```
In []: from tensorflow.keras.models import Sequential
            from tensorflow.keras.layers import Conv2D, MaxPooling2D, Dropout, Flatten, Dense
            from tensorflow.keras.optimizers import Adam
            from tensorflow.keras.preprocessing.image import ImageDataGenerator
            import numpy as np
            import matplotlib.pyplot as plt
            import os
In [ ]: data = '../input/covid-face-mask-detection-dataset/New Masks Dataset'
            train = os.path.join(data, 'Train')
            test = os.path.join(data, 'Test')
            valid = os.path.join(data, 'Validation')
            train mask = os.path.join(train, 'Mask')
            train_nomask = os.path.join(train,'Non Mask')
            train mask = os.listdir(train mask)
            print(train mask[15:25])
            train_nomask_names = os.listdir(train_nomask)
            print(train_nomask_names[15:25])
            ['0372.jpg', '1548.jpg', '0932.jpg', '0006.jpg', '0559.jpg', '1473.jpg', '1669.jpg', '0862.jpg', '1058.jpg', '1
            166.jpg']
            ['215.jpg', '185.jpg', '243.jpg', '153.jpg', '143.jpg', '327.jpg', '253.jpg', '115.jpg', '131.jpg', '151.jpg']
In [ ]: import matplotlib.image as mpimg
            n rows = 4
            ncols = 4
            plt.figure(figsize=(12,12))
            mask_picture = []
            for i in train_mask[10:20]:
              mask_picture.append(os.path.join(train_mask,i))
            nomask_pic = []
            for i in train nomask names[10:20]:
              nomask pic.append(os.path.join(train nomask,i))
            print(mask_picture)
            print(nomask_pic)
            merged pics = mask picture+nomask pic
            for i in range(0,len(merged pics)):
               data = merged_pics[i].split('/',4)[4]
               sp = plt.subplot(4,4,i+1)
               sp.axis('Off')
               image = mpimg.imread(merged_pics[i])
               sp.set_title(data,fontsize=10)
               plt.imshow(image,cmap='gray')
            plt.show()
            ['../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Mask/1037.png', '../input/covid-face-mask-
            detection-dataset/New Masks Dataset/Train/Mask/1488.jpg', '../input/covid-face-mask-detection-dataset/New Masks
            Dataset/Train/Mask/1539.jpg', '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Mask/1409.jpg
               '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Mask/1154.jpg', '../input/covid-face-mas
            k-detection-dataset/New Masks Dataset/Train/Mask/0372.jpg', '../input/covid-face-mask-detection-dataset/New Mas
            ks Dataset/Train/Mask/1548.jpg', '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Mask/0932.
            jpg', \ '.../input/covid-face-mask-detection-dataset/New\ Masks\ Dataset/Train/Mask/0006.jpg', \ '.../input/covid-face-mask-detection-dataset/New\ Masks\ Dataset/New\ Masks\ Datas
            mask-detection-dataset/New Masks Dataset/Train/Mask/0559.jpg']
            ['../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Non Mask/109.jpg', '../input/covid-face-ma
            sk-detection-dataset/New Masks Dataset/Train/Non Mask/149.jpg', '../input/covid-face-mask-detection-dataset/New
            Masks Dataset/Train/Non Mask/187.jpg', '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Non
            Mask/76.jpg', '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Non Mask/270.jpg', '../input/
            covid-face-mask-detection-dataset/New Masks Dataset/Train/Non Mask/215.jpg', '../input/covid-face-mask-detectio
            n-dataset/New Masks Dataset/Train/Non Mask/185.jpg', '../input/covid-face-mask-detection-dataset/New Masks Data
            set/Train/Non Mask/243.jpg', '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Non Mask/153.j
                    '../input/covid-face-mask-detection-dataset/New Masks Dataset/Train/Non Mask/143.jpg']
                                                                          Traceback (most recent call last)
            /tmp/ipykernel 23/282198932.py in <module>
                   19 for i in range(0,len(merged_pics)):
                   20 data = merged_pics[i].split('/',4)[4]
                          sp = plt.subplot(4,4,i+1)
            ---> 21
                   22
                           sp.axis('Off'
                          image = mpimg.imread(merged pics[i])
            /opt/conda/lib/python3.7/site-packages/matplotlib/pyplot.py in subplot(*args, **kwargs)
                1285
                1286
                              # First, search for an existing subplot with a matching spec.
             -> 1287
                              key = SubplotSpec._from_subplot_args(fig, args)
                1288
                              for ax in fig.axes:
                1289
            /opt/conda/lib/python3.7/site-packages/matplotlib/gridspec.py in _from_subplot_args(figure, args)
                  607
                                         if not isinstance(num, Integral) or num < 1 or num > rows*cols:
                  608
                                                raise ValueError(
                                                      f"num must be 1 <= num <= {rows*cols}, not {num!r}")
            --> 609
                  610
                                          i = j = num
                                   return gs[i-1:j]
                  611
            ValueError: num must be 1 <= num <= 16, not 17</pre>
```

