```
import tensorflow as tf
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
import pandas as pd
import os
import cv2
```

File paths for Images

```
train_data_path = 'C://Users//gaura//Documents//GitHub//Scaler-
Projects//Computer
Vision//Project//data//ninjacart_data//ninjacart_data//train'
test_data_path = 'C://Users//gaura//Documents//GitHub//Scaler-
Projects//Computer
Vision//Project//data//ninjacart_data//ninjacart_data//test'
os.listdir(train_data_path)
['indian market', 'onion', 'potato', 'tomato']
```

Training image data count check

```
train_image_count_check = {}
for value in os.listdir(train_data_path):
    tmp_path = train_data_path+"//"+str(value)
    train_image_count_check[str(value)] = len(os.listdir(tmp_path))
train_image_count_check
{'indian market': 599, 'onion': 849, 'potato': 898, 'tomato': 789}
```

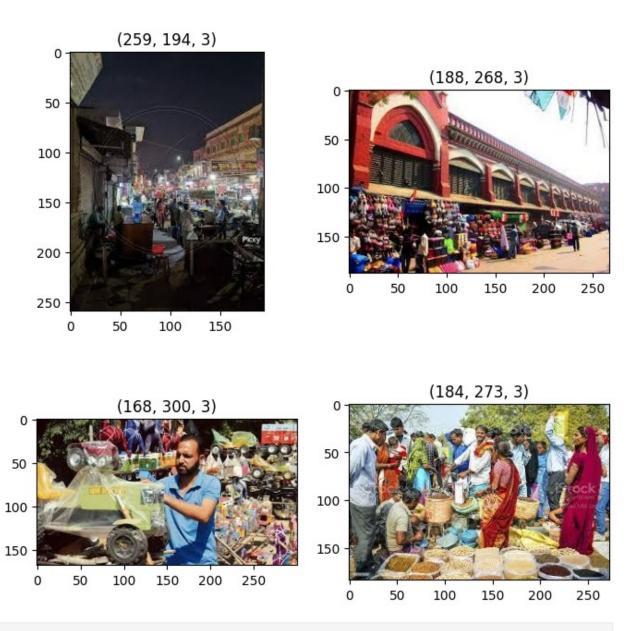
Testing image data count check

```
test_image_count_check = {}
for value in os.listdir(test_data_path):
    tmp_path = test_data_path+"//"+str(value)
    test_image_count_check[str(value)] = len(os.listdir(tmp_path))
test_image_count_check
{'indian market': 81, 'onion': 83, 'potato': 81, 'tomato': 106}
```

Ploting Images from Training Folder

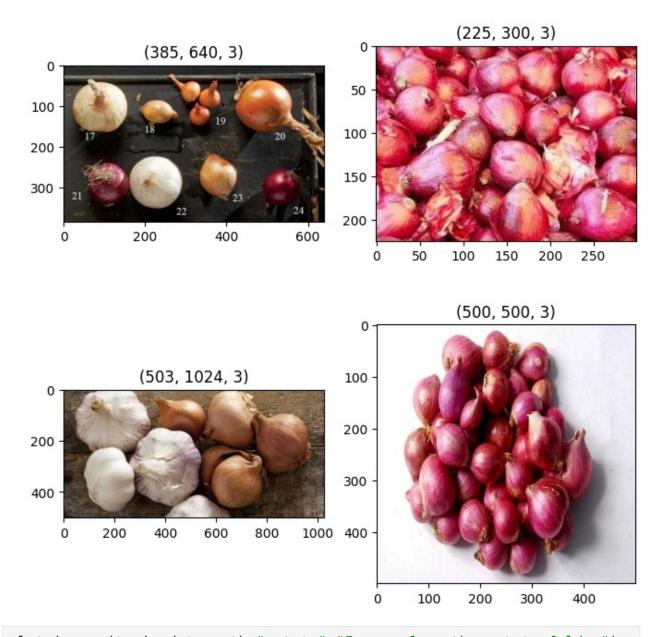
```
def plot images(path,label,title):
    data = path+"//"+label
    first four images = os.listdir(data)[:4]
    fig, axs = plt.subplots(2,2, figsize=(8,8))
    for img,ax in enumerate(axs.flat):
        img_path = data+"//"+first_four_images[img]
        image = plt.imread(img path)
        ax.set_title(image.shape)
        ax.imshow(image)
    fig.suptitle(title)
    plt.show()
os.listdir(train_data_path+"//"+'indian market')[-1:-4:-1]
['village-market-827122__340.jpg', 'vendor-4049916__340.jpg',
'uyg.jpeg']
plot images(train data path, "indian market", "Images from the Indian
Market")
```

Images from the Indian Market



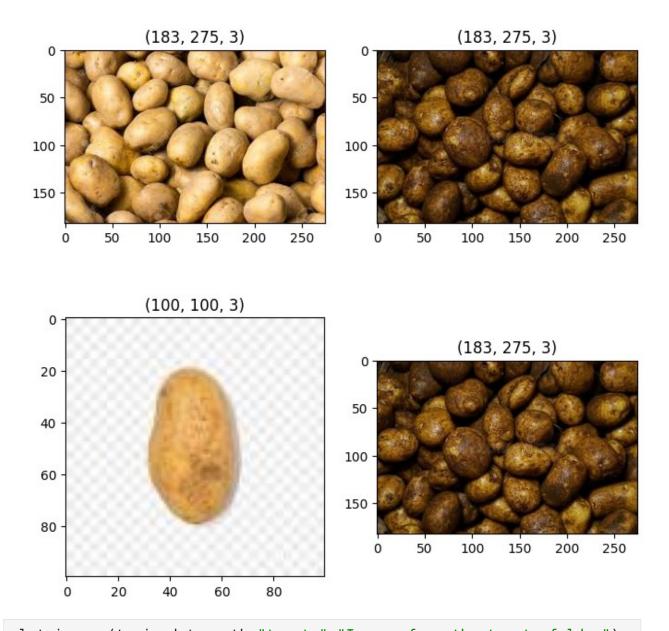
plot_images(train_data_path,"onion","Images from the onion folder")

