

```
1 # install the dependencies
2 !pip install kaggle
```

```
Requirement already satisfied: kaggle in /usr/local/lib/python3.6/dist-packages (1.5.
Requirement already satisfied: certifi in /usr/local/lib/python3.6/dist-packages (fr
Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packages (fr
Requirement already satisfied: urllib3<1.25,>=1.21.1 in /usr/local/lib/python3.6/dist
Requirement already satisfied: python-dateutil in /usr/local/lib/python3.6/dist-packa
Requirement already satisfied: python-slugify in /usr/local/lib/python3.6/dist-packag
Requirement already satisfied: tqdm in /usr/local/lib/python3.6/dist-packages (from k
Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.6/dist-packages (f
Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.6/dist-pac
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.6/dist-packages
Requirement already satisfied: text-unidecode>=1.3 in /usr/local/lib/python3.6/dist-p
```

```
1 from google.colab import files
2 files.upload()
```

```
Choose Files kaggle.json
• kaggle.json(application/json) - 70 bytes, last modified: 5/17/2020 - 100% done
Saving kaggle.json to kaggle.json
{'kaggle.json': b'{"username":"solveyourerror","key":"ba2ddc59efb623337c4f7d945d370ct
```

```
1 !mkdir -p ~/.kaggle
2 !cp kaggle.json ~/.kaggle/
3 # change the permission
4 !chmod 600 ~/.kaggle/kaggle.json
```

```
1 !kaggle datasets download -d iarunava/cell-images-for-detecting-malaria
```

```
Downloading cell-images-for-detecting-malaria.zip to /content
100% 672M/675M [00:06<00:00, 116MB/s]
100% 675M/675M [00:06<00:00, 107MB/s]
```

```
1 from zipfile import ZipFile
2 file_name = "/content/cell-images-for-detecting-malaria.zip"
3 with ZipFile(file_name,'r') as zip:
4     zip.extractall()
5     print('Done')
```

```
Done
```

```
1 !pip install tensorflow-gpu==2.0.0-rc0
```

Collecting tensorflow-gpu==2.0.0-rc0

Downloading <https://files.pythonhosted.org/packages/6a/12/8c64cc62149cc21c70c550185>

|██| 380.5MB 37kB/s

Requirement already satisfied: wheel>=0.26 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: keras-applications>=1.0.8 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: opt-einsum>=2.3.2 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: keras-preprocessing>=1.0.5 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: grpcio>=1.8.6 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: gast>=0.2.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: absl-py>=0.7.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: astor>=0.6.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: six>=1.10.0 in /usr/local/lib/python3.6/dist-packages

Collecting tb-nightly<1.15.0a20190807,>=1.15.0a20190806

Downloading <https://files.pythonhosted.org/packages/bc/88/24b5fb7280e74c7cf65bde47c>

|██| 4.3MB 42.5MB/s

Requirement already satisfied: numpy<2.0,>=1.16.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: wrapt>=1.11.1 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: protobuf>=3.6.1 in /usr/local/lib/python3.6/dist-packages

Collecting tf-estimator-nightly<1.14.0.dev2019080602,>=1.14.0.dev2019080601

Downloading <https://files.pythonhosted.org/packages/21/28/f2a27a62943d5f041e4a6fd46>

|██| 501kB 43.7MB/s

Requirement already satisfied: google-pasta>=0.1.6 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: h5py in /usr/local/lib/python3.6/dist-packages (from k

Requirement already satisfied: setuptools>=41.0.0 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.6/dist-packages

Requirement already satisfied: werkzeug>=0.11.15 in /usr/local/lib/python3.6/dist-packages

```
1 import tensorflow as tf
2 from tensorflow.keras import Sequential
3 from tensorflow.keras.layers import Flatten, Dense, Conv2D, MaxPool2D, Dropout
4
5 from tensorflow.keras.preprocessing.image import ImageDataGenerator
6
7 print(tf.__version__)
```

2.0.0-rc0

```
1 import numpy as np
2 import matplotlib.pyplot as plt
```

```
1 img_width = 64
2 img_height = 64
```

```
1 datagen = ImageDataGenerator(rescale=1/255.0, validation_split=0.2)
```

```
1 train_data_generator = datagen.flow_from_directory(directory='/content/cell_
2                                                         target_size = (img_width,
3                                                         class_mode = 'binary',
4                                                         batch_size = 16,
5                                                         subset = 'training'
6 )
```

Found 22048 images belonging to 2 classes.

```
1 validation_data_generator = datagen.flow_from_directory(directory='/content',
2                                                         target_size = (img_width,
3                                                         class_mode = 'binary',
4                                                         batch_size = 16,
5                                                         subset = 'validation'
6 )
```

Found 5510 images belonging to 2 classes.

```
1 train_data_generator.labels
```

array([0, 0, 0, ..., 1, 1, 1], dtype=int32)

▼ CNN Model Building

```
1 model = Sequential()
2
3 model.add(Conv2D(16, (3,3), input_shape = (img_width, img_height, 3), activation='relu'))
4 model.add(MaxPool2D(2,2))
5 model.add(Dropout(0.2))
6
7 model.add(Conv2D(32, (3,3), activation='relu'))
8 model.add(MaxPool2D(2,2))
9 model.add(Dropout(0.3))
10
11 model.add(Flatten())
12 model.add(Dense(64, activation='relu'))
13 model.add(Dropout(0.5))
14
15 model.add(Dense(1, activation='sigmoid'))
```

```
1 model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 16)	448
max_pooling2d (MaxPooling2D)	(None, 31, 31, 16)	0
dropout (Dropout)	(None, 31, 31, 16)	0
conv2d_1 (Conv2D)	(None, 29, 29, 32)	4640
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 32)	0

