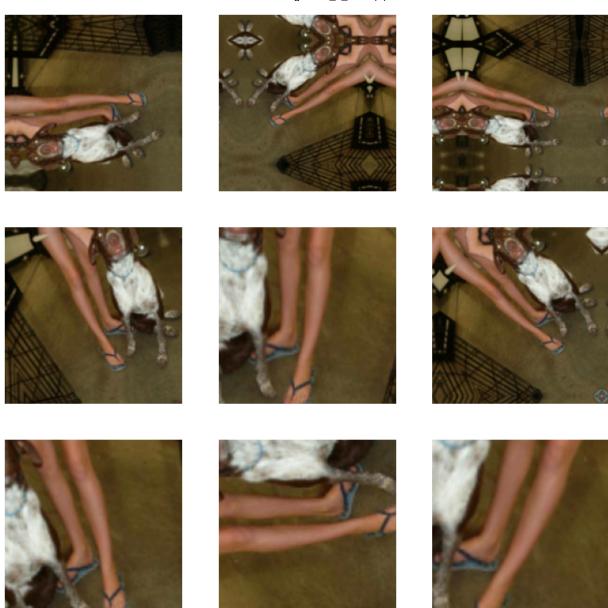
## Training and validation loss



```
In [12]: test_model = keras.models.load_model("convnet_from_scratch.keras")
    test_loss, test_acc = test_model.evaluate(test_dataset)
    print(f"Test_accuracy: {test_acc:.3f}")
```

Adding Data Augmentation

```
import warnings
warnings.filterwarnings("ignore")
plt.figure(figsize=(10, 10))
for images, _ in train_dataset.take(1):
    for i in range(9):
        augmented_images = data_augmentation(images)
        ax = plt.subplot(3, 3, i + 1)
        plt.imshow(augmented_images[0].numpy().astype("uint8"))
        plt.axis("off")
```



```
In [15]: inputs = keras.Input(shape=(180, 180, 3))
         x = data_augmentation(inputs)
         x = layers.Rescaling(1./255)(inputs)
         x = layers.Conv2D(filters=32,kernel_size=3,activation="relu")(x)
         x = layers.MaxPooling2D(pool_size=2)(x)
         x = layers.Conv2D(filters=64, kernel_size=3, activation="relu")(x)
         x = layers.MaxPooling2D(pool_size=2)(x)
         x = layers.Conv2D(filters=128,kernel_size=3, activation="relu")(x)
         x = layers.MaxPooling2D(pool_size=2)(x)
         x = layers.Conv2D(filters=256,kernel_size=3, activation="relu")(x)
         x = layers.MaxPooling2D(pool_size=2)(x)
         x = layers.Conv2D(filters=256, kernel_size=3, activation="relu")(x)
         \#x = Layers.Conv2D(filters=512,strides=2, kernel_size=3, activation="relu")(x)
         x = layers.Flatten()(x)
         x = layers.Dense(512, activation='relu', kernel_regularizer=regularizers.12(0.0001))(x
         x = layers.Dropout(0.5)(x)
         outputs = layers.Dense(1, activation="sigmoid")(x)
         model = keras.Model(inputs=inputs, outputs=outputs)
         model.summary()
         from keras.optimizers import Adam
```

```
model.compile(loss="binary_crossentropy",
             optimizer=Adam(learning_rate=0.0001),
              metrics=["accuracy"])
```

Model: "model\_1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)		0
rescaling_1 (Rescaling)	(None, 180, 180, 3)	0
conv2d_5 (Conv2D)	(None, 178, 178, 32)	896
<pre>max_pooling2d_4 (MaxPoolin g2D)</pre>	(None, 89, 89, 32)	0
conv2d_6 (Conv2D)	(None, 87, 87, 64)	18496
<pre>max_pooling2d_5 (MaxPoolin g2D)</pre>	(None, 43, 43, 64)	0
conv2d_7 (Conv2D)	(None, 41, 41, 128)	73856
<pre>max_pooling2d_6 (MaxPoolin g2D)</pre>	(None, 20, 20, 128)	0
conv2d_8 (Conv2D)	(None, 18, 18, 256)	295168
<pre>max_pooling2d_7 (MaxPoolin g2D)</pre>	(None, 9, 9, 256)	0
conv2d_9 (Conv2D)	(None, 7, 7, 256)	590080
flatten_1 (Flatten)	(None, 12544)	0
dense_1 (Dense)	(None, 512)	6423040
dropout (Dropout)	(None, 512)	0
dense_2 (Dense)	(None, 1)	513
Total params: 7402049 (28.24 MB) Trainable params: 7402049 (28.24 MB)		