Images from the onion folder



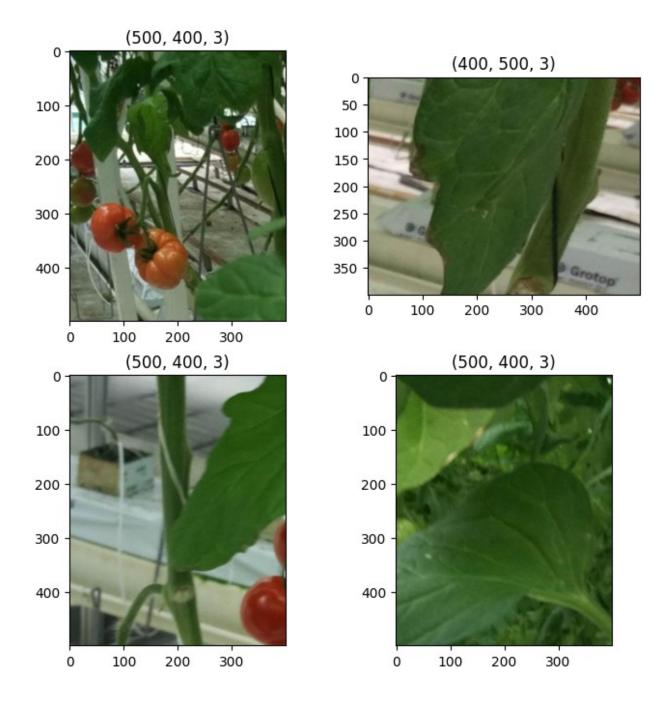
plot_images(train_data_path,"potato","Images from the potato folder")

Images from the potato folder



plot_images(train_data_path,"tomato","Images from the tomato folder")

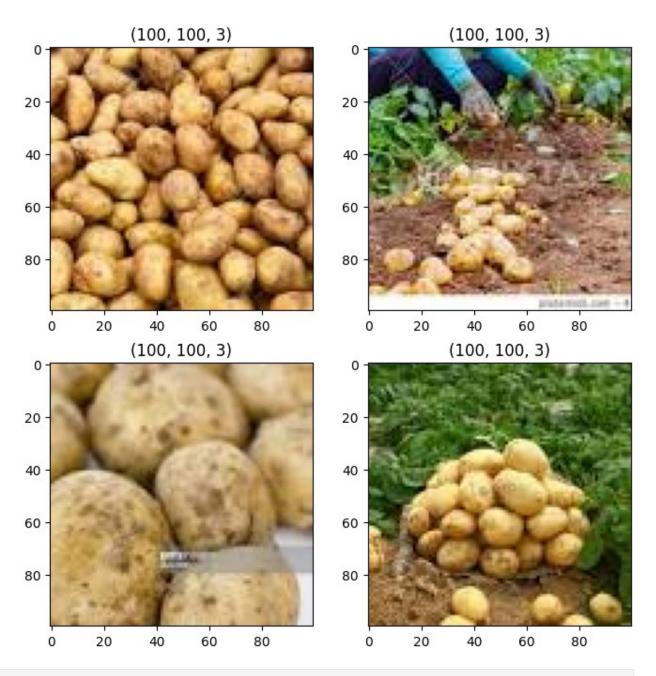
Images from the tomato folder



Ploting Images from Test Folder

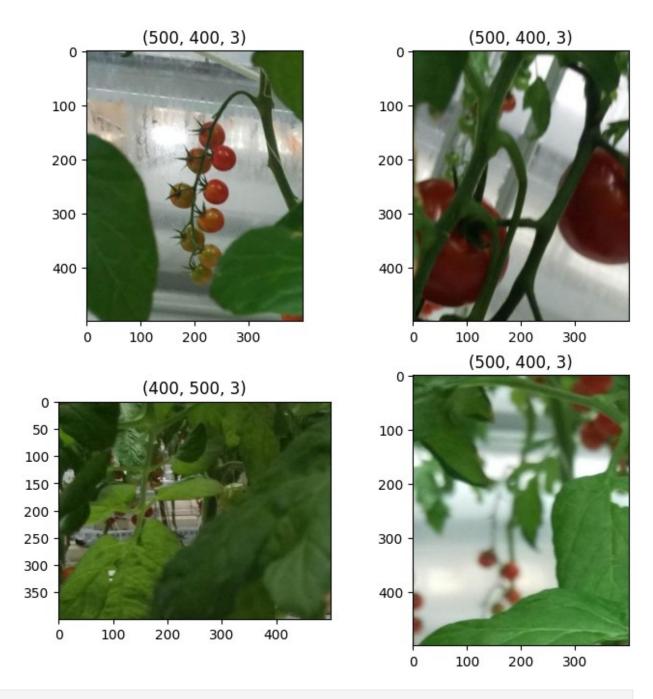
plot_images(test_data_path, "potato", "Images from the potato Test folder")

Images from the potato Test folder



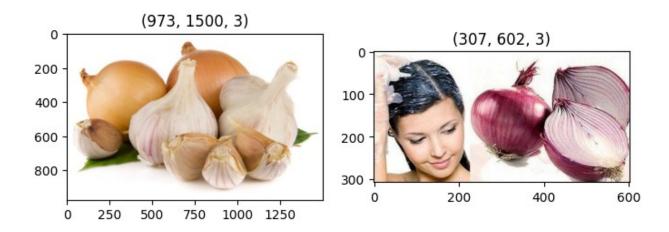
plot_images(test_data_path,"tomato","Images from the tomato Test
folder")

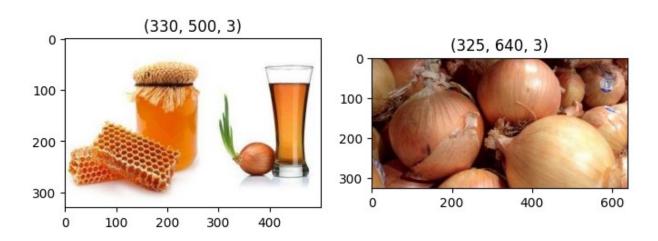
Images from the tomato Test folder



plot_images(test_data_path,"onion","Images from the onion Test
folder")

Images from the onion Test folder





plot_images(test_data_path,"indian market","Images from the indian
market Test folder")

Images from the indian market Test folder



Loading Training and Test Data

```
input_shape = (128,128)
train_data =
tf.keras.utils.image_dataset_from_directory(train_data_path,image_size
=input_shape,batch_size=64,)
test_data =
```

```
tf.keras.utils.image_dataset_from_directory(test_data_path,image_size=
input_shape,batch_size=64)
Found 3135 files belonging to 4 classes.
Found 351 files belonging to 4 classes.

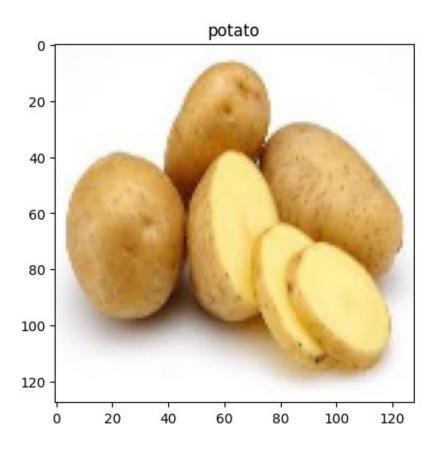
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Rescaling,RandomFlip,RandomTranslation,RandomRotation,RandomZoom,Rando
mContrast,RandomBrightness
```