```
1 # data optimization is adam
```

```
2 model.compile(optimizer='adam', loss='binary_crossentropy', metrics = ['accu
```

```
1 # train
```

```
2 history = model.fit_generator(generator=train_data_generator,
3                               steps_per_epoch = len(train_data_generator),
4                               epochs = 5,
5                               validation_data = validation_data_generator,
6                               validation_steps = len(validation_data_generat
```



```
from /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/ops/math_grad.py:13
lating:
```

```
which has the same broadcast rule as np.where
```

```
=====] - 164s 119ms/step - loss: 0.4860 - accuracy: 0.7667 - val_loss
```

```
=====] - 163s 118ms/step - loss: 0.2736 - accuracy: 0.9195 - val_loss
```

```
=====] - 163s 119ms/step - loss: 0.2463 - accuracy: 0.9321 - val_loss
```

```
=====] - 165s 120ms/step - loss: 0.1964 - accuracy: 0.9365 - val_loss
```

```
=====] - 168s 122ms/step - loss: 0.1882 - accuracy: 0.9371 - val_loss
```

```
1 history.history
```



```
{'accuracy': [0.7667362, 0.9195392, 0.9321027, 0.9365475, 0.9370918],
 'loss': [0.48602462820276204,
 0.27357066065310914,
 0.24625902730263283,
 0.19637806036108119,
 0.1881721192722159],
 'val_accuracy': [0.91288567, 0.91107076, 0.93139744, 0.9323049, 0.9353902],
 'val_loss': [0.2681370893250341,
 0.27332649928504144,
 0.22460641115903854,
 0.18770437418226746,
 0.1834789704841872]}
```

```
1 def plot_learningCurve(history, epoch):
```

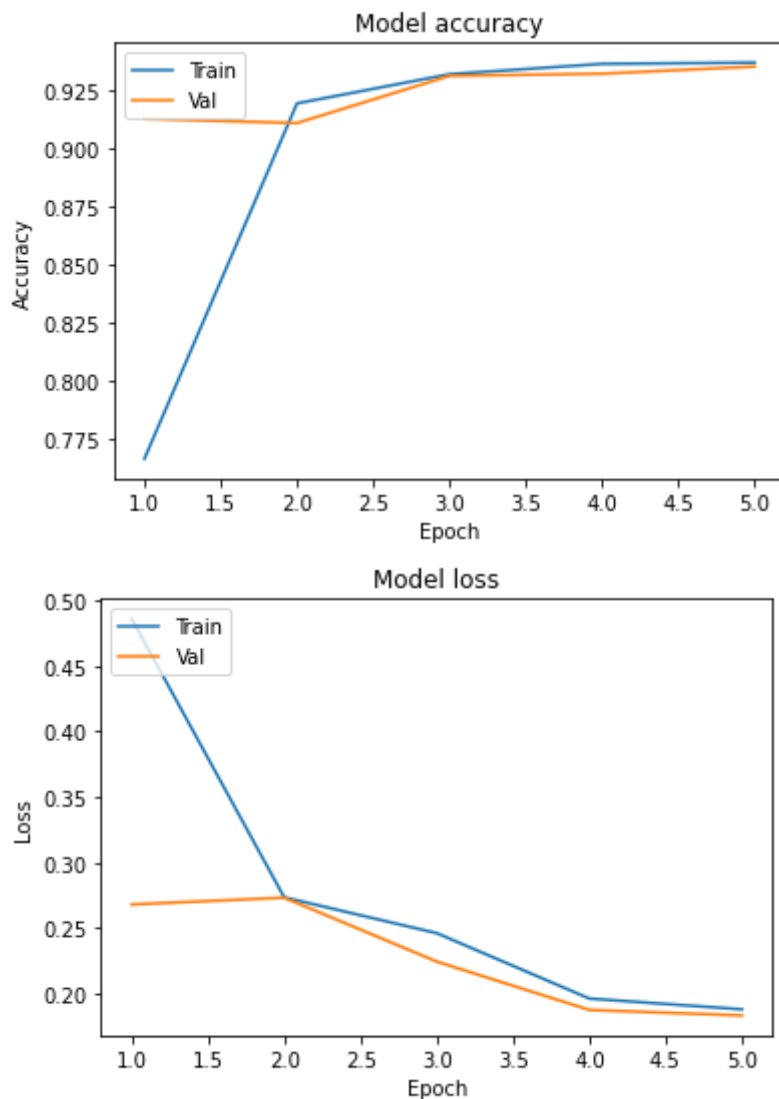
```
2 # Plot training & validation accuracy values
```

```

2  # Plot training & validation accuracy values
3  epoch_range = range(1, epoch+1)
4  plt.plot(epoch_range, history.history['accuracy'])
5  plt.plot(epoch_range, history.history['val_accuracy'])
6  plt.title('Model accuracy')
7  plt.ylabel('Accuracy')
8  plt.xlabel('Epoch')
9  plt.legend(['Train', 'Val'], loc='upper left')
10 plt.show()
11
12 # Plot training & validation loss values
13 plt.plot(epoch_range, history.history['loss'])
14 plt.plot(epoch_range, history.history['val_loss'])
15 plt.title('Model loss')
16 plt.ylabel('Loss')
17 plt.xlabel('Epoch')
18 plt.legend(['Train', 'Val'], loc='upper left')
19 plt.show()

```

```
1 plot_learningCurve(history, 5)
```



▼ **Confusion Matrix**

1