Train/Mask/1037.png





Train/Mask/0006.jpg



Train/Non Mask/187.jpg



Train/Mask/1488.jpg



Train/Mask/0372.jpg



Train/Mask/0559.jpg



Train/Non Mask/76.jpg



Train/Mask/1539.jpg



Train/Mask/1548.jpg



Train/Non Mask/109.jpg



Train/Non Mask/270.jpg





Train/Mask/0932.jpg



Train/Non Mask/149.jpg



Train/Non Mask/215.jpg



```
data_train = ImageDataGenerator(rescale=1./255, zoom_range = 0.2,rotation_range = 40,horizontal_flip = True)
In [ ]:
        data test = ImageDataGenerator(rescale=1./255)
        valid data = ImageDataGenerator(rescale=1./255)
        train generator = data train.flow from directory(train, target size=(150,150), batch size = 32, class mode = 'bina'
        test generator = data test.flow from directory(test, target size=(150,150), batch size = 32, class mode = 'binary'
        valid_generator = valid_data.flow_from_directory(valid, target_size=(150,150), batch_size = 32, class_mode = 'bina'
        Found 600 images belonging to 2 classes.
```

```
Found 100 images belonging to 2 classes.
        Found 306 images belonging to 2 classes.
In [ ]:
        model = Sequential()
        model.add(Conv2D(32,(3,3),padding='SAME',activation='relu',input_shape=(150,150,3)))
        model.add(MaxPooling2D(pool_size=(2,2)))
        model.add(Dropout(0.5))
        model.add(Conv2D(64,(3,3),padding='SAME',activation='relu'))
        model.add(MaxPooling2D(pool_size=(2,2)))
        model.add(Dropout(0.5))
        model.add(Flatten())
        model.add(Dense(256,activation='relu'))
        model.add(Dropout(0.5))
        model.add(Dense(1,activation='sigmoid'))
        model.summary()
```

Output Shape

Layer (type)

In [ ]:

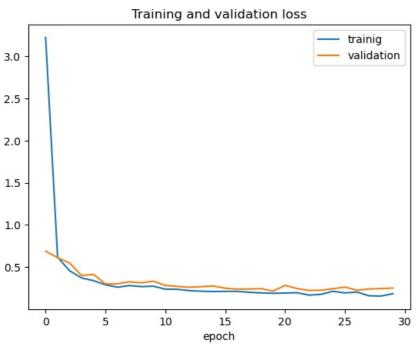
In [ ]:

Conv2d (Conv2D)	Layer (type)	output Sila	•	raralli #			
Argument					==		
conv2d_1 (conv2d) (None, 75, 75, 64) 18496  mav_pooling2d_1 (MaxPooling (None, 37, 37, 64) 0 20)  dropput_1 (Dropout) (None, 37, 37, 64) 0 flatten (Flatten) (None, 87616) 0  dense (Dense) (None, 256) 22429952  dropput_2 (Dropout) (None, 256) 0 dense_1 (Dense) (None, 1) 257  Total params: 22, 449, 601 Trainable params: 22, 449, 601 Non-trainable params: 0  model.complie(Adam(l=0.001),loss='binary_crossentropy',metrics=['accuracy']) history = model.fit(train_generator, epochs = 30, volidation_data = valid_generator)  Epoch 1/30  223-02-26_38:37;36.855618: tensorftow/core/grappler/optimizers/meta_optimizer.cc:954] layout failed: IN ARGUMENT: Size of values 0 does not match size of permutation 4 of farin shape insequential/dropout/dropout-cett22-TransposeNMMCTOMUN_apput_optimizers/meta_optimizers/ 0.5117 - val_loss: 0.6895  Epoch 1/30  19/19 [====================================	<pre>max_pooling2d (MaxPooling2D )</pre>	(None, 75	, 75, 32)	0			
max pooling2d_1 (MaxPooling (None, 37, 37, 64) 0  dropout 1 (Dropout) (None, 37, 37, 64) 0  flatten (Flatten) (None, 256) 0  dense (Dense) (None, 256) 22429952  dropout 2 (Dropout) (None, 256) 0  dense 1 (Dense) (None, 256) 0  dense 1 (Dense, 256) 0  d	dropout (Dropout)	(None, 75,	75, 32)	Θ			
dropout 1 (Dropout) (None, 37, 37, 64) 8  flatten (Flatten) (None, 37, 37, 64) 8  dense (Dense) (Mone, 256) 9  dense (Dropout) (None, 256) 9  dense 1 (Dense) (None, 256) 9  dense 1 (Dense 1	conv2d_1 (Conv2D)	(None, 75,	75, 64)	18496			
flatten (Flatten) (None, 27616) 0  dense (Dense) (None, 256) 22429952  dropout_2 (Dropout) (None, 256) 0  dense_1 (Dense) (None, 1) 257  Total params: 22,449,601 Trainable params: 22,449,601 Trainable params: 22,449,601 Non-trainable params: 20,449,601 Non-trainable params: 20,449,401 Non-trainab		(None, 37	, 37, 64)	0			
dense (Dense) (None, 256) 22429952  dropout_2 (Dropout) (None, 256) 8  dense_1 (Dense) (None, 1) 257	<pre>dropout_1 (Dropout)</pre>	(None, 37,	37, 64)	0			
dense_1 (Dense)	flatten (Flatten)	(None, 876	16)	Θ			
Total params: 22,449,601 Torial params: 22,449,601 Mon-trainable params: 22,449,601 Mon-trainable params: 2,449,601 Mon-traina	dense (Dense)	(None, 256	)	22429952			
Total params: 22,449,601 Non-trainable params: 22,449,601 Non-trainable params: 2,2449,601 Non-trainable params: 0  model.compile(Adam(tr=0.001),loss='binary_crossentropy',metrics=['accuracy']) history = model.fit(train_generator, epochs = 30, validation_data = valid_generator)  Epoch 1/30 2023-02-26 18/37:36.835618: E tensorflow/core/grappler/optimizers/meta_optimizer.cc:954] layout failed: IR ARGUNENT: Size of values 0 does not match size of permutation 4 @ fanin_shape_insequential/dropout/dropout cctv2_2-1ransposeNeWcToNCHM-LayoutOptimizer 19/19 [====================================	<pre>dropout_2 (Dropout)</pre>	(None, 256	)	Θ			
Trainable params: 22,449,601 Non-trainable params: 0  model.compile(Adam(lr=0.001),loss='binary_crossentropy',metrics=['accuracy']) history = model.fit(train_generator, epochs = 30,	dense_1 (Dense)	(None, 1)		257			
history = model.fit(train generator, epochs = 30, validation_data = valid_generator)	Trainable params: 22,449,601				==		
history = model.fit(train_generator, epochs = 30, validation_data = valid_generator)  Depoch 1/30  2023-02-26 18:37:36.855618: E tensorflow/core/grappler/optimizers/meta_optimizer.cc:954] layout failed: IN ARGUMENT: Size of values 0 does not match size of permutation 4 @ famin shape insequential/dropout/dropoutevt2-2-TransposeNHMCTOKHM-layoutDytimizer 19/19 [====================================							
. 2023-02.26 18:37:36.855618: E tensorflow/core/grappler/optimizers/meta optimizer.cc:954] layout failed: IM ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential/dropout/dropout/dropout/cc-educy: 0.5000 [	<pre>history = model.fit(train_ge</pre>	nerator, 30,			=['accuracy'])		
ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin shape insequential/dropout/dropoutectv2.2-transposeNHWCTONCHW-LayoutOptimizer 19/19 [====================================	Epoch 1/30						
accuracy: 0.5000 Epoch 2/30 19/19 [====================================	_ARGUMENT: Size of values 0 ectV2-2-TransposeNHWCToNCHW-	does not ma LayoutOptim	tch size of pe izer	rmutation 4	4 @ fanin shape in	sequential/dropo	ut/dropout/Sel
19/19 [====================================	_accuracy: 0.5000	======]	- 29s Is/step	- loss: 3.2	2225 - accuracy: 0	.511/ - val_loss	: 0.6895 - val
Epoch 3/30  19/19 [====================================	19/19 [====================================	=====]	- 14s 735ms/st	ep - loss:	0.6169 - accuracy	: 0.6583 - val_l	oss: 0.6118 -
val accuracy: 0.7876 Epoch 4/30 19/19 [====================================	Epoch 3/30	1	- 1/c 738mc/ct	en - lossi	0 4560 - accuracy	. A 915A - val l	oss: 0 5/06 -
val_accuracy: 0.8954 Epoch 5/30 19/19 [====================================	val_accuracy: 0.7876		143 /30113/36	.ер 10331	0.4303 accuracy	. 0.0130 Vac_c	0.5450
19/19 [====================================	val_accuracy: 0.8954	======]	- 15s 784ms/st	ep - loss:	0.3713 - accuracy	: 0.8650 - val_l	oss: 0.4006 -
Epoch 6/30 19/19 [====================================	19/19 [===========	=====]	- 14s 736ms/st	ep - loss:	0.3395 - accuracy	: 0.8733 - val_l	oss: 0.4136 -
Val_accuracy: 0.9020 Epoch 7/30 19/19 [====================================	Epoch 6/30	======1	- 14s 770ms/st	ep - loss:	0.2896 - accuracy	: 0.8950 - val l	oss: 0.3012 -
Val_accuracy: 0.8954 Epoch 8/30 19/19 [====================================	val_accuracy: 0.9020	-			•	_	
19/19 [====================================		=====]	- 14s 729ms/st	ep - loss:	0.2623 - accuracy	: 0.9117 - val_l	oss: 0.3023 -
Epoch 9/30  19/19 [====================================	19/19 [===========	=====]	- 14s 765ms/st	ep - loss:	0.2809 - accuracy	: 0.9000 - val_l	oss: 0.3261 -
Val_accuracy: 0.8889 Epoch 10/30 19/19 [====================================	Epoch 9/30			_			
19/19 [====================================	val_accuracy: 0.8889	=====]	- 14s 726ms/st	ep - loss:	0.2689 - accuracy	: 0.8983 - val_l	oss: 0.3134 -
Epoch 11/30  19/19 [====================================	19/19 [==========	=====]	- 14s 718ms/st	ep - loss:	0.2739 - accuracy	: 0.9117 - val_l	oss: 0.3328 -
<pre>val_accuracy: 0.8987 Epoch 12/30 19/19 [===========] - 14s 742ms/step - loss: 0.2369 - accuracy: 0.9167 - val_loss: 0.27 val_accuracy: 0.9118 Epoch 13/30 19/19 [==========] - 14s 766ms/step - loss: 0.2213 - accuracy: 0.9183 - val_loss: 0.26 val_accuracy: 0.9020 Epoch 14/30 19/19 [==========] - 14s 720ms/step - loss: 0.2143 - accuracy: 0.9200 - val_loss: 0.26 val_accuracy: 0.8987 Epoch 15/30 19/19 [===========] - 19s 1s/step - loss: 0.2107 - accuracy: 0.9233 - val_loss: 0.2756 accuracy: 0.8922</pre>	Epoch 11/30		14- 704 ( )		0.2200	0.0150	0 2021
19/19 [==========] - 14s 742ms/step - loss: 0.2369 - accuracy: 0.9167 - val_loss: 0.27 val_accuracy: 0.9118 Epoch 13/30   19/19 [==========] - 14s 766ms/step - loss: 0.2213 - accuracy: 0.9183 - val_loss: 0.26 val_accuracy: 0.9020   Epoch 14/30   19/19 [============] - 14s 720ms/step - loss: 0.2143 - accuracy: 0.9200 - val_loss: 0.26 val_accuracy: 0.8987   Epoch 15/30   19/19 [====================================	val_accuracy: 0.8987	======]	- 145 /b4MS/St	ep - LOSS:	พ.2388 - accuracy	: מ.פו_l	USS: U.2821 -
19/19 [==========] - 14s 766ms/step - loss: 0.2213 - accuracy: 0.9183 - val_loss: 0.26 val_accuracy: 0.9020  Epoch 14/30  19/19 [=========] - 14s 720ms/step - loss: 0.2143 - accuracy: 0.9200 - val_loss: 0.26 val_accuracy: 0.8987  Epoch 15/30  19/19 [=========] - 19s 1s/step - loss: 0.2107 - accuracy: 0.9233 - val_loss: 0.2756 _accuracy: 0.8922	19/19 [====================================	=====]	- 14s 742ms/st	ep - loss:	0.2369 - accuracy	: 0.9167 - val_l	oss: 0.2713 -
Epoch 14/30  19/19 [====================================	19/19 [===========	=====]	- 14s 766ms/st	ep - loss:	0.2213 - accuracy	: 0.9183 - val_l	oss: 0.2613 -
val_accuracy: 0.8987 Epoch 15/30 19/19 [====================================	Epoch 14/30	1	- 14s 720ms/s+	ep - loss.	0.2143 - accuracy	: 0.9200 - val 1	oss: 0.2682 -
_accuracy: 0.8922	val_accuracy: 0.8987 Epoch 15/30				-	_	
Epoch 16/30		======]	- 19s 1s/step	- loss: 0.2	2107 - accuracy: 0	.9233 - val_loss	: 0.2756 - val
	<del>-</del> ,						