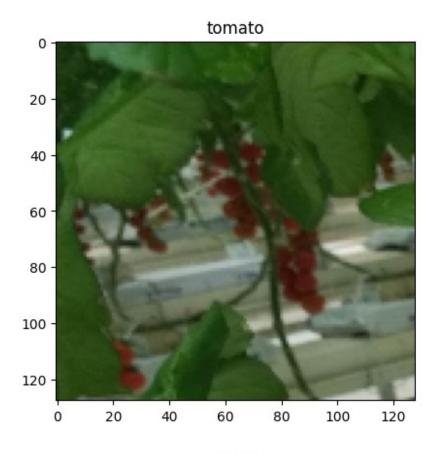
Data Preprocessing and Data augumentation

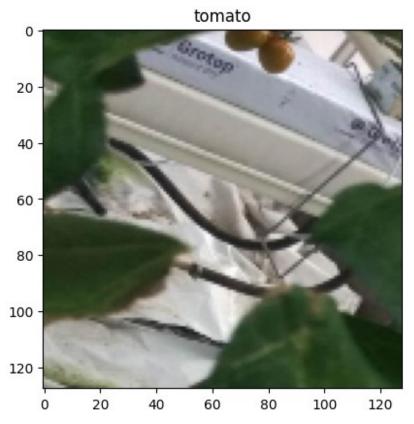
```
preproces image = Sequential([
    Rescaling(scale=1./255)
])
train data = train data.map(lambda x,y :
(preproces image(x,training=True),y),num parallel calls =
tf.data.AUTOTUNE)
test data = test data.map(lambda x,y :
(preproces image(x,training=True),y),num parallel calls =
tf.data.AUTOTUNE)
len(train data)
49
# image augumentation = Sequential([
      RandomFlip("horizontal"),
      RandomFlip("vertical"),
#
      RandomRotation(factor=0.5),
#
      RandomZoom(height factor=-0.1, width factor=-0.1)
# 1)
\# train ds = train data.map(lambda x, y :
(image augumentation(x,training=True),y),num parallel calls
=tf.data.AUTOTUNE)
val data = train data.take(15)
train ds = train data.skip(15)
print(tf.data.experimental.cardinality(train ds))
tf.Tensor(34, shape=(), dtype=int64)
images,label = next(iter(train ds))
reference_mapping label = {
    2: 'potato',
    3: 'tomato',
```

```
0:'indian market',
1:'onion'
}

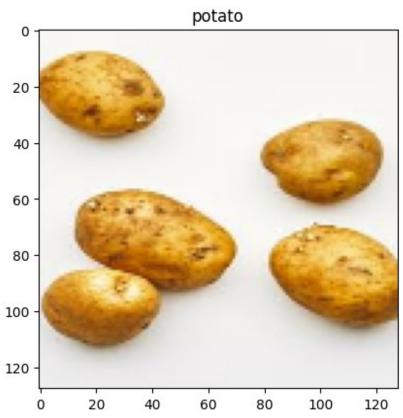
for img,label in zip(images.numpy()[:10],label.numpy()[:10]):
    plt.title(reference_mapping_label[label])
    # plt.title(label)
    plt.imshow(img)
    plt.show()
```

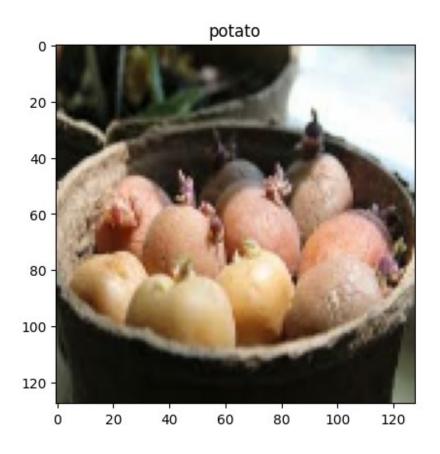




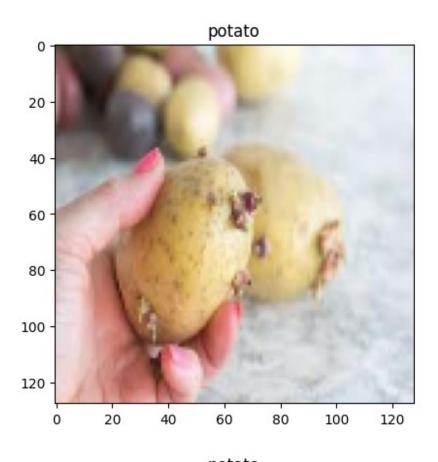


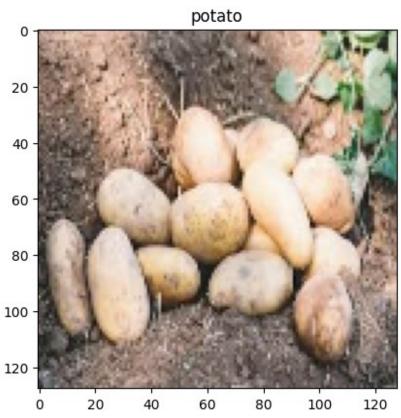


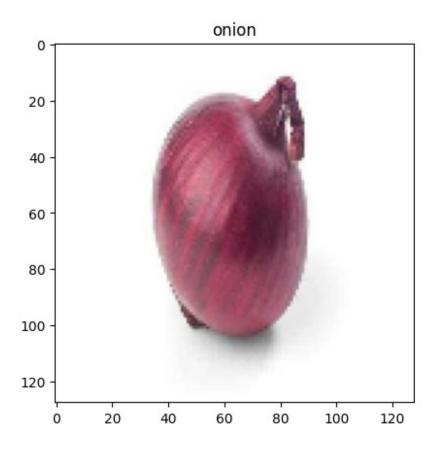












Scratch Implementation of CNN Model

```
Dropout (0.2),
Conv2D(filters=64, kernel size=3, activation='relu', padding="same", kerne
l regularizer=regularizers.L2(1e-4)),
Conv2D(filters=64, kernel size=3, activation='relu', padding="same", kernel
l regularizer=regularizers.L2(1e-4)),
    MaxPooling2D(),
    Dropout (0.3),
Conv2D(filters=128, kernel_size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
Conv2D(filters=128, kernel size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
    MaxPooling2D(),
    Dropout (0.2),
Conv2D(filters=256, kernel size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
Conv2D(filters=256, kernel size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
    MaxPooling2D(),
    Dropout (0.2),
Conv2D(filters=512, kernel size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
Conv2D(filters=512, kernel size=3, activation='relu', padding="same", kern
el regularizer=regularizers.L2(1e-4)),
    MaxPooling2D(),
    Dropout (0.2),
    GlobalAveragePooling2D(),
    Dense(units=4,activation='softmax')
])
model.summary()
Model: "sequential 1"
                              Output Shape
Layer (type)
                                                         Param #
                              (None, 128, 128, 16)
 conv2d (Conv2D)
                                                         448
```

