

tablenet final

April 20, 2021

1 TableNet: Deep Learning model for end-to-end Table detection and Tabular data extraction from Scanned Document Images

With the widespread use of mobile phones and scanners to photograph and upload documents, the need for extracting the information trapped in unstructured document images such as retail receipts, insurance claim forms and financial invoices is becoming more acute.

A major hurdle to this objective is that these images often contain information in the form of tables and extracting data from tabular sub-images presents a unique set of challenges. This includes accurate detection of the tabular region within an image, and subsequently detecting and extracting information from the rows and columns of the detected table.

While some progress has been made in table detection, extracting the table contents is still a challenge since this involves more fine grained table structure(rows & columns) recognition. Prior approaches have attempted to solve the table detection and structure recognition problems independently using two separate models.

TableNet: a novel end-to- end deep learning model for both table detection and structure recognition. The model exploits the interdependence between the twin tasks of table detection and table structure recognition to segment out the table and column regions.

2 problem statement

We need to detect tabular structure from a document image, then extract tabular information (row-col format).

3 data source

<https://drive.google.com/drive/folders/1410iMmQCXbA9GJP5CqLEMfjjv-hOWlac?usp=sharing>

4 ML/DL problem type

classification - we need to predict each pixel either white(background) or black(tabular region).

5 Performance Metric

1. precision
2. recall

3. f1 score

6 google drive mount

```
[1]: from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

7 dependencies

```
[2]: !sudo apt install tesseract-ocr
!pip install pytesseract

import warnings
warnings.filterwarnings('ignore')

import tensorflow as tf
import os
import matplotlib.pyplot as plt
import numpy as np
import cv2
import xml.etree.ElementTree as ET
from PIL import Image
import pandas as pd
import pytesseract
from sklearn.model_selection import train_test_split
import tensorflow as tf
from google.colab.patches import cv2_imshow
```

Reading package lists... Done

Building dependency tree

Reading state information... Done

The following additional packages will be installed:

tesseract-ocr-eng tesseract-ocr-osd

The following NEW packages will be installed:

tesseract-ocr tesseract-ocr-eng tesseract-ocr-osd

0 upgraded, 3 newly installed, 0 to remove and 31 not upgraded.

Need to get 4,795 kB of archives.

After this operation, 15.8 MB of additional disk space will be used.

Get:1 <http://archive.ubuntu.com/ubuntu> bionic/universe amd64 tesseract-ocr-eng
all 4.00~git24-0e00fe6-1.2 [1,588 kB]

Get:2 <http://archive.ubuntu.com/ubuntu> bionic/universe amd64 tesseract-ocr-osd
all 4.00~git24-0e00fe6-1.2 [2,989 kB]

Get:3 <http://archive.ubuntu.com/ubuntu> bionic/universe amd64 tesseract-ocr amd64
4.00~git2288-10f4998a-2 [218 kB]

```

Fetched 4,795 kB in 2s (2,860 kB/s)
debconf: unable to initialize frontend: Dialog
debconf: (No usable dialog-like program is installed, so the dialog based
frontend cannot be used. at /usr/share/perl5/Debconf/FrontEnd/Dialog.pm line 76,
<> line 3.)
debconf: falling back to frontend: Readline
debconf: unable to initialize frontend: Readline
debconf: (This frontend requires a controlling tty.)
debconf: falling back to frontend: Teletype
dpkg-preconfigure: unable to re-open stdin:
Selecting previously unselected package tesseract-ocr-eng.
(Reading database ... 160983 files and directories currently installed.)
Preparing to unpack .../tesseract-ocr-eng_4.00~git24-0e00fe6-1.2_all.deb ...
Unpacking tesseract-ocr-eng (4.00~git24-0e00fe6-1.2) ...
Selecting previously unselected package tesseract-ocr-osd.
Preparing to unpack .../tesseract-ocr-osd_4.00~git24-0e00fe6-1.2_all.deb ...
Unpacking tesseract-ocr-osd (4.00~git24-0e00fe6-1.2) ...
Selecting previously unselected package tesseract-ocr.
Preparing to unpack .../tesseract-ocr_4.00~git2288-10f4998a-2_amd64.deb ...
Unpacking tesseract-ocr (4.00~git2288-10f4998a-2) ...
Setting up tesseract-ocr-osd (4.00~git24-0e00fe6-1.2) ...
Setting up tesseract-ocr-eng (4.00~git24-0e00fe6-1.2) ...
Setting up tesseract-ocr (4.00~git2288-10f4998a-2) ...
Processing triggers for man-db (2.8.3-2ubuntu0.1) ...
Collecting pytesseract
  Downloading https://files.pythonhosted.org/packages/a0/e6/a4e9fc8a93c1318540e8
de6d8d4beb5749b7960388a7c7f27799fc2dd016/pytesseract-0.3.7.tar.gz
Requirement already satisfied: Pillow in /usr/local/lib/python3.7/dist-packages
(from pytesseract) (7.1.2)
Building wheels for collected packages: pytesseract
  Building wheel for pytesseract (setup.py) ... done
  Created wheel for pytesseract: filename=pytesseract-0.3.7-py2.py3-none-any.whl
size=13945
sha256=48bbfd21121b97c690b2ca92b9558dbc4ca5b70ac92e0092152525045099beca
  Stored in directory: /root/.cache/pip/wheels/81/20/7e/1dd0daad1575d5260916bb1e
9781246430647adaef4b3ca3b3
Successfully built pytesseract
Installing collected packages: pytesseract
Successfully installed pytesseract-0.3.7

```

8 paths

```

[3]: originalImage = "/content/drive/MyDrive/case study - II/tablenet/data/
    ↳Marmot_data/10.1.1.1.2006_3.bmp"
imageMask = "/content/drive/MyDrive/case study - II/tablenet/data/Marmot_data/
    ↳10.1.1.1.2006_3.xml"

```

```

fileSavepath = "/content/drive/MyDrive/case study - II/tablenet/data/final data/
↳"
table_mask_path = "/content/drive/MyDrive/case study - II/tablenet/data/final_
↳data/tablemask/"
col_mask_path = "/content/drive/MyDrive/case study - II/tablenet/data/final_
↳data/colmask/"
org_image_path = "/content/drive/MyDrive/case study - II/tablenet/data/final_
↳data/orgimage/"
dataPath = "/content/drive/MyDrive/case study - II/tablenet/data/Marmot_data/"

```

9 creating dataframe of image and xml files

```

[ ]: """
CREATE DATAFRAME OF PATHS.
dataframe
-----
image_path, xml_path

* go through every file in mammoth folder (dataPath).
* check a .bmp file, extract name, check if .xml file is present or not -->_
↳store in row
"""

image_xml_dict = {"image_path": [], "xml_path": []}

for file in os.listdir(dataPath):
    if ".bmp" in file:
        name = file.split(".bmp")[0]
        if os.path.exists(dataPath+name+".xml"):
            image_xml_dict['image_path'].append(name+".bmp")
            image_xml_dict['xml_path'].append(name+".xml")

image_xml_df = pd.DataFrame(image_xml_dict)

image_xml_df.head(2)

```

```

[ ]:
      image_path      xml_path
0  10.1.1.1.2044_7.bmp  10.1.1.1.2044_7.xml
1  10.1.1.1.34.330_9.bmp  10.1.1.1.34.330_9.xml

```

10 creating mask from .xml files

10.1 xml file for every image defining regions of every table in an image.

1. 'size' element denotes the height, width and depth of image.

2. 'object' element denotes every single col region in a image.