Param #

```
19/19 [============== ] - 14s 738ms/step - loss: 0.2120 - accuracy: 0.9333 - val loss: 0.2495 -
     val_accuracy: 0.8954
     Epoch 17/30
     19/19 [=======
                  ==========] - 14s 762ms/step - loss: 0.2124 - accuracy: 0.9250 - val loss: 0.2390 -
     val accuracy: 0.9085
     Epoch 18/30
                19/19 [=====
     val_accuracy: 0.9118
     Epoch 19/30
     19/19 [==
                        ========] - 14s 772ms/step - loss: 0.1940 - accuracy: 0.9317 - val loss: 0.2446 -
     val accuracy: 0.9020
     Epoch 20/30
     19/19 [=====
                     =========] - 14s 730ms/step - loss: 0.1896 - accuracy: 0.9267 - val_loss: 0.2154 -
     val_accuracy: 0.9150
     Epoch 21/30
                       :========] - 14s 738ms/step - loss: 0.1921 - accuracy: 0.9317 - val_loss: 0.2840 -
     19/19 [====
     val_accuracy: 0.8922
     Epoch 22/30
     val_accuracy: 0.8987
     Epoch 23/30
     19/19 [=========] - 13s 720ms/step - loss: 0.1678 - accuracy: 0.9333 - val loss: 0.2236 -
     val_accuracy: 0.9150
     Epoch 24/30
     val_accuracy: 0.9118
     Epoch 25/30
     19/19 [=====
                 val_accuracy: 0.9052
     Epoch 26/30
     19/19 [=========] - 14s 758ms/step - loss: 0.1939 - accuracy: 0.9317 - val loss: 0.2632 -
     val accuracy: 0.8987
     Epoch 27/30
     val_accuracy: 0.9248
     Epoch 28/30
               19/19 [=====
     val accuracy: 0.9052
     Epoch 29/30
     19/19 [===
                        ========] - 14s 745ms/step - loss: 0.1564 - accuracy: 0.9450 - val_loss: 0.2461 -
     val accuracy: 0.9052
     Epoch 30/30
     19/19 [==
                          ======] - 14s 725ms/step - loss: 0.1853 - accuracy: 0.9200 - val loss: 0.2513 -
     val_accuracy: 0.8954
In [ ]: history.history.keys()
     dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])
     plt.plot(history.history['loss'])
In [ ]:
      plt.plot(history.history['val_loss'])
      plt.legend(['trainig','validation'])
     plt.title('Training and validation loss')
     plt.xlabel('epoch')
     Text(0.5, 0, 'epoch')
```