conv2d_1 (Conv2D)	(None, 126, 126, 16)	2320
dropout (Dropout)	(None, 126, 126, 16)	0
conv2d_2 (Conv2D)	(None, 126, 126, 32)	4640
conv2d_3 (Conv2D)	(None, 126, 126, 32)	9248
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 63, 63, 32)	0
dropout_1 (Dropout)	(None, 63, 63, 32)	Θ
conv2d_4 (Conv2D)	(None, 63, 63, 64)	18496
conv2d_5 (Conv2D)	(None, 63, 63, 64)	36928
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(None, 31, 31, 64)	0
dropout_2 (Dropout)	(None, 31, 31, 64)	0
conv2d_6 (Conv2D)	(None, 31, 31, 128)	73856
conv2d_7 (Conv2D)	(None, 31, 31, 128)	147584
<pre>max_pooling2d_2 (MaxPooling 2D)</pre>	(None, 15, 15, 128)	0
dropout_3 (Dropout)	(None, 15, 15, 128)	0
conv2d_8 (Conv2D)	(None, 15, 15, 256)	295168
conv2d_9 (Conv2D)	(None, 15, 15, 256)	590080
<pre>max_pooling2d_3 (MaxPooling 2D)</pre>	(None, 7, 7, 256)	0
dropout_4 (Dropout)	(None, 7, 7, 256)	0
conv2d_10 (Conv2D)	(None, 7, 7, 512)	1180160
conv2d_11 (Conv2D)	(None, 7, 7, 512)	2359808
<pre>max_pooling2d_4 (MaxPooling 2D)</pre>	(None, 3, 3, 512)	0
dropout_5 (Dropout)	(None, 3, 3, 512)	0

```
0
global average pooling2d (G (None, 512)
lobalAveragePooling2D)
dense (Dense)
                            (None, 4)
                                                     2052
Total params: 4,720,788
Trainable params: 4,720,788
Non-trainable params: 0
from tensorflow.keras.callbacks import EarlyStopping,ModelCheckpoint
model.compile(optimizer='adam',
               loss='sparse categorical crossentropy',
               metrics=['accuracy'])
filepath =
"C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models//"
callbacks = [
       tf.keras.callbacks.ReduceLROnPlateau(
           monitor="val loss", factor=0.3, patience=3, min lr=0.00001
       tf.keras.callbacks.ModelCheckpoint(filepath,
monitor='val accuracy', mode='max', save best only=True),
       tf.keras.callbacks.EarlyStopping(
           monitor="val loss", patience=3, min delta=0.001,
mode='min'
   1
tf.keras.backend.clear session()
hist =
model.fit(train ds, validation data=val data, verbose=1, epochs=60, batch
size=32, callbacks=callbacks)
Epoch 1/60
accuracy: 0.2602
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
```

```
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.2602 - val loss: 1.4708 - val accuracy: 0.2677 - lr:
0.0010
Epoch 2/60
- accuracy: 0.2717 - val loss: 1.4456 - val accuracy: 0.2646 - lr:
0.0010
Epoch 3/60
accuracy: 0.3149
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.3149 - val loss: 1.3761 - val_accuracy: 0.4187 - lr:
0.0010
Epoch 4/60
accuracy: 0.4782
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
```

```
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.4782 - val loss: 0.9694 - val accuracy: 0.5344 - lr:
0.0010
Epoch 5/60
accuracy: 0.5375
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 12). These functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.5375 - val loss: 1.0074 - val accuracy: 0.5698 - lr:
0.0010
Epoch 6/60
accuracy: 0.6106
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.6106 - val loss: 0.8137 - val accuracy: 0.6396 - lr:
0.0010
Epoch 7/60
```

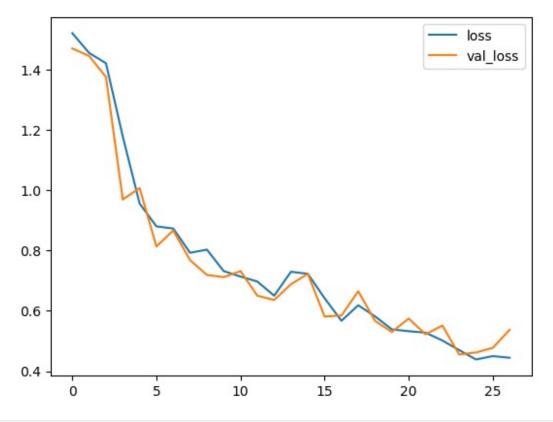
```
- accuracy: 0.6078 - val loss: 0.8669 - val accuracy: 0.5948 - lr:
0.0010
Epoch 8/60
accuracy: 0.6317
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
__jit_compiled_convolution_op, __jit_compiled_convolution_op, __jit_compiled_convolution_op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.6317 - val loss: 0.7682 - val accuracy: 0.6438 - lr:
0.0010
Epoch 9/60
accuracy: 0.6487
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit_compiled_convolution_op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.6487 - val loss: 0.7193 - val accuracy: 0.6865 - lr:
0.0010
Epoch 10/60
- accuracy: 0.6874 - val loss: 0.7117 - val accuracy: 0.6750 - lr:
0.0010
Epoch 11/60
- accuracy: 0.6947 - val loss: 0.7320 - val accuracy: 0.6635 - lr:
```

```
0.0010
Epoch 12/60
accuracy: 0.7140
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op while saving (showing 5 of 12). These functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.7140 - val loss: 0.6503 - val accuracy: 0.7240 - lr:
0.0010
Epoch 13/60
- accuracy: 0.7361 - val loss: 0.6361 - val accuracy: 0.7146 - lr:
0.0010
Epoch 14/60
accuracy: 0.6791
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
__jit_compiled_convolution_op, __jit_compiled_convolution_op, __jit_compiled_convolution_op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.6791 - val loss: 0.6884 - val accuracy: 0.7302 - lr:
0.0010
Epoch 15/60
- accuracy: 0.6952 - val loss: 0.7227 - val accuracy: 0.6573 - lr:
0.0010
```