```
[ ]: """
<size>
    <width>793</width>
    <height>1123</height>
    <depth>3</depth>
</size>

<object>
    <name>column</name>
    <pose>Unspecified</pose>
    <truncated>0</truncated>
    <difficult>0</difficult>
    <bndbox>
        <xmin>458</xmin>
        <ymin>710</ymin>
        <xmax>517</xmax>
        <ymax>785</ymax>
    </bndbox>
</object>

"""

# /content/drive/MyDrive/case study - II/tablenet/data/final data/

def euc_dist(point1, point2):
    dist = np.linalg.norm(point1 - point2)
    return dist

def show_image_plt(image_arr):
    plt.figure(figsize=(5,5))
    plt.imshow(image_arr)
    plt.show()

def save_image(name, image_arr):
    im = Image.fromarray(image_arr)
    im.save(name)

final_dataframe_dict = {"image": [], "table_mask": [], "col_mask": []}

for index, row in image_xml_df.iterrows():
```

```

# per row --> xml_path
org_img_mask_xml = row['xml_path'] # .xml path
image = dataPath + row['image_path'] # image .bmp path

# file name
name = org_img_mask_xml.split(".xml")[0]

# reading xml file
tree = ET.parse(dataPath + org_img_mask_xml)
root = tree.getroot()

size = root.find('size')
width = int(size.find('width').text)
height = int(size.find('height').text)
depth = int(size.find('depth').text)

# creating empty mask image
col_mask_empty = np.zeros(shape=(height, width), dtype=np.uint8)
table_mask_empty = np.zeros(shape=(height, width), dtype=np.uint8)

# finding objects
objects = tree.findall('object')
table_xmin = 0
table_ymin = 0
table_xmax = 0
table_ymax = 0
prev_dist = 0
dist = 0
forward_flag = False
backward_flag = False
newtable_flag = True

# creating empty mask image
col_mask_empty = np.zeros(shape=(height, width), dtype=np.uint8)
table_mask_empty = np.zeros(shape=(height, width), dtype=np.uint8)

plt.figure(figsize=(5, 5))

objects = tree.findall('object')

for index, object in enumerate(objects):

    bndbox = object.find('bndbox')

```

```

xmin = int(bndbox.find('xmin').text)
xmax = int(bndbox.find('xmax').text)
ymin = int(bndbox.find('ymin').text)
ymax = int(bndbox.find('ymax').text)

col_mask_empty[ymin:ymax, xmin:xmax] = 255

if index == 0:

    prev_xmin = int(bndbox.find('xmin').text)
    prev_ymin = int(bndbox.find('ymin').text)
    prev_xmax = int(bndbox.find('xmax').text)
    prev_ymax = int(bndbox.find('ymax').text)

else:

    if xmin > prev_xmin and newtable_flag:

        table_xmin = prev_xmin
        table_ymin = prev_ymin
        newtable_flag = False
        forward_flag = True
        backward_flag = False

    if xmin < prev_xmin and newtable_flag:

        table_xmax = prev_xmax
        table_ymax = prev_ymax

        newtable_flag = False
        backward_flag = True
        forward_flag = False

    if forward_flag:
        dist = euc_dist(np.array([xmin, ymin]), np.array([prev_xmax,
→prev_ymin]))

        if prev_dist == 0:

```

```

        prev_dist = dist
    else:

        if int(np.divide(dist, prev_dist)) > 5:
            newtable_flag = True
            table_mask_empty[table_ymin:prev_ymax, table_xmin:
↪prev_xmax] = 255

            prev_dist = 0

        if index==len(objects)-1:
            newtable_flag = True
            table_mask_empty[table_ymin:ymax, table_xmin:xmax] = 255

            prev_dist = 0

    if backward_flag:
        dist = euc_dist(np.array([xmax, ymin]), np.array([prev_xmin,
↪prev_ymin]))

        if prev_dist == 0:
            prev_dist = dist
        else:
            if int(np.divide(dist, prev_dist)) > 5 or
↪index==len(objects)-1:
                newtable_flag = True
                table_mask_empty[ymin:table_ymax, xmin:table_xmax] = 255
                prev_dist = 0

    prev_xmin = int(bndbox.find('xmin').text)
    prev_ymin = int(bndbox.find('ymin').text)
    prev_xmax = int(bndbox.find('xmax').text)
    prev_ymax = int(bndbox.find('ymax').text)
    prev_dist = dist

    save_image(table_mask_path+ name+".jpeg", table_mask_empty)
    save_image(col_mask_path + name+".jpeg", col_mask_empty)

    final_dataframe_dict['table_mask'].append(table_mask_path+ name+".jpeg")
    final_dataframe_dict['col_mask'].append(col_mask_path + name+".jpeg")
    final_dataframe_dict['image'].append(image)

# creating dataframe --> (original_image, table_mask, col_mask)
final_dataframe = pd.DataFrame(final_dataframe_dict)

```



```
final_dataframe.head(2)
final_dataframe.to_csv("/content/drive/MyDrive/case study - II/tablenet/data/
↳final_dataframe.csv", index=False)
```

<Figure size 360x360 with 0 Axes>

<Figure size 360x360 with 0 Axes>

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[illegible]

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```

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<Figure size 360x360 with 0 Axes>

```

11 reading dataframe

```

[ ]: final_dataframe = pd.read_csv("/content/drive/MyDrive/case study - II/tablenet/
    ↳data/final_dataframe.csv")
    final_dataframe.head(2)

```

```

[ ]:                                     image ...
    col_mask
0  /content/drive/MyDrive/case study - II/tablene... ...
   /content/drive/MyDrive/case study - II/tablene...
1  /content/drive/MyDrive/case study - II/tablene... ...
   /content/drive/MyDrive/case study - II/tablene...

[2 rows x 3 columns]

```

12 data generator

```

[ ]: X_train, X_test = train_test_split(final_dataframe, test_size=0.2)