Non-trainable params: 0 (0.00 Byte)

```
In [16]: callbacks = [
             keras.callbacks.ModelCheckpoint(
                 filepath="convnet_from_scratch_with_augmentation.keras",
                 save_best_only=True,
                 monitor="val_loss")
         history = model.fit(
             train_dataset,
             epochs=30,
             validation_data=validation_dataset,
             callbacks=callbacks)
```

```
Epoch 1/30
63/63 [============= - - 5s 31ms/step - loss: 0.7761 - accuracy: 0.50
35 - val_loss: 0.7564 - val_accuracy: 0.5000
Epoch 2/30
63/63 [============= - 2s 29ms/step - loss: 0.7297 - accuracy: 0.58
20 - val_loss: 0.7165 - val_accuracy: 0.5720
Epoch 3/30
63/63 [=======================] - 2s 30ms/step - loss: 0.7017 - accuracy: 0.62
05 - val_loss: 0.6932 - val_accuracy: 0.6610
63/63 [============= - 2s 29ms/step - loss: 0.6748 - accuracy: 0.64
05 - val_loss: 0.6517 - val_accuracy: 0.6600
Epoch 5/30
25 - val loss: 0.6266 - val accuracy: 0.6810
Epoch 6/30
30 - val_loss: 0.5905 - val_accuracy: 0.7150
Epoch 7/30
63/63 [============= - 1s 22ms/step - loss: 0.5872 - accuracy: 0.72
20 - val loss: 0.5940 - val accuracy: 0.7160
Epoch 8/30
90 - val_loss: 0.5710 - val_accuracy: 0.7130
Epoch 9/30
63/63 [============ - 1s 22ms/step - loss: 0.5278 - accuracy: 0.75
70 - val_loss: 0.5740 - val_accuracy: 0.7070
Epoch 10/30
35 - val_loss: 0.5583 - val_accuracy: 0.7270
Epoch 11/30
63/63 [================== ] - 2s 29ms/step - loss: 0.4733 - accuracy: 0.79
80 - val_loss: 0.5515 - val_accuracy: 0.7410
Epoch 12/30
50 - val_loss: 0.6010 - val_accuracy: 0.7390
Epoch 13/30
80 - val_loss: 0.5505 - val_accuracy: 0.7550
Epoch 14/30
65 - val_loss: 0.5605 - val_accuracy: 0.7460
Epoch 15/30
63/63 [=======================] - 1s 22ms/step - loss: 0.3649 - accuracy: 0.85
35 - val_loss: 0.5726 - val_accuracy: 0.7580
Epoch 16/30
63/63 [============= - - 1s 22ms/step - loss: 0.3540 - accuracy: 0.85
40 - val_loss: 0.5677 - val_accuracy: 0.7630
Epoch 17/30
10 - val_loss: 0.5734 - val_accuracy: 0.7590
Epoch 18/30
63/63 [============= - 1s 22ms/step - loss: 0.2627 - accuracy: 0.90
70 - val_loss: 0.5719 - val_accuracy: 0.7740
95 - val_loss: 0.5632 - val_accuracy: 0.7850
Epoch 20/30
63/63 [======================= ] - 1s 22ms/step - loss: 0.2048 - accuracy: 0.93
90 - val_loss: 0.5720 - val_accuracy: 0.7730
```

```
Epoch 21/30
       63/63 [=================== ] - 1s 22ms/step - loss: 0.1636 - accuracy: 0.95
       05 - val_loss: 0.6330 - val_accuracy: 0.7830
       Epoch 22/30
       63/63 [============] - 2s 23ms/step - loss: 0.1818 - accuracy: 0.94
       30 - val_loss: 0.6799 - val_accuracy: 0.7770
       Epoch 23/30
       63/63 [==================== ] - 2s 23ms/step - loss: 0.1473 - accuracy: 0.95
       50 - val_loss: 0.7129 - val_accuracy: 0.7740
       Epoch 24/30
       63/63 [================== ] - 2s 23ms/step - loss: 0.1113 - accuracy: 0.97
       65 - val_loss: 0.7209 - val_accuracy: 0.7780
       Epoch 25/30
       20 - val loss: 0.9464 - val accuracy: 0.7350
       Epoch 26/30
       90 - val_loss: 0.7943 - val_accuracy: 0.7700
       Epoch 27/30
       63/63 [============== - 1s 22ms/step - loss: 0.0723 - accuracy: 0.99
       10 - val_loss: 0.8400 - val_accuracy: 0.7620
       Epoch 28/30
       63/63 [============== - - 1s 22ms/step - loss: 0.0689 - accuracy: 0.99
       20 - val_loss: 0.8382 - val_accuracy: 0.7700
       Epoch 29/30
       63/63 [=======================] - 1s 22ms/step - loss: 0.0574 - accuracy: 0.99
       50 - val_loss: 0.8519 - val_accuracy: 0.7810
       Epoch 30/30
       70 - val_loss: 0.9228 - val_accuracy: 0.7710
In [17]: test_model = keras.models.load model(
          "convnet_from_scratch_with_augmentation.keras")
       test loss, test acc = test model.evaluate(test dataset)
       print(f"Test accuracy: {test_acc:.3f}")
       32/32 [================= ] - 1s 9ms/step - loss: 0.5681 - accuracy: 0.741
```