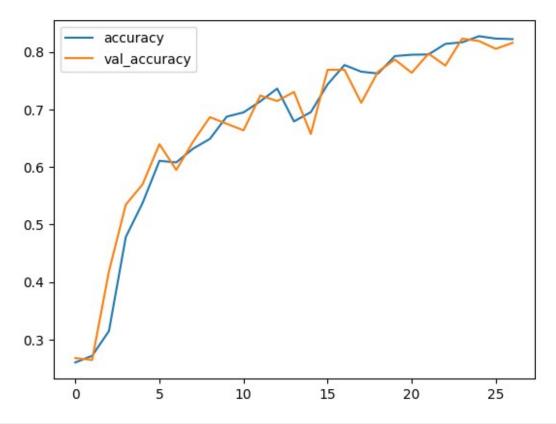
```
Epoch 16/60
accuracy: 0.7434
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.7434 - val loss: 0.5810 - val accuracy: 0.7688 - lr:
0.0010
Epoch 17/60
- accuracy: 0.7770 - val loss: 0.5846 - val accuracy: 0.7688 - lr:
0.0010
Epoch 18/60
- accuracy: 0.7655 - val loss: 0.6650 - val accuracy: 0.7115 - lr:
0.0010
Epoch 19/60
- accuracy: 0.7623 - val loss: 0.5665 - val accuracy: 0.7656 - lr:
0.0010
Epoch 20/60
accuracy: 0.7926
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
```

```
- accuracy: 0.7926 - val loss: 0.5301 - val accuracy: 0.7865 - lr:
0.0010
Epoch 21/60
- accuracy: 0.7949 - val loss: 0.5746 - val accuracy: 0.7635 - lr:
0.0010
Epoch 22/60
accuracy: 0.7954
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
- accuracy: 0.7954 - val loss: 0.5222 - val accuracy: 0.7969 - lr:
0.0010
Epoch 23/60
- accuracy: 0.8138 - val loss: 0.5512 - val accuracy: 0.7760 - lr:
0.0010
Epoch 24/60
accuracy: 0.8166
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
jit compiled convolution op while saving (showing 5 of 12). These
functions will not be directly callable after loading.
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
INFO:tensorflow:Assets written to:
C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//Models\/assets
```

```
- accuracy: 0.8166 - val loss: 0.4552 - val accuracy: 0.8229 - lr:
0.0010
Epoch 25/60
- accuracy: 0.8271 - val loss: 0.4619 - val accuracy: 0.8188 - lr:
0.0010
Epoch 26/60
- accuracy: 0.8230 - val loss: 0.4773 - val accuracy: 0.8052 - lr:
0.0010
Epoch 27/60
- accuracy: 0.8221 - val loss: 0.5372 - val accuracy: 0.8156 - lr:
0.0010
plt.plot(hist.history['loss'],label='loss')
plt.plot(hist.history['val loss'],label='val loss')
plt.leaend()
plt.show()
```



```
plt.plot(hist.history['accuracy'],label='accuracy')
plt.plot(hist.history['val_accuracy'],label='val_accuracy')
plt.legend()
plt.show()
```

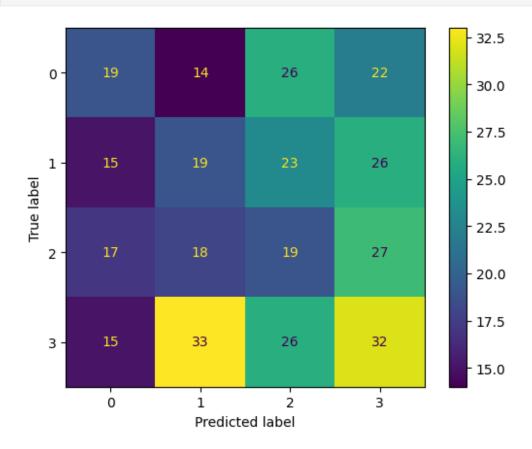


```
load model =
tf.keras.models.load model('C://Users//gaura//Documents//GitHub//Scale
r-Projects//Computer Vision//Project//Models')
# Evaluate the model
loss, acc = load_model.evaluate(test_data, verbose=2)
print("Restored model, accuracy: {:5.2f}%".format(100 * acc))
6/6 - 4s - loss: 0.5625 - accuracy: 0.7521 - 4s/epoch - 600ms/step
Restored model, accuracy: 75.21%
y_pred = load_model.predict(test_data)
6/6 [=======] - 1s 55ms/step
pred = np.argmax(y_pred,axis=1)
ytest = []
for img, label in test data:
    labels = label.numpy()
    ytest.extend(labels)
from sklearn.metrics import confusion matrix, ConfusionMatrixDisplay,
classification report
print(classification report(ytest,pred))
```

	precision	recall	f1-score	support
0 1	0.29 0.23	0.23 0.23	0.26 0.23	81 83
2 3	0.20 0.30	0.23 0.30	0.22 0.30	81 106
accuracy macro avg weighted avg	0.25 0.26	0.25 0.25	0.25 0.25 0.25	351 351 351

ConfusionMatrixDisplay(confusion matrix(ytest,pred)).plot()

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x225772bfeb0>

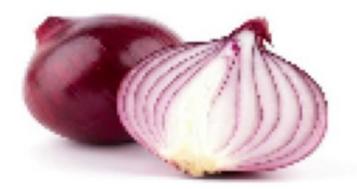


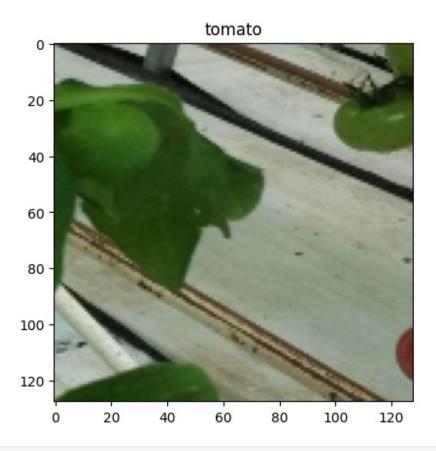
```
images,label = next(iter(test_data))
images = images.numpy()
label = label.numpy()

reference_mapping_label = {
    2:'potato',
    3:'tomato',
```

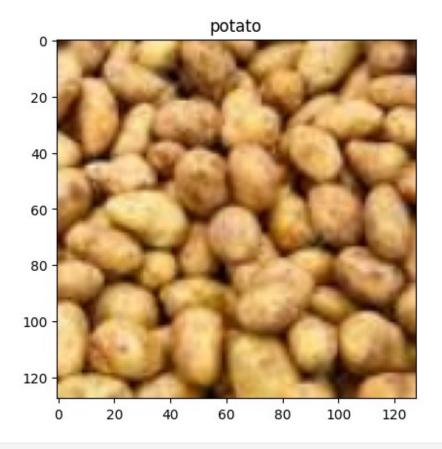
```
0:'indian market',
   1: 'onion'
}
CLASS NAMES = ['indian market', 'onion', 'potato', 'tomato']
for image, label in test data.take(2):
      plt.imshow(image[2])
      plt.title(CLASS_NAMES[label[2].numpy()])
      plt.axis('off')
prediction = load model.predict(test data.take(2))[2]
prediction = np.argmax(prediction)
#print(prediction)
scores = [1 - prediction, prediction]
for score, name in zip(scores, CLASS NAMES):
   print("This image is %.2f percent %s" % ((100 * score), name))
%time
2/2 [=======] - 0s 149ms/step
This image is -200.00 percent indian market
This image is 300.00 percent onion
CPU times: total: 0 ns
Wall time: 0 ns
```