```



```
[ ]: training_dataset = (
    tf.data.Dataset.from_tensor_slices(
        (
            tf.cast(X_train['image'].values, tf.string),
            tf.cast(X_train['table_mask'].values, tf.string),
            tf.cast(X_train['col_mask'].values, tf.string),
        )
    )
)

testing_dataset = (
    tf.data.Dataset.from_tensor_slices(
        (
            tf.cast(X_test['image'].values, tf.string),
            tf.cast(X_test['table_mask'].values, tf.string),
            tf.cast(X_test['col_mask'].values, tf.string),
        )
    )
)
```

```
[ ]: # https://www.tensorflow.org/tutorials/load\_data/images

@tf.function
def load_image(image, table_mask, col_mask):

    image = tf.io.read_file(image)
    table_mask=tf.io.read_file(table_mask)
    col_mask=tf.io.read_file(col_mask)

    image=tf.io.decode_bmp(image, channels=3)
    image=tf.image.resize(image, [1024, 1024])
    image = tf.cast(image, tf.float32) / 255.0

    table_mask=tf.io.decode_jpeg(table_mask, channels=1)
    table_mask=tf.image.resize(table_mask, [1024, 1024])
    table_mask = table_mask / 255.0

    col_mask=tf.io.decode_jpeg(col_mask, channels=1)
    col_mask=tf.image.resize(col_mask, [1024, 1024])
    col_mask = col_mask / 255.0

    return image, {"table_mask":table_mask, "col_mask":col_mask}
```

```
# creating dataset object
train = training_dataset.map(load_image, num_parallel_calls=tf.data.AUTOTUNE)
test = testing_dataset.map(load_image)
```

```
[ ]: BATCH_SIZE = 2
      BUFFER_SIZE = 10
      train_steps = len(X_train) // BATCH_SIZE

      # for feeding to training
      train_dataset = train.cache().shuffle(BUFFER_SIZE).batch(BATCH_SIZE).repeat()
      train_dataset = train_dataset.prefetch(buffer_size=tf.data.experimental.
      ↪AUTOTUNE)
      test_dataset = test.batch(BATCH_SIZE)
```

```
[ ]: def display(display_list):
      plt.figure(figsize=(15, 15))
      title = ['Input Image', 'Table Mask', 'Column Mask', 'Masked image']
      for i in range(len(display_list)):
          plt.subplot(1, len(display_list), i+1)
          plt.title(title[i])

          image = display_list[i]

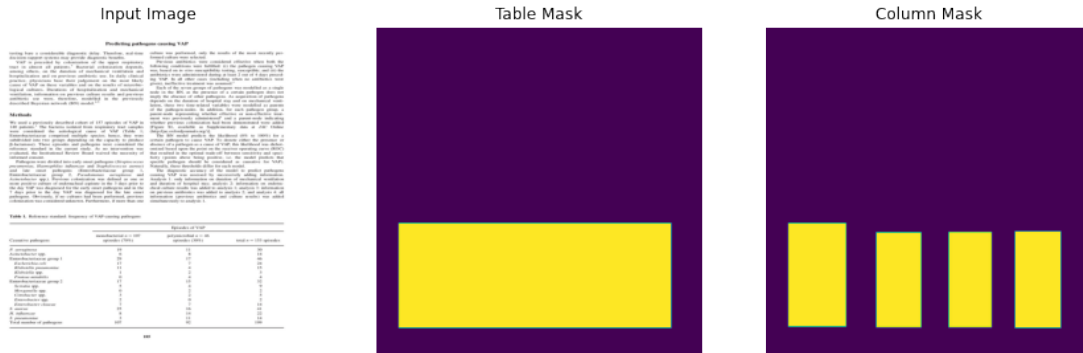
          plt.imshow(tf.keras.preprocessing.image.array_to_img(image))
          plt.axis('off')
      plt.show()

      for image, mask in train.take(1):

          sample_image = image
          sample_table_mask = mask['table_mask']
          sample_col_mask = mask['col_mask']

          print(image.shape)
          print(mask['table_mask'].shape)
          print(mask['col_mask'].shape)
          display([image, mask['table_mask'], mask['col_mask']])
```

```
(1024, 1024, 3)
(1024, 1024, 1)
(1024, 1024, 1)
```



13 model

```
[4]: import tensorflow as tf
from tensorflow.keras.applications import VGG19
from tensorflow.keras.layers import Input
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Conv2D
from tensorflow.keras.layers import Dropout
from tensorflow.keras.layers import UpSampling2D
from tensorflow.keras.layers import Concatenate
from tensorflow.keras.layers import Layer
from tensorflow.keras.layers import Activation
from tensorflow.keras.layers import Conv2DTranspose
from tensorflow.keras.utils import plot_model
from tensorflow.keras import backend as K
```

```
[5]: """
Table decoder
-----

x = conv7(1x1)
x = x(upscaled) + vgg19(pool4)
x = x(upscaled) + vgg19(pool3)
x = upscaled to match input dimention (1024)

column decoder
-----

x = conv7(1x1, relu)
x = dropout(0.8)
x = conv8(1x1)
x = upscaled() + vgg19(pool4)
x = x(upscaled) + vgg19(pool3)
x = upscaled (1024)
```

```
"""
```

```
tf.keras.backend.clear_session()
```

```
class table_mask(Layer):
```

```
    def __init__(self):
        super().__init__()
        self.conv_7 = Conv2D(kernel_size=(1,1), filters=128,
↪kernel_regularizer=tf.keras.regularizers.l2(0.002))
        self.upsample_pool4 = UpSampling2D(size=(2, 2),
↪interpolation='bilinear')
        self.upsample_pool3 = UpSampling2D(size=(2, 2),
↪interpolation='bilinear')
        self.upsample_final = Conv2DTranspose(filters=2, kernel_size=3,
↪strides=2, padding='same', activation='softmax')

    def call(self, input, pool3, pool4):

        x = self.conv_7(input)
        x = self.upsample_pool4(x)
        x = Concatenate()([x, pool4])

        x = self.upsample_pool3(x)
        x = Concatenate()([x, pool3])

        x = UpSampling2D((2,2))(x)
        x = UpSampling2D((2,2))(x)

        x = self.upsample_final(x)

    return x
```

```
class col_mask(Layer):
```

```
    def __init__(self):
        super().__init__()
        self.conv_7 = Conv2D(kernel_size=(1,1), filters=128,
↪kernel_regularizer=tf.keras.regularizers.l2(0.004),
↪kernel_initializer='he_normal',)
        self.drop = Dropout(0.8)
```

```

        self.conv_8 = Conv2D(kernel_size=(1,1), filters=128,
↪kernel_regularizer=tf.keras.regularizers.l2(0.004),
↪kernel_initializer='he_normal',)
        self.upsample_pool4 = UpSampling2D(size=(2, 2),
↪interpolation='bilinear')
        self.upsample_pool3 = UpSampling2D(size=(2, 2),
↪interpolation='bilinear')
        self.upsample_final = Conv2DTranspose(filters=2, kernel_size=3,
↪strides=2, padding='same', activation='softmax')

    def call(self, input, pool3, pool4):

        x = self.conv_7(input)
        x = self.drop(x)
        x = self.conv_8(x)

        x = self.upsample_pool4(x)
        x = Concatenate()([x, pool4])

        x = self.upsample_pool3(x)
        x = Concatenate()([x, pool3])

        x = UpSampling2D((2,2))(x)
        x = UpSampling2D((2,2))(x)

        x = self.upsample_final(x)

    return x

input_shape = (1024, 1024, 3)
input_ = Input(shape=input_shape)

vgg19_ = VGG19(
    include_top=False,
    weights="imagenet",
    input_tensor=input_,
    input_shape=None,
    pooling=None,
    classes=1000,
    classifier_activation="softmax",
)

for layer in vgg19_.layers:

```

```

layer.trainable = False

pool3 = vgg19_.get_layer('block3_pool').output
pool4 = vgg19_.get_layer('block4_pool').output

conv_1_1_1 = Conv2D(filters=128, kernel_size=(1, 1), activation='relu',
    ↳name="block6_conv1", kernel_regularizer=tf.keras.regularizers.l2(0.
    ↳004))(vgg19_.output)
conv_1_1_1_drop = Dropout(0.8)(conv_1_1_1)

conv_1_1_2 = Conv2D(filters=128, kernel_size=(1, 1), activation='relu',
    ↳name="block6_conv2", kernel_regularizer=tf.keras.regularizers.l2(0.
    ↳004))(conv_1_1_1_drop)
conv_1_1_2_drop = Dropout(0.8)(conv_1_1_2)

table_mask = table_mask()(conv_1_1_2_drop, pool3, pool4)
col_mask = col_mask()(conv_1_1_2_drop, pool3, pool4)

model = Model(input_, [table_mask, col_mask])

model.summary()

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels_notop.h5
 80142336/80134624 [=====] - 0s 0us/step
 Model: "model"

```

-----
Layer (type)                Output Shape              Param #   Connected to
-----
input_1 (InputLayer)        [(None, 1024, 1024, 0)   0         input_1[0][0]
-----
block1_conv1 (Conv2D)        (None, 1024, 1024, 6)    1792      input_1[0][0]
-----
block1_conv2 (Conv2D)        (None, 1024, 1024, 6)    36928     block1_conv1[0][0]
-----
block1_pool (MaxPooling2D)   (None, 512, 512, 64)    0         block1_conv2[0][0]
-----
block2_conv1 (Conv2D)        (None, 512, 512, 128)    73856     block1_pool[0][0]

```

```

-----
-----
block2_conv2 (Conv2D)          (None, 512, 512, 128 147584
block2_conv1[0][0]

-----
-----
block2_pool (MaxPooling2D)     (None, 256, 256, 128 0
block2_conv2[0][0]

-----
-----
block3_conv1 (Conv2D)          (None, 256, 256, 256 295168
block2_pool[0][0]

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-----
block3_conv2 (Conv2D)          (None, 256, 256, 256 590080
block3_conv1[0][0]

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-----
block3_conv3 (Conv2D)          (None, 256, 256, 256 590080
block3_conv2[0][0]

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-----
block3_conv4 (Conv2D)          (None, 256, 256, 256 590080
block3_conv3[0][0]

-----
-----
block3_pool (MaxPooling2D)     (None, 128, 128, 256 0
block3_conv4[0][0]

-----
-----
block4_conv1 (Conv2D)          (None, 128, 128, 512 1180160
block3_pool[0][0]

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-----
block4_conv2 (Conv2D)          (None, 128, 128, 512 2359808
block4_conv1[0][0]

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block4_conv3 (Conv2D)          (None, 128, 128, 512 2359808
block4_conv2[0][0]

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block4_conv4 (Conv2D)          (None, 128, 128, 512 2359808
block4_conv3[0][0]

-----
-----
block4_pool (MaxPooling2D)     (None, 64, 64, 512) 0
block4_conv4[0][0]

```

```

-----
block5_conv1 (Conv2D)          (None, 64, 64, 512) 2359808
block4_pool[0][0]

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block5_conv2 (Conv2D)          (None, 64, 64, 512) 2359808
block5_conv1[0][0]

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block5_conv3 (Conv2D)          (None, 64, 64, 512) 2359808
block5_conv2[0][0]

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block5_conv4 (Conv2D)          (None, 64, 64, 512) 2359808
block5_conv3[0][0]

-----
block5_pool (MaxPooling2D)      (None, 32, 32, 512) 0
block5_conv4[0][0]

-----
block6_conv1 (Conv2D)          (None, 32, 32, 128) 65664
block5_pool[0][0]

-----
dropout (Dropout)              (None, 32, 32, 128) 0
block6_conv1[0][0]

-----
block6_conv2 (Conv2D)          (None, 32, 32, 128) 16512      dropout[0][0]

-----
dropout_1 (Dropout)            (None, 32, 32, 128) 0
block6_conv2[0][0]

-----
table_mask (table_mask)        (None, 1024, 1024, 2 32642      dropout_1[0][0]
block3_pool[0][0]
block4_pool[0][0]

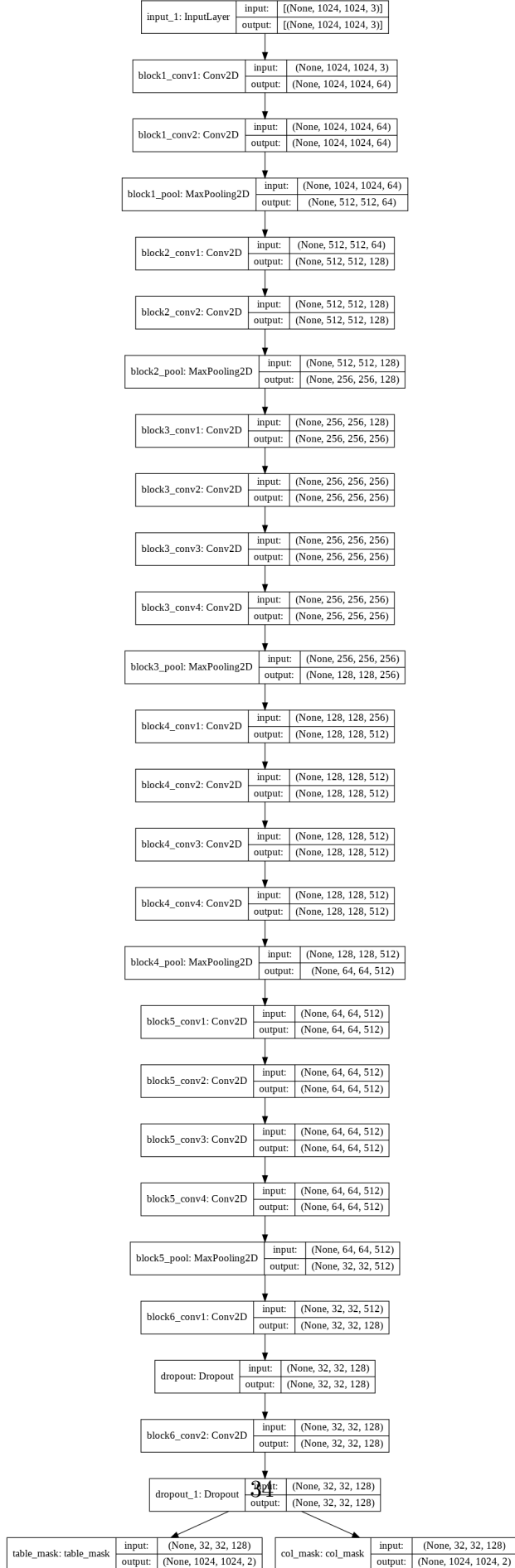
-----
col_mask (col_mask)            (None, 1024, 1024, 2 49154      dropout_1[0][0]
block3_pool[0][0]
block4_pool[0][0]
=====
Total params: 20,188,356

```


Trainable params: 163,972
Non-trainable params: 20,024,384