Choosing a Random train sample size

Train Sample 2000

Test accuracy: 0.741

```
make_subset("test", start_index=500, end_index=1000)
         make_subset("train", start_index=1000, end_index=3000)
In [19]: | from tensorflow.keras.utils import image_dataset_from_directory
         train_dataset = image_dataset_from_directory(
             new_base_dir / "train",
             image_size=(180, 180),
             batch_size=32)
         validation_dataset = image_dataset_from_directory(
             new_base_dir / "validation",
             image_size=(180, 180),
             batch size=32)
         test_dataset = image_dataset_from_directory(
             new_base_dir / "test",
             image_size=(180, 180),
             batch_size=32)
         Found 4000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
In [20]: callbacks = [
             keras.callbacks.ModelCheckpoint(
                 filepath="convnet_from_scratch_with_augmentation_2000.keras",
                 save_best_only=True,
                 monitor="val_loss")
         history = model.fit(
             train_dataset,
```

epochs=30,

callbacks=callbacks)

validation\_data=validation\_dataset,

```
Epoch 1/30
8455 - val_loss: 0.5073 - val_accuracy: 0.7910
Epoch 2/30
8817 - val_loss: 0.5004 - val_accuracy: 0.7980
Epoch 3/30
9115 - val_loss: 0.5062 - val_accuracy: 0.8030
Epoch 4/30
9252 - val_loss: 0.5214 - val_accuracy: 0.7930
Epoch 5/30
9452 - val loss: 0.5477 - val accuracy: 0.8050
Epoch 6/30
9595 - val_loss: 0.5830 - val_accuracy: 0.8150
Epoch 7/30
9615 - val loss: 0.5807 - val accuracy: 0.8040
9765 - val_loss: 0.6959 - val_accuracy: 0.7950
Epoch 9/30
9858 - val loss: 0.6770 - val accuracy: 0.8170
Epoch 10/30
9875 - val_loss: 0.7055 - val_accuracy: 0.7980
Epoch 11/30
9948 - val_loss: 0.7215 - val_accuracy: 0.8170
Epoch 12/30
9970 - val_loss: 0.7941 - val_accuracy: 0.8040
Epoch 13/30
9942 - val_loss: 0.8104 - val_accuracy: 0.7900
Epoch 14/30
9872 - val_loss: 0.8395 - val_accuracy: 0.8000
Epoch 15/30
9937 - val_loss: 0.7983 - val_accuracy: 0.8040
Epoch 16/30
9983 - val loss: 0.8633 - val accuracy: 0.8020
Epoch 17/30
9983 - val_loss: 0.9108 - val_accuracy: 0.8080
Epoch 18/30
9912 - val_loss: 0.8319 - val_accuracy: 0.7970
Epoch 19/30
9977 - val_loss: 0.8391 - val_accuracy: 0.7990
Epoch 20/30
125/125 [=======================] - 3s 20ms/step - loss: 0.0445 - accuracy: 0.
9967 - val_loss: 0.9177 - val_accuracy: 0.8100
```

```
Epoch 21/30
      9975 - val_loss: 1.3359 - val_accuracy: 0.7650
      Epoch 22/30
      9872 - val_loss: 0.8737 - val_accuracy: 0.7900
      Epoch 23/30
      9965 - val_loss: 0.9912 - val_accuracy: 0.7900
      Epoch 24/30
      9992 - val_loss: 0.9905 - val_accuracy: 0.8000
      Epoch 25/30
      125/125 [======================] - 3s 20ms/step - loss: 0.0305 - accuracy: 1.
      0000 - val_loss: 1.0272 - val_accuracy: 0.8030
      Epoch 26/30
      9990 - val_loss: 0.9589 - val_accuracy: 0.8030
      Epoch 27/30
      9975 - val loss: 1.1412 - val accuracy: 0.7770
      Epoch 28/30
      9930 - val_loss: 1.0767 - val_accuracy: 0.7930
      Epoch 29/30
      9880 - val loss: 0.9004 - val accuracy: 0.8050
      Epoch 30/30
      9977 - val_loss: 0.9505 - val_accuracy: 0.7920
In [21]: test_model = keras.models.load_model(
        "convnet_from_scratch_with_augmentation_2000.keras")
      test loss, test acc = test model.evaluate(test dataset)
      print(f"Test accuracy: {test_acc:.3f}")
      Test accuracy: 0.781
     Train Sample: 3000
In [22]: import os, shutil, pathlib
      original_dir = pathlib.Path("train")
      new_base_dir = pathlib.Path("cats_vs_dogs_small_5")
      def make_subset(subset_name, start_index, end_index):
        for category in ("cat", "dog"):
           dir = new_base_dir / subset_name / category
           os.makedirs(dir)
           fnames = [f"{category}.{i}.jpg" for i in range(start_index, end_index)]
           for fname in fnames:
             shutil.copyfile(src=original_dir / fname,
                        dst=dir / fname)
      make_subset("validation", start_index=0, end_index=500)
      make_subset("test", start_index=500, end_index=1000)
      make_subset("train", start_index=2000, end_index=4000)
```

```
In [23]: from tensorflow.keras.utils import image_dataset_from_directory
         train_dataset = image_dataset_from_directory(
             new_base_dir / "train",
             image_size=(180, 180),
             batch_size=32)
         validation_dataset = image_dataset_from_directory(
             new_base_dir / "validation",
              image_size=(180, 180),
             batch_size=32)
         test_dataset = image_dataset_from_directory(
             new_base_dir / "test",
             image_size=(180, 180),
             batch_size=32)
         Found 4000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
In [24]: callbacks = [
             keras.callbacks.ModelCheckpoint(
                 filepath="convnet_from_scratch_with_augmentation_30000.keras",
                 save_best_only=True,