onion

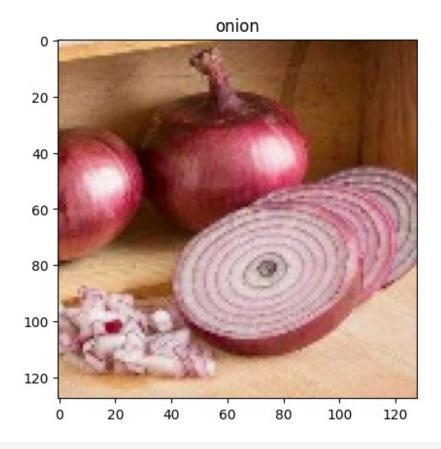




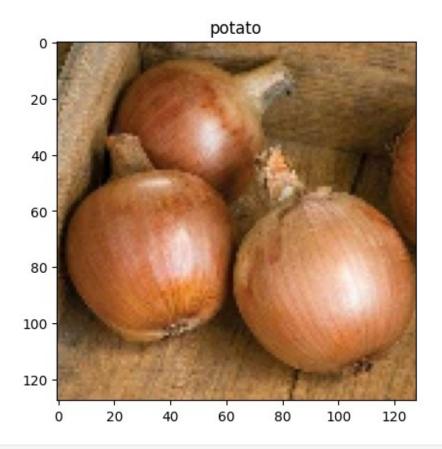
1/1 [======] - 0s 34ms/step



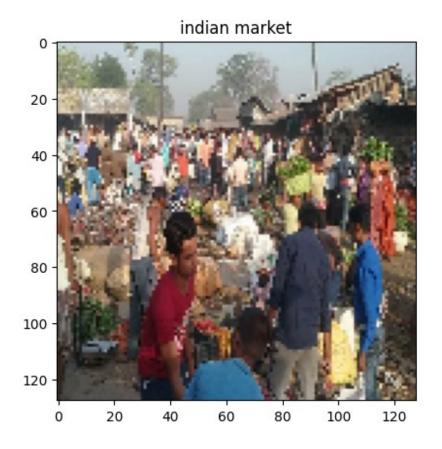
1/1 [======] - 0s 31ms/step



1/1 [======] - 0s 40ms/step



1/1 [======] - 0s 40ms/step



Applying Transfer Learning Method

```
base model =
tf.keras.applications.ResNet50(input shape=(128,128,3),include top=Fal
se,weights='imagenet')
for layers in base_model.layers[-1:-1000:-1]:
    #print(layers, layers.trainable)
    layers.trainable = False
inputs = tf.keras.Input(shape=(128,128,3))
x = base model(inputs,training=False)
x = tf.keras.layers.GlobalAveragePooling2D()(x)
outputs = tf.keras.layers.Dense(4,activation='softmax')(x)
n model = tf.keras.Model(inputs,outputs)
# for layers in base model.layers:
      if(layers.trainable == False):
          print(layers, layers.trainable)
# for layers in base model.layers:
      print(layers, layers.trainable)
```

```
n model.summary()
Model: "model"
Layer (type)
                           Output Shape
                                                   Param #
                           [(None, 128, 128, 3)]
input_2 (InputLayer)
 resnet50 (Functional)
                           (None, 4, 4, 2048)
                                                   23587712
global average pooling2d (G (None, 2048)
                                                   0
lobalAveragePooling2D)
dense (Dense)
                           (None, 4)
                                                   8196
______
Total params: 23,595,908
Trainable params: 8,196
Non-trainable params: 23,587,712
n model.compile(optimizer=tf.keras.optimizers.Adam(),
               loss='sparse categorical crossentropy',
               metrics=['accuracy'])
filepath =
"C://Users//gaura//Documents//GitHub//Scaler-Projects//Computer
Vision//Project//TL_MODEL"
callbacks = [
       tf.keras.callbacks.ReduceLROnPlateau(
           monitor="val loss", factor=0.3, patience=5, min lr=0.00001
       tf.keras.callbacks.ModelCheckpoint(filepath,
save weights only=True, monitor='val accuracy', mode='max',
save best only=True),
       tf.keras.callbacks.EarlyStopping(
           monitor="val loss", patience=10, min delta=0.001,
mode='min'
tf.keras.backend.clear session()
t hist =
n model.fit(train ds, validation data=val data, verbose=1, epochs=100, bat
ch_size=32,callbacks=callbacks)
Epoch 1/100
- accuracy: 0.2855 - val loss: 1.3536 - val accuracy: 0.2865 - lr:
```

```
0.0010
Epoch 2/100
- accuracy: 0.3131 - val loss: 1.3322 - val accuracy: 0.3729 - lr:
0.0010
Epoch 3/100
- accuracy: 0.3756 - val loss: 1.3173 - val accuracy: 0.3646 - lr:
0.0010
Epoch 4/100
- accuracy: 0.3825 - val loss: 1.3040 - val accuracy: 0.3625 - lr:
0.0010
Epoch 5/100
- accuracy: 0.4014 - val loss: 1.2867 - val accuracy: 0.3760 - lr:
0.0010
Epoch 6/100
- accuracy: 0.4230 - val loss: 1.2723 - val accuracy: 0.3833 - lr:
0.0010
Epoch 7/100
- accuracy: 0.4299 - val loss: 1.2561 - val accuracy: 0.4573 - lr:
0.0010
Epoch 8/100
- accuracy: 0.4667 - val loss: 1.2540 - val accuracy: 0.4354 - lr:
0.0010
Epoch 9/100
- accuracy: 0.4749 - val loss: 1.2362 - val accuracy: 0.4490 - lr:
0.0010
Epoch 10/100
- accuracy: 0.4464 - val loss: 1.2383 - val accuracy: 0.4292 - lr:
0.0010
Epoch 11/100
- accuracy: 0.4892 - val loss: 1.2097 - val accuracy: 0.4521 - lr:
0.0010
Epoch 12/100
- accuracy: 0.5076 - val loss: 1.1983 - val accuracy: 0.4833 - lr:
0.0010
Epoch 13/100
- accuracy: 0.5053 - val loss: 1.2043 - val accuracy: 0.4708 - lr:
0.0010
```