```
[6]: plot_model(model, show_shapes=True, show_layer_names=True)
```

[6]:



14 model training

[]:

```
[ ]: losses = {
    "table_mask": 'sparse_categorical_crossentropy',
    "col_mask": 'sparse_categorical_crossentropy',
}

# tf.keras.losses.SparseCategoricalCrossentropy(from_logits=False)

filepath = "/content/drive/MyDrive/case study - II/tablenet/model checkpoint/
↳table_net.h5"
model_checkpoint = tf.keras.callbacks.ModelCheckpoint(filepath, monitor =
↳"val_table_mask_loss", save_best_only=True, verbose = 0, mode="min")

es = tf.keras.callbacks.EarlyStopping(monitor='val_loss', mode='min',
↳patience=5,)

class F1_Score(tf.keras.metrics.Metric):

    def __init__(self, name='f1_score', **kwargs):
        super().__init__(name=name, **kwargs)
        self.f1 = self.add_weight(name='f1', initializer='zeros')
        self.precision_fn = tf.keras.metrics.Precision(thresholds=0.5)
        self.recall_fn = tf.keras.metrics.Recall(thresholds=0.5)

    def update_state(self, y_true, y_pred, sample_weight=None):
        p = self.precision_fn(y_true, tf.argmax(y_pred, axis=-1))
        r = self.recall_fn(y_true, tf.argmax(y_pred, axis=-1))
        # since f1 is a variable, we use assign
        self.f1.assign(2 * ((p * r) / (p + r + 1e-6)))

    def result(self):
        return self.f1

    def reset_states(self):
        # we also need to reset the state of the precision and recall objects
        self.precision_fn.reset_states()
        self.recall_fn.reset_states()
```

```

        self.f1.assign(0)

metrics = [F1_Score()]

global init_lr
init_lr = 0.0001

model.compile(optimizer=tf.keras.optimizers.Adam(learning_rate=init_lr,
↪epsilon=1e-8,),
              loss=losses,
              metrics=metrics, )

```

```

[ ]: def show_predictions(dataset=None, num=1):

    if dataset:
        for image, mask in dataset.take(1):
            table_mask_pred, col_mask_pred = model.predict(image)

            table_mask_pred = tf.argmax(table_mask_pred, axis=-1)
            table_mask_pred = table_mask_pred[..., tf.newaxis][0]

            col_mask_pred = tf.argmax(col_mask_pred, axis=-1)
            col_mask_pred = col_mask_pred[..., tf.newaxis][0]

            im=tf.keras.preprocessing.image.array_to_img(image[0])
            im.save('image.png')

            im=tf.keras.preprocessing.image.array_to_img(table_mask_pred)
            im.save('table_mask_pred.png')

            im=tf.keras.preprocessing.image.array_to_img(col_mask_pred)
            im.save('col_mask_pred.png')

            img_org = Image.open('./image.png')
            table_mask = Image.open('./table_mask_pred.png')
            col_mask = Image.open('./col_mask_pred.png')

            # convert images
            img_mask = table_mask.convert('L')
            # img_mask = col_mask.convert('L')

            # grayscale
            # add alpha channel

```

```

img_org.putalpha(img_mask)

# save as png which keeps alpha channel
img_org.save('output.png')

display([image[0], table_mask_pred, col_mask_pred, img_org])

pytesseract.pytesseract.tesseract_cmd = r'/usr/bin/tesseract'
text = pytesseract.image_to_string(Image.open('./output.png'),
↳ lang='eng' ) # config='--psm 11'
print(text)

class DisplayCallback(tf.keras.callbacks.Callback):

    def __init__(self):
        self.history = {'val_table_mask_loss': []}
        self.init_lr = init_lr

    def on_epoch_end(self, epoch, logs=None):
        if epoch % 1 == 0:
            show_predictions(test_dataset, 1)

            self.history['val_table_mask_loss'].append(logs.
↳ get('val_table_mask_loss'))
            if epoch > 2:
                cur_loss = self.history['val_table_mask_loss'][epoch]
                prev_loss = self.history['val_table_mask_loss'][epoch-1]

                if cur_loss > prev_loss:
                    self.init_lr = self.init_lr * 0.93
                    K.set_value(self.model.optimizer.learning_rate, self.
↳ init_lr)

```