                 monitor="val_loss")
         history = model.fit(
             train_dataset,
             epochs=20,
             validation_data=validation_dataset,
             callbacks=callbacks)
```

```
Epoch 1/20
8595 - val_loss: 0.5851 - val_accuracy: 0.7700
Epoch 2/20
9243 - val_loss: 0.5646 - val_accuracy: 0.8050
Epoch 3/20
9532 - val_loss: 0.6713 - val_accuracy: 0.8070
Epoch 4/20
9730 - val_loss: 0.7232 - val_accuracy: 0.8080
Epoch 5/20
9847 - val loss: 0.7073 - val accuracy: 0.8110
Epoch 6/20
9927 - val_loss: 0.8641 - val_accuracy: 0.8060
Epoch 7/20
9960 - val loss: 0.8803 - val accuracy: 0.8060
9933 - val_loss: 0.8813 - val_accuracy: 0.8090
Epoch 9/20
9965 - val loss: 0.8989 - val accuracy: 0.8060
Epoch 10/20
9977 - val_loss: 0.9687 - val_accuracy: 0.8120
Epoch 11/20
9927 - val_loss: 0.9135 - val_accuracy: 0.8040
Epoch 12/20
9973 - val_loss: 0.8961 - val_accuracy: 0.8080
Epoch 13/20
9995 - val_loss: 0.9677 - val_accuracy: 0.8130
Epoch 14/20
0000 - val_loss: 0.9682 - val_accuracy: 0.8210
Epoch 15/20
9998 - val_loss: 1.0560 - val_accuracy: 0.7950
Epoch 16/20
0000 - val loss: 1.0084 - val accuracy: 0.8040
Epoch 17/20
0000 - val_loss: 1.0253 - val_accuracy: 0.8090
Epoch 18/20
0000 - val_loss: 1.0311 - val_accuracy: 0.8130
Epoch 19/20
9992 - val_loss: 1.0240 - val_accuracy: 0.8000
Epoch 20/20
125/125 [======================] - 3s 20ms/step - loss: 0.0254 - accuracy: 1.
0000 - val_loss: 1.0623 - val_accuracy: 0.8050
```

```
test model = keras.models.load model(
In [25]:
             "convnet_from_scratch_with_augmentation_30000.keras")
         test_loss, test_acc = test_model.evaluate(test_dataset)
         print(f"Test accuracy: {test acc:.3f}")
         32/32 [============== ] - 1s 9ms/step - loss: 0.5876 - accuracy: 0.788
         0
         Test accuracy: 0.788
         Train Sample: 4000
In [26]: import os, shutil, pathlib
         original_dir = pathlib.Path("train")
         new_base_dir = pathlib.Path("cats_vs_dogs_small_3")
         def make_subset(subset_name, start_index, end_index):
             for category in ("cat", "dog"):
                 dir = new_base_dir / subset_name / category
                 os.makedirs(dir)
                 fnames = [f"{category}.{i}.jpg" for i in range(start_index, end_index)]
                 for fname in fnames:
                     shutil.copyfile(src=original_dir / fname,
                                      dst=dir / fname)
         make_subset("validation", start_index=0, end_index=500)
         make_subset("test", start_index=500, end_index=1000)
         make_subset("train", start_index=1000, end_index=5000)
In [27]: | from tensorflow.keras.utils import image_dataset_from_directory
         train dataset = image dataset from directory(
             new_base_dir / "train",
             image_size=(180, 180),
             batch_size=32)
         validation_dataset = image_dataset_from_directory(
             new_base_dir / "validation",
             image_size=(180, 180),
             batch_size=32)
         test_dataset = image_dataset_from_directory(
             new_base_dir / "test",
             image size=(180, 180),
             batch_size=32)
         Found 8000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
         Found 1000 files belonging to 2 classes.
In [28]: callbacks = [
             keras.callbacks.ModelCheckpoint(
                 filepath="convnet_from_scratch_with_augmentation_4000.keras",
                 save_best_only=True,
                 monitor="val_loss")
         history = model.fit(
             train_dataset,
             epochs=20,
             validation_data=validation_dataset,
             callbacks=callbacks)
```

```
Epoch 1/20
9166 - val_loss: 0.4791 - val_accuracy: 0.8240
Epoch 2/20
9644 - val_loss: 0.5674 - val_accuracy: 0.8210
Epoch 3/20
250/250 [======================] - 5s 18ms/step - loss: 0.0995 - accuracy: 0.
9746 - val_loss: 0.5733 - val_accuracy: 0.8140
Epoch 4/20
9909 - val_loss: 0.7475 - val_accuracy: 0.8150
Epoch 5/20
9934 - val loss: 0.7044 - val accuracy: 0.8260
Epoch 6/20
9930 - val_loss: 0.7537 - val_accuracy: 0.8200
Epoch 7/20
9979 - val loss: 0.8301 - val accuracy: 0.8090
9893 - val_loss: 0.7506 - val_accuracy: 0.8290
Epoch 9/20
9970 - val loss: 0.8698 - val accuracy: 0.8100
Epoch 10/20
9967 - val_loss: 0.8518 - val_accuracy: 0.8190
Epoch 11/20
9998 - val_loss: 0.9072 - val_accuracy: 0.8200
Epoch 12/20
9971 - val_loss: 1.0367 - val_accuracy: 0.7930
Epoch 13/20
9971 - val_loss: 0.9628 - val_accuracy: 0.8230
Epoch 14/20
9948 - val_loss: 0.8865 - val_accuracy: 0.8010
Epoch 15/20
9945 - val_loss: 0.8942 - val_accuracy: 0.8270
Epoch 16/20
9933 - val loss: 0.7595 - val accuracy: 0.8390
Epoch 17/20
9987 - val_loss: 0.9124 - val_accuracy: 0.8140
Epoch 18/20
9960 - val_loss: 0.9227 - val_accuracy: 0.8180
Epoch 19/20
9951 - val_loss: 0.9651 - val_accuracy: 0.8100
Epoch 20/20
250/250 [================== ] - 5s 18ms/step - loss: 0.0363 - accuracy: 0.
9969 - val_loss: 0.9699 - val_accuracy: 0.8250
```