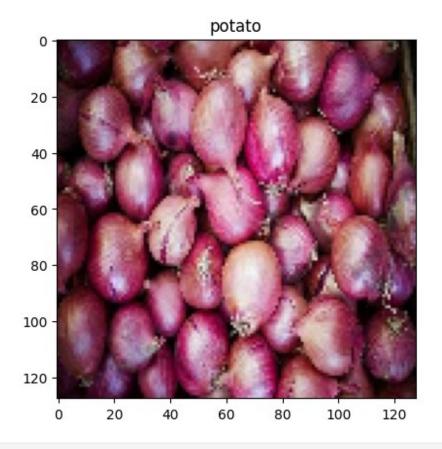
```
Epoch 14/100
- accuracy: 0.5159 - val loss: 1.1924 - val accuracy: 0.4875 - lr:
0.0010
Epoch 15/100
- accuracy: 0.5076 - val loss: 1.1900 - val accuracy: 0.4812 - lr:
0.0010
Epoch 16/100
34/34 [============== ] - 9s 223ms/step - loss: 1.1714
- accuracy: 0.4970 - val loss: 1.1774 - val accuracy: 0.5052 - lr:
0.0010
Epoch 17/100
- accuracy: 0.5168 - val loss: 1.1684 - val accuracy: 0.5104 - lr:
0.0010
Epoch 18/100
- accuracy: 0.5333 - val loss: 1.1629 - val accuracy: 0.5000 - lr:
0.0010
Epoch 19/100
- accuracy: 0.5255 - val loss: 1.1693 - val accuracy: 0.4792 - lr:
0.0010
Epoch 20/100
- accuracy: 0.5343 - val loss: 1.1596 - val_accuracy: 0.4927 - lr:
0.0010
Epoch 21/100
- accuracy: 0.5434 - val_loss: 1.1515 - val_accuracy: 0.5260 - lr:
0.0010
Epoch 22/100
- accuracy: 0.5554 - val loss: 1.1380 - val accuracy: 0.5281 - lr:
0.0010
Epoch 23/100
- accuracy: 0.5490 - val loss: 1.1341 - val accuracy: 0.5156 - lr:
0.0010
Epoch 24/100
- accuracy: 0.5586 - val_loss: 1.1309 - val_accuracy: 0.5167 - lr:
0.0010
Epoch 25/100
- accuracy: 0.5582 - val loss: 1.1187 - val accuracy: 0.5188 - lr:
0.0010
Epoch 26/100
```

```
- accuracy: 0.5453 - val loss: 1.1226 - val accuracy: 0.5375 - lr:
0.0010
Epoch 27/100
- accuracy: 0.5683 - val loss: 1.1190 - val accuracy: 0.5292 - lr:
0.0010
Epoch 28/100
- accuracy: 0.5379 - val loss: 1.1167 - val accuracy: 0.5271 - lr:
0.0010
Epoch 29/100
- accuracy: 0.5678 - val loss: 1.1130 - val accuracy: 0.5281 - lr:
0.0010
Epoch 30/100
- accuracy: 0.5623 - val loss: 1.1104 - val accuracy: 0.5198 - lr:
0.0010
Epoch 31/100
- accuracy: 0.5724 - val loss: 1.1021 - val accuracy: 0.5344 - lr:
0.0010
Epoch 32/100
- accuracy: 0.5807 - val loss: 1.1107 - val accuracy: 0.5177 - lr:
0.0010
Epoch 33/100
- accuracy: 0.5747 - val loss: 1.0952 - val accuracy: 0.5500 - lr:
0.0010
Epoch 34/100
- accuracy: 0.5802 - val loss: 1.0988 - val accuracy: 0.5312 - lr:
0.0010
Epoch 35/100
- accuracy: 0.5775 - val loss: 1.0749 - val accuracy: 0.5531 - lr:
0.0010
Epoch 36/100
- accuracy: 0.5816 - val loss: 1.0812 - val accuracy: 0.5396 - lr:
0.0010
Epoch 37/100
- accuracy: 0.5743 - val_loss: 1.0912 - val_accuracy: 0.5312 - lr:
0.0010
Epoch 38/100
```

```
- accuracy: 0.5793 - val loss: 1.0860 - val accuracy: 0.5375 - lr:
0.0010
Epoch 39/100
- accuracy: 0.5802 - val loss: 1.0759 - val accuracy: 0.5500 - lr:
0.0010
Epoch 40/100
- accuracy: 0.5867 - val loss: 1.0761 - val accuracy: 0.5635 - lr:
0.0010
Epoch 41/100
- accuracy: 0.5940 - val loss: 1.0767 - val accuracy: 0.5469 - lr:
3.0000e-04
Epoch 42/100
- accuracy: 0.5954 - val loss: 1.0771 - val_accuracy: 0.5479 - lr:
3.0000e-04
Epoch 43/100
- accuracy: 0.5991 - val loss: 1.0655 - val accuracy: 0.5573 - lr:
3.0000e-04
Epoch 44/100
- accuracy: 0.5913 - val loss: 1.0743 - val accuracy: 0.5490 - lr:
3.0000e-04
Epoch 45/100
- accuracy: 0.5903 - val loss: 1.0703 - val accuracy: 0.5708 - lr:
3.0000e-04
Epoch 46/100
- accuracy: 0.5977 - val loss: 1.0683 - val accuracy: 0.5510 - lr:
3.0000e-04
Epoch 47/100
- accuracy: 0.5908 - val loss: 1.0564 - val accuracy: 0.5792 - lr:
3.0000e-04
Epoch 48/100
- accuracy: 0.5949 - val loss: 1.0747 - val accuracy: 0.5437 - lr:
3.0000e-04
Epoch 49/100
- accuracy: 0.5926 - val loss: 1.0637 - val accuracy: 0.5583 - lr:
3.0000e-04
Epoch 50/100
- accuracy: 0.6014 - val loss: 1.0674 - val accuracy: 0.5698 - lr:
```

```
3.0000e-04
Epoch 51/100
- accuracy: 0.5899 - val loss: 1.0623 - val accuracy: 0.5625 - lr:
3.0000e-04
Epoch 52/100
- accuracy: 0.5931 - val loss: 1.0768 - val accuracy: 0.5510 - lr:
3.0000e-04
Epoch 53/100
- accuracy: 0.5982 - val loss: 1.0484 - val accuracy: 0.5781 - lr:
9.0000e-05
Epoch 54/100
- accuracy: 0.5926 - val loss: 1.0583 - val accuracy: 0.5635 - lr:
9.0000e-05
Epoch 55/100
- accuracy: 0.5986 - val loss: 1.0615 - val accuracy: 0.5615 - lr:
9.0000e-05
Epoch 56/100
- accuracy: 0.5917 - val loss: 1.0691 - val accuracy: 0.5594 - lr:
9.0000e-05
Epoch 57/100
- accuracy: 0.5926 - val loss: 1.0703 - val accuracy: 0.5531 - lr:
9.0000e-05
Epoch 58/100
- accuracy: 0.5945 - val loss: 1.0573 - val accuracy: 0.5698 - lr:
9.0000e-05
Epoch 59/100
- accuracy: 0.5926 - val loss: 1.0700 - val accuracy: 0.5500 - lr:
2.7000e-05
Epoch 60/100
- accuracy: 0.5963 - val loss: 1.0611 - val accuracy: 0.5635 - lr:
2.7000e-05
Epoch 61/100
- accuracy: 0.5926 - val loss: 1.0610 - val accuracy: 0.5615 - lr:
2.7000e-05
Epoch 62/100
- accuracy: 0.5899 - val loss: 1.0579 - val accuracy: 0.5625 - lr:
2.7000e-05
Epoch 63/100
```