```

[ ]: EPOCHS = 50
VAL_SUBSPLITS = 30
VALIDATION_STEPS = len(X_test)//BATCH_SIZE//VAL_SUBSPLITS

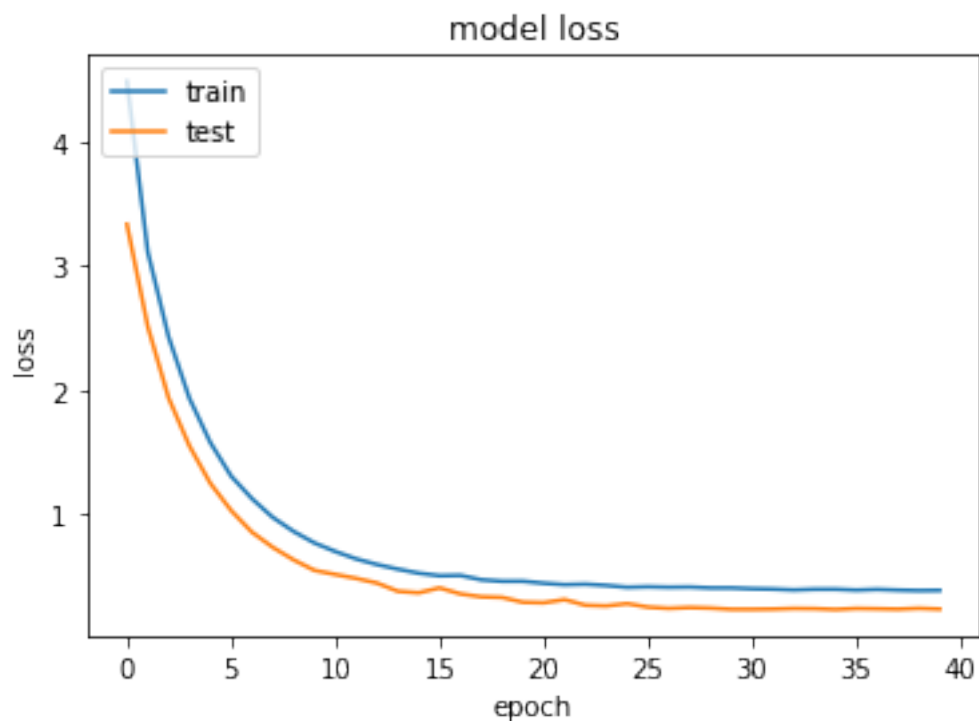
history = model.fit(train_dataset,
                    epochs=EPOCHS,
                    steps_per_epoch=train_steps,
                    validation_data=test_dataset,
                    validation_steps=VALIDATION_STEPS,

```

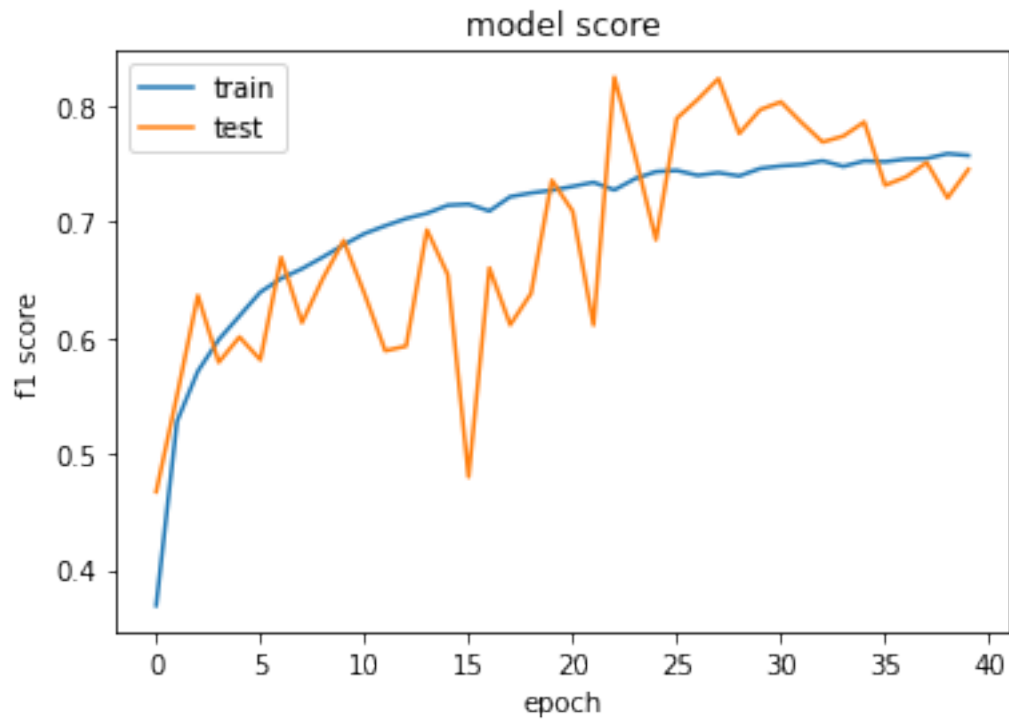
```
callbacks=[model_checkpoint, es, ↳DisplayCallback()])
```

Output hidden; open in <https://colab.research.google.com> to view.

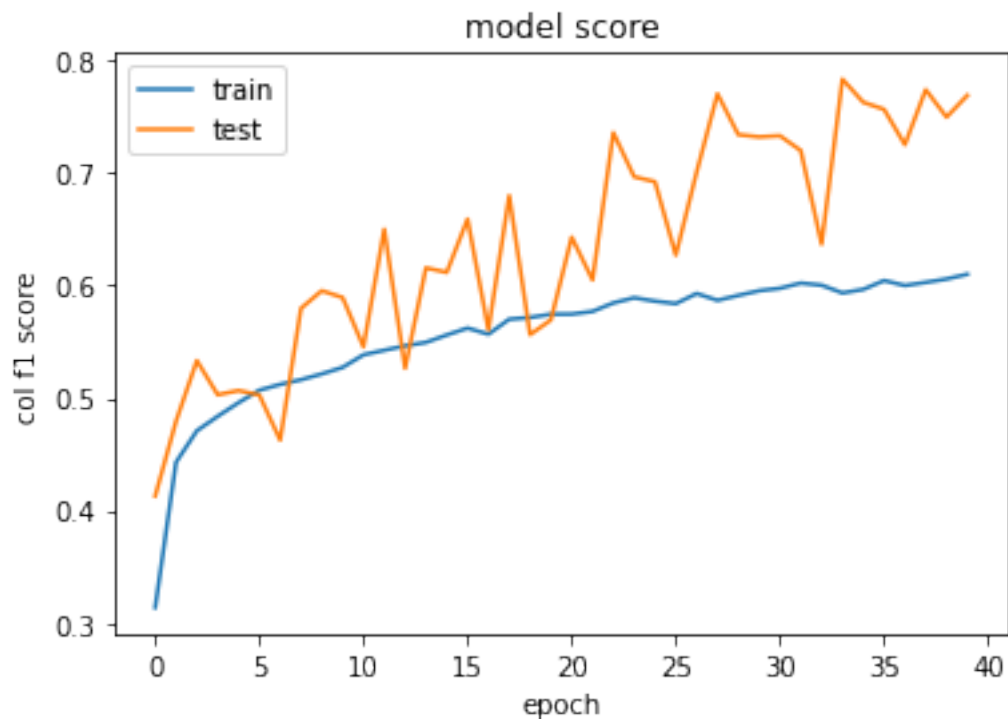
```
[ ]: plt.plot(history.history['loss'])  
plt.plot(history.history['val_loss'])  
plt.title('model loss')  
plt.ylabel('loss')  
plt.xlabel('epoch')  
plt.legend(['train', 'test'], loc='upper left')  
plt.show()
```



```
[ ]: plt.plot(history.history['table_mask_f1_score'])  
plt.plot(history.history['val_table_mask_f1_score'])  
plt.title('model score')  
plt.ylabel('table f1 score')  
plt.xlabel('epoch')  
plt.legend(['train', 'test'], loc='upper left')  
plt.show()
```