Layer (type)	Output Shape	Param #
input_3 (InputLayer)	[(None, 180, 180, 3)]	0
block1_conv1 (Conv2D)	(None, 180, 180, 64)	1792
block1_conv2 (Conv2D)	(None, 180, 180, 64)	36928
<pre>block1_pool (MaxPooling2D)</pre>	(None, 90, 90, 64)	0
block2_conv1 (Conv2D)	(None, 90, 90, 128)	73856
block2_conv2 (Conv2D)	(None, 90, 90, 128)	147584
<pre>block2_pool (MaxPooling2D)</pre>	(None, 45, 45, 128)	0
block3_conv1 (Conv2D)	(None, 45, 45, 256)	295168
block3_conv2 (Conv2D)	(None, 45, 45, 256)	590080
block3_conv3 (Conv2D)	(None, 45, 45, 256)	590080
<pre>block3_pool (MaxPooling2D)</pre>	(None, 22, 22, 256)	0
block4_conv1 (Conv2D)	(None, 22, 22, 512)	1180160
block4_conv2 (Conv2D)	(None, 22, 22, 512)	2359808
block4_conv3 (Conv2D)	(None, 22, 22, 512)	2359808
<pre>block4_pool (MaxPooling2D)</pre>	(None, 11, 11, 512)	0
block5_conv1 (Conv2D)	(None, 11, 11, 512)	2359808
block5_conv2 (Conv2D)	(None, 11, 11, 512)	2359808
block5_conv3 (Conv2D)	(None, 11, 11, 512)	2359808
block5_pool (MaxPooling2D)	(None, 5, 5, 512)	0