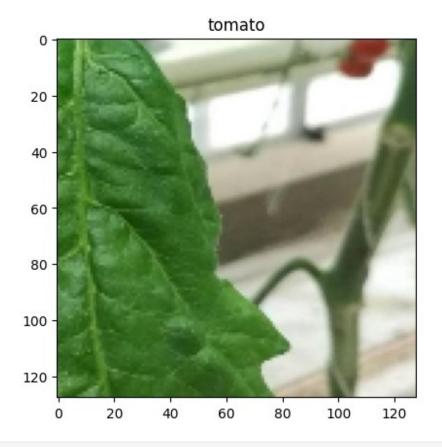
```
- accuracy: 0.5995 - val loss: 1.0613 - val accuracy: 0.5615 - lr:
2.7000e-05
# n model.save('tl model.pb')
WARNING:absl:Found untraced functions such as
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
_jit_compiled_convolution_op, _jit_compiled_convolution_op,
__jit_compiled_convolution_op while saving (showing 5 of 53). These functions will not be directly callable after loading.
INFO:tensorflow:Assets written to: tl model.pb\assets
INFO:tensorflow:Assets written to: tl model.pb\assets
load model =
tf.keras.models.load model("C://Users//gaura//Documents//GitHub//Scale
r-Projects//Computer Vision//Project//Notebook//tl model.pb")
# Evaluate the model
loss, acc = load model.evaluate(test data, verbose=2)
print("Restored model, accuracy: {:5.2f}%".format(100 * acc))
6/6 - 2s - loss: 1.1665 - accuracy: 0.4644 - 2s/epoch - 338ms/step
Restored model, accuracy: 46.44%
images,label = next(iter(test data))
for img in images.numpy()[:5]:
   #print(img.shape)
   temp = (np.expand dims(imq,axis=0))
   pred = load model.predict(temp)
    labels = reference mapping label[np.argmax(pred)]
   plt.imshow(img)
   plt.title(labels)
   plt.show()
```



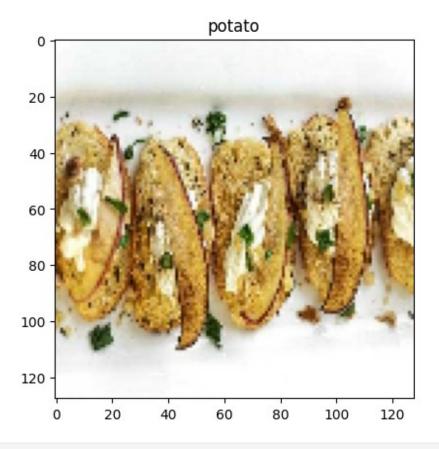
1/1 [======] - 0s 33ms/step



1/1 [=======] - 0s 24ms/step



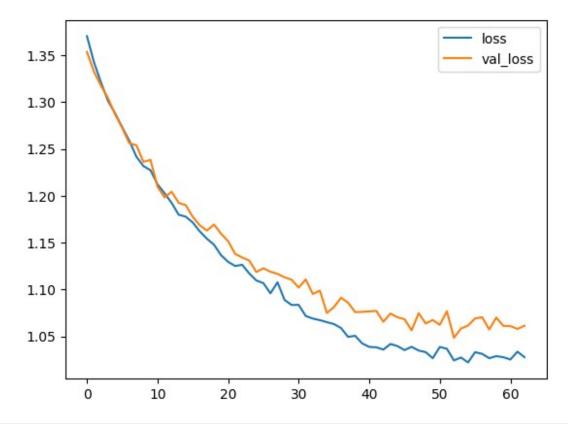
1/1 [======] - 0s 26ms/step



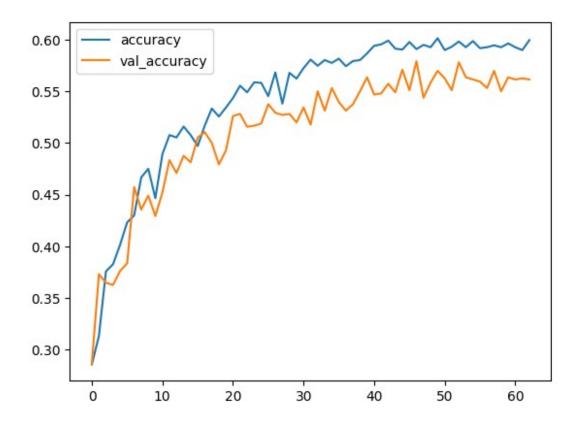
1/1 [======] - 0s 32ms/step



```
plt.plot(t_hist.history['loss'],label='loss')
plt.plot(t_hist.history['val_loss'],label='val_loss')
plt.legend()
plt.show()
```



```
plt.plot(t_hist.history['accuracy'],label='accuracy')
plt.plot(t_hist.history['val_accuracy'],label='val_accuracy')
plt.legend()
plt.show()
```



Summary of Model

- 1. Both the model have there own way of generalization.
- 2. CNN from scratch has achived accuracy of 75.21% but was unable to learn the params and end up with missclassification.
- 3. Using Transfer learning the model has achived accuracy of 46% accuracy on test but has less missclassification problems.

```
! pip install nb-pdf-template
Collecting nb-pdf-template
  Downloading nb_pdf_template-4.0.0.tar.gz (5.6 kB)
  Preparing metadata (setup.py): started
  Preparing metadata (setup.py): finished with status 'done'
Building wheels for collected packages: nb-pdf-template
  Building wheel for nb-pdf-template (setup.py): started
  Building wheel for nb-pdf-template (setup.py): finished with status
'done'
  Created wheel for nb-pdf-template: filename=nb_pdf_template-4.0.0-
py3-none-any.whl size=6794
sha256=2b9974e5f8cd09c25fdc7878d5380d9447572157352e135be2c5dec591f47a4
a
  Stored in directory: c:\users\gaura\appdata\local\pip\cache\wheels\
5c\8c\c4\c70aa7f988af2d421bbd58610f5ae777babda912999bb2acb1
```

Successfully built nb-pdf-template Installing collected packages: nb-pdf-template Successfully installed nb-pdf-template-4.0.0