```
[ ]: plt.plot(history.history['col_mask_f1_score'])
plt.plot(history.history['val_col_mask_f1_score'])
plt.title('model score')
plt.ylabel('col f1 score')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
```

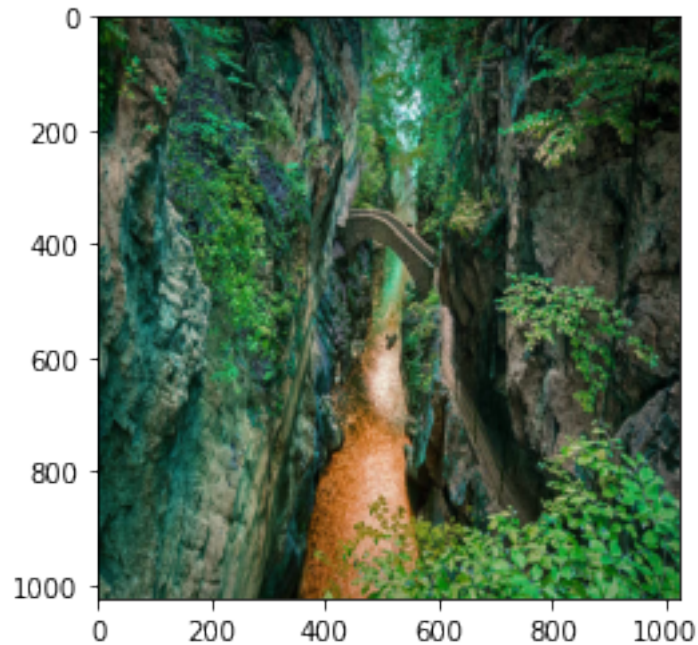


15 prediction

```
[7]: model_path = "/content/drive/MyDrive/case study - II/tablenet/model checkpoint/
      ↳table_net.h5"
      model.load_weights(model_path)
```

```
[16]: image = cv2.imread('/content/9cdb1c60321e4636bc37363cdccc0e38.jpg', cv2.
      ↳IMREAD_UNCHANGED)
      image = cv2.resize(image,(1024, 1024), cv2.INTER_NEAREST)
      plt.imshow(image)
      image = np.expand_dims(image, axis=0)
      print(image.shape)
```

```
(1, 1024, 1024, 3)
```

```
[17]: table_mask_pred, col_mask_pred = model.predict(image)

table_mask_pred = tf.argmax(table_mask_pred, axis=-1)
table_mask_pred = table_mask_pred[..., tf.newaxis][0]

col_mask_pred = tf.argmax(col_mask_pred, axis=-1)
col_mask_pred = col_mask_pred[..., tf.newaxis][0]

im=tf.keras.preprocessing.image.array_to_img(image[0])
im.save('image.png')

im=tf.keras.preprocessing.image.array_to_img(table_mask_pred)
im.save('table_mask_pred.png')

im=tf.keras.preprocessing.image.array_to_img(col_mask_pred)
im.save('col_mask_pred.png')
```

```
[ ]: count = 0

for image, mask in test_dataset.take(10):

    print(image.shape)

    table_mask_pred, col_mask_pred = model.predict(image)
```

```

table_mask_pred = tf.argmax(table_mask_pred, axis=-1)
table_mask_pred = table_mask_pred[..., tf.newaxis][0]

col_mask_pred = tf.argmax(col_mask_pred, axis=-1)
col_mask_pred = col_mask_pred[..., tf.newaxis][0]

im=tf.keras.preprocessing.image.array_to_img(image[0])
im.save('image.png')

im=tf.keras.preprocessing.image.array_to_img(table_mask_pred)
im.save('table_mask_pred.png')

im=tf.keras.preprocessing.image.array_to_img(col_mask_pred)
im.save('col_mask_pred.png')

img_org = Image.open('./image.png')
table_mask = Image.open('./table_mask_pred.png')
col_mask = Image.open('./col_mask_pred.png')

# convert images
img_mask = table_mask.convert('L')
# img_mask = col_mask.convert('L')

# grayscale
# add alpha channel
img_org.putalpha(img_mask)

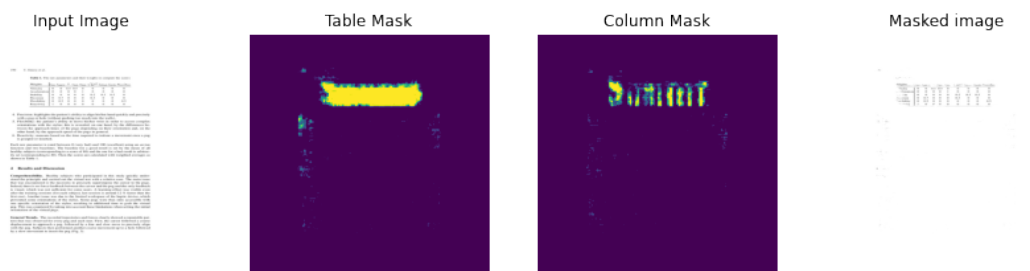
# save as png which keeps alpha channel
img_org.save('output.png')

display([image[0], table_mask_pred, col_mask_pred, img_org])