Total params: 14714688 (56.13 MB)
Trainable params: 14714688 (56.13 MB)
Non-trainable params: 0 (0.00 Byte)

```
Epoch 1/20
0.9001 - val_loss: 0.0966 - val_accuracy: 0.9630
Epoch 2/20
250/250 [================= ] - 11s 43ms/step - loss: 0.2154 - accuracy:
0.9118 - val_loss: 0.0861 - val_accuracy: 0.9650
Epoch 3/20
0.9212 - val_loss: 0.0774 - val_accuracy: 0.9750
Epoch 4/20
0.9144 - val_loss: 0.0551 - val_accuracy: 0.9790
Epoch 5/20
0.9243 - val loss: 0.0692 - val accuracy: 0.9670
Epoch 6/20
0.9289 - val_loss: 0.1031 - val_accuracy: 0.9590
Epoch 7/20
0.9365 - val loss: 0.0618 - val accuracy: 0.9770
0.9381 - val_loss: 0.0627 - val_accuracy: 0.9820
Epoch 9/20
0.9460 - val_loss: 0.0615 - val_accuracy: 0.9780
Epoch 10/20
0.9457 - val_loss: 0.0783 - val_accuracy: 0.9770
Epoch 11/20
0.9490 - val_loss: 0.0629 - val_accuracy: 0.9780
Epoch 12/20
0.9481 - val_loss: 0.1300 - val_accuracy: 0.9590
Epoch 13/20
250/250 [================= ] - 11s 43ms/step - loss: 0.1284 - accuracy:
0.9491 - val_loss: 0.0510 - val_accuracy: 0.9820
Epoch 14/20
0.9500 - val_loss: 0.0448 - val_accuracy: 0.9790
Epoch 15/20
0.9536 - val_loss: 0.0606 - val_accuracy: 0.9740
Epoch 16/20
0.9569 - val loss: 0.0638 - val accuracy: 0.9750
Epoch 17/20
0.9578 - val_loss: 0.0497 - val_accuracy: 0.9820
Epoch 18/20
0.9556 - val_loss: 0.0604 - val_accuracy: 0.9790
Epoch 19/20
0.9544 - val_loss: 0.0543 - val_accuracy: 0.9780
Epoch 20/20
0.9620 - val_loss: 0.0604 - val_accuracy: 0.9760
```