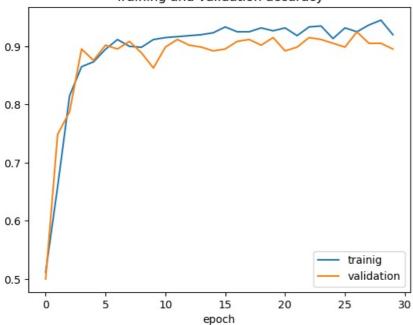
## out[]:



In [ ]: plt.plot(history.history['accuracy'])
 plt.plot(history.history['val\_accuracy'])

```
plt.legend(['trainig','validation'])
plt.title('Training and validation accuracy')
plt.xlabel('epoch')
Text(0.5, 0, 'epoch')
```

## Training and validation accuracy



```
In [ ]: test_loss , test_acc = model.evaluate(test_generator)
        print('test acc :{} test loss:{}'.format(test_acc,test_loss))
        4/4 [===========] - 2s 548ms/step - loss: 0.2142 - accuracy: 0.9000
        test acc :0.8999999761581421 test loss:0.2141808718442917
In [ ]: import numpy as np
        from google.colab import files
        from keras.preprocessing import image
        upload_file = files.upload()
        for i in upload_file.keys():
         image_path='/content/'+i
          img = image.load_img(image_path , target_size=(150,150))
          images = image.img_to_array(img)
          images=np.expand_dims(images,axis=0)
          prediction = model.predict(images)
          print(i)
          if prediction==0:
            print('mask')
          else:
            print('nomask')
```

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