pytesseract.pytesseract.tesseract_cmd = r'/usr/bin/tesseract'
text = pytesseract.image_to_string(Image.open('./output.png'), lang='eng' )
→# config='--psm 11'
print(text)

```

(2, 1024, 1024, 3)

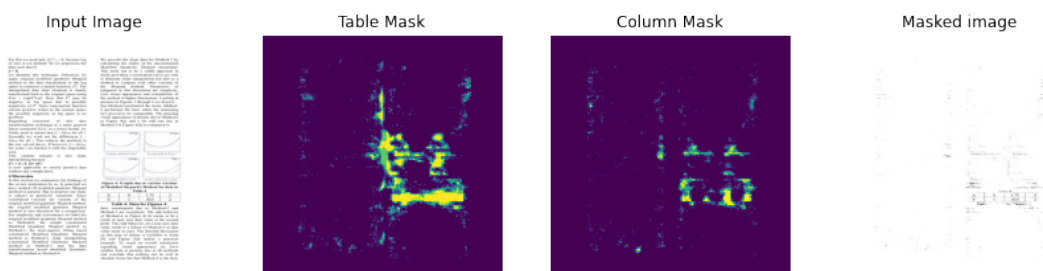


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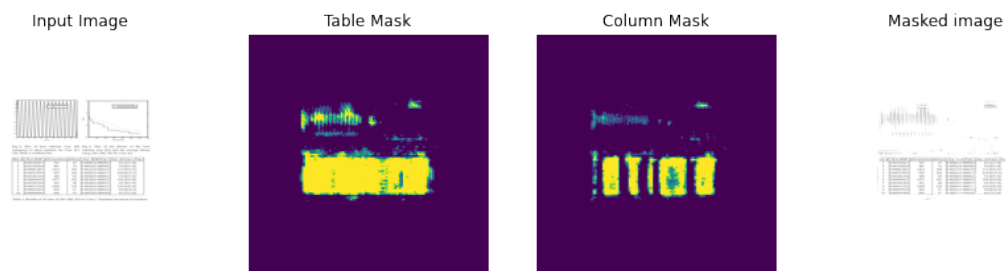
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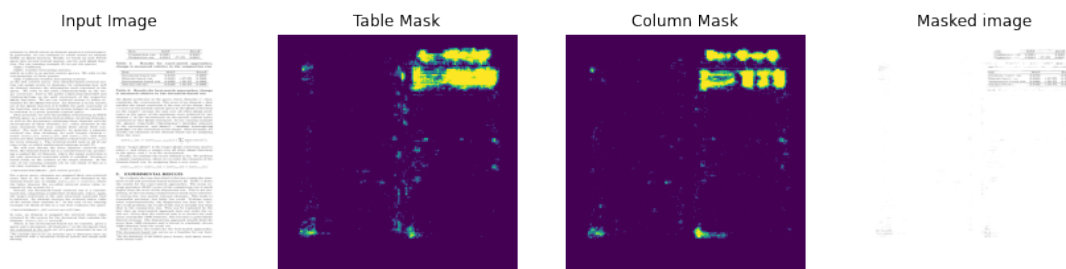
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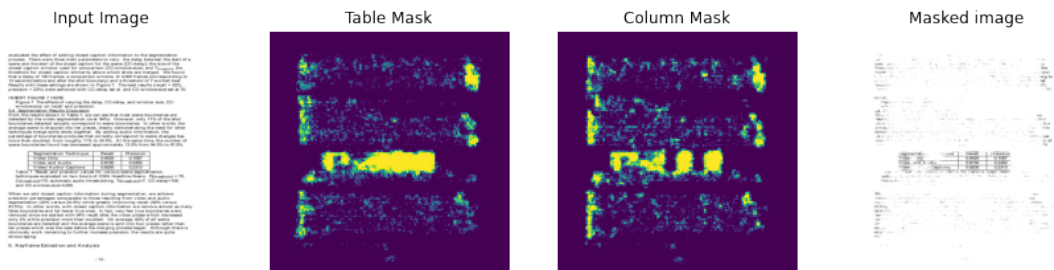
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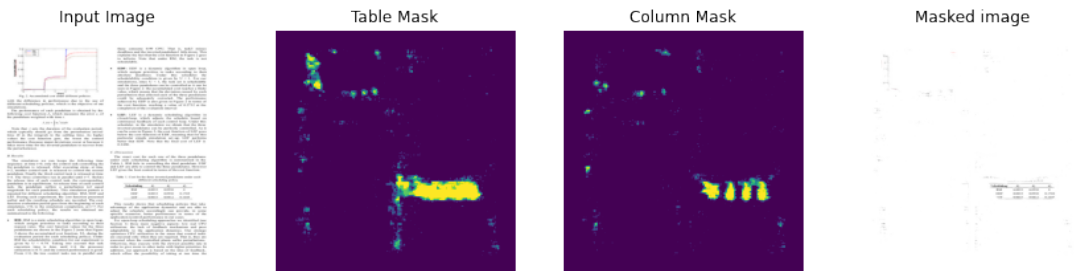
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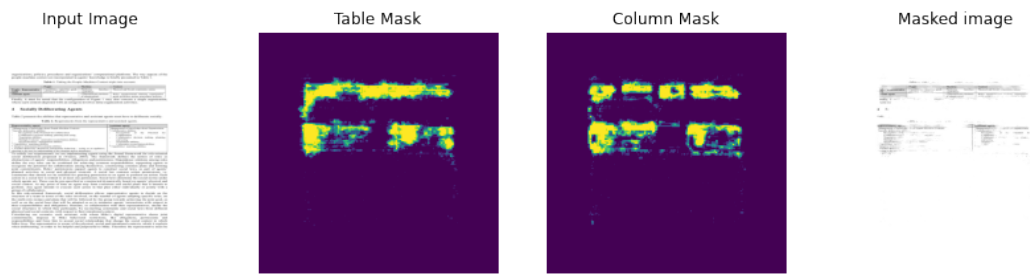
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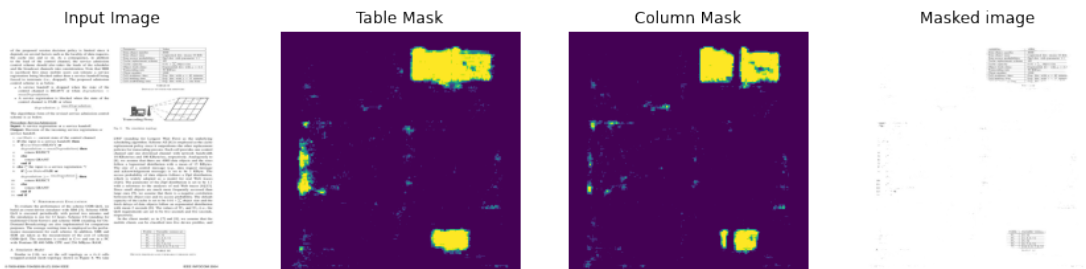
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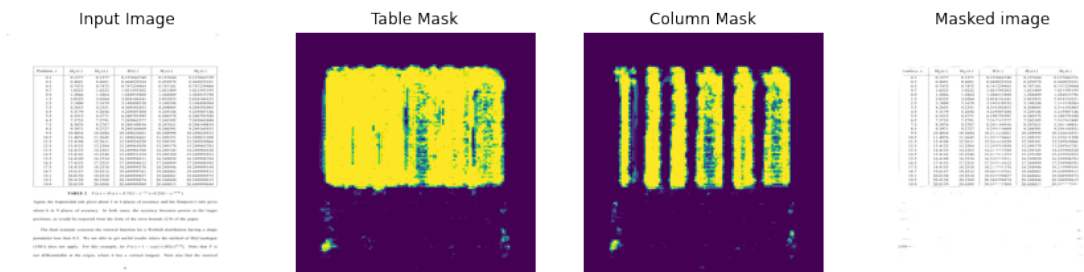
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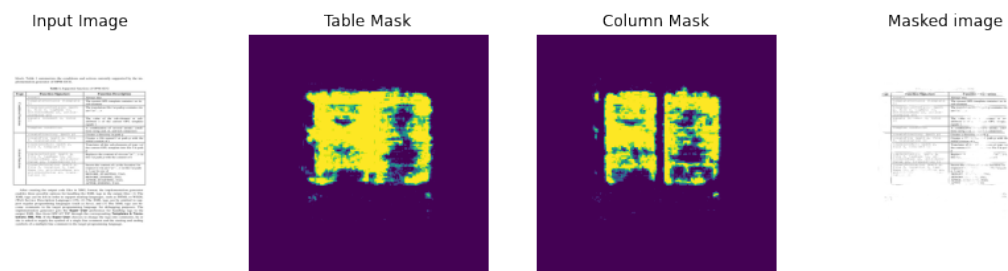
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