- Import Libraries

wandb.login()

```
#Import all libraries
#Install EMIST library, import datasets of letters, Matplotlib
!pip3 install emnist
from emnist import list_datasets
from emnist import extract_training_samples
import matplotlib as mpl
import os
import numpy as np #linear algebra
import pandas as pd
from pandas_profiling import ProfileReport
import matplotlib.pyplot as plt
%matplotlib inline
import torch
import torchvision
import tensorflow as tf
import torch
import torch.nn as nn
print(tf.__version__)
import torch
import cv2
import torchvision.transforms as transforms
from tensorflow.keras.layers import *
from tensorflow.keras.models import *
import tensorflow datasets as tfds
# import numpy as np
# import torch.nn.functional as F
# from torchvision.datasets import EMNIST
# from torch.utils.data import DataLoader
# import torchvision.transforms as tt
# from torch.utils.data import random split
# from torchvision.utils import make grid
# from tensorflow.keras.datasets import mnist
# from tensorflow.keras.utils import to_categorical
# from tensorflow.keras.models import Sequential
# from tensorflow.keras.layers import Conv2D
# from tensorflow.keras.layers import MaxPooling2D
# from tensorflow.keras.layers import Dense
# from tensorflow.keras.layers import Flatten
# from tensorflow.keras.optimizers import SGD
from PIL import Image
!pip install pyyaml h5py
import os
# from tensorflow import keras
# !pip install extra_keras_datasets
# from extra_keras_datasets import emnist
!pip install wandb
import wandb
from wandb.keras import WandbCallback
```

```
!pip install tqdm
from numpy import argmax
from keras.preprocessing.image import load_img
from keras.preprocessing.image import img_to_array
from keras.models import load_model

!pip install tqdm
from tqdm import tqdm
import matplotlib
import matplotlib
import matplotlib.pyplot as plt
%matplotlib inline
matplotlib.rcParams['figure.facecolor'] = '#ffffff'
```

```
tqdm in /usr/local/lib/python3.7/dist-packages (from emnist) (4.63.0)
        requests in /usr/local/lib/python3.7/dist-packages (from emnist) (2.23.0)
        numpy in /usr/local/lib/python3.7/dist-packages (from emnist) (1.21.5)
        idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->emnist) (2.10)
        certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests->emnist) (2021.10
        urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (from reques
        chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->emnist) (3.0.4)
        emnist
       0.0
        pyyaml in /usr/local/lib/python3.7/dist-packages (3.13)
        h5py in /usr/local/lib/python3.7/dist-packages (3.1.0)
        numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages (from h5py) (1.21.5)
        cached-property in /usr/local/lib/python3.7/dist-packages (from h5py) (1.5.2)
       2.py3-none-any.whl (1.7 MB)
             1.7 MB 3.9 MB/s
       .0
       .4.0-py2.py3-none-any.whl (9.0 kB)
        nvthon-dateutil>=2.6.1 in /usr/local/lib/nvthon3.7/dist-nackages (from wandh) (2.8.2)
  #Download dataset and it is 536 MB
  list_datasets()
       Downloading emnist.zip: 43.9MB [00:01, 40.5MB/s]
        KeyboardInterrupt
                                                  Traceback (most recent call last)
        <ipython-input-2-aa4b3a38056e> in <module>()
              1 #Download dataset and it is 536 MB
        ----> 2 list_datasets()
                                           4 frames
        /usr/local/lib/python3.7/dist-packages/urllib3/response.py in stream(self, amt,
        decode content)
           493
                                yield line
            494
                        else:
        --> 495
                            while not is_fp_closed(self._fp):
                                data = self.read(amt=amt, decode content=decode content)
            496
            497
        KeyboardInterrupt:
         SEARCH STACK OVERFLOW
             I 63 kB 1.3 MB/s
   images, labels = extract_training_samples('byclass')
  images.shape
        urllib3!=1.25.0.!=1.25.1.<1.26.>=1.21.1 in /usr/local/lib/nvthon3.7/dist-nackages (from request
GPU Running
       nackagas · nathtools
  # Get the GPU device name.
  def get_default_device():
       """Pick GPU if available, else CPU"""
      if torch.cuda.is_available():
          return torch.device('cuda')
      else:
```

one-any.whl (7.3 kB)

return torch.device('cpu')

```
def to_device(data, device):
    """Move tensor(s) to chosen device"""
    if isinstance(data, (list,tuple)):
        return [to_device(x, device) for x in data]
    return data.to(device, non_blocking=True)
class DeviceDataLoader():
    """Wrap a dataloader to move data to a device"""
    def __init__(self, dl, device):
        self.dl = dl
        self.device = device
    def __iter__(self):
    """Yield a batch of data after moving it to device"""
        for b in self.dl:
            yield to_device(b, self.device)
    def __len__(self):
        """Number of batches"""
        return len(self.dl)
# Returns current working runtype
device = get_default_device()
device
```

DataLoader Preparation

```
transform=tt.Compose([
                    lambda img: tt.functional.rotate(img, -90),
                    lambda img: tt.functional.hflip(img),
                    tt.ToTensor()
                ]))
random\_seed = 50
torch.manual_seed(random_seed);
val_size = 50000
train_size = len(dataset) - val_size
train_ds, val_ds = random_split(dataset, [train_size, val_size])
len(train_ds), len(val_ds)
batch\_size = 400
train_dl = DataLoader(train_ds, batch_size, shuffle=True, num_workers=4, pin_memory=True)
val_dl = DataLoader(val_ds, batch_size*2, num_workers=4, pin_memory=True)
# Lets see a batch of images
def show_batch(dl):
    for images, labels in dl:
        fig = 2x - pl + subplots (figsize - (12 - 12))
```

dataset = EMNIST(root="data/", split="byclass", download=True, train=True,

```
ax.set_xticks([]); ax.set_yticks([])
ax.imshow(make_grid(images, nrow=20).permute(1, 2, 0))
break
show_batch(train_dl)

train_dl = DeviceDataLoader(train_dl, device)
val_dl = DeviceDataLoader(val_dl, device)
```

show_batch(train_dl_1)

MNIST

MNIST recognition refernce

```
# save the final model to file
# load train and test dataset
def load_dataset(data):
   # load dataset
    (trainX, trainY), (testX, testY) = data
    # reshape dataset to have a single channel
    trainX = trainX.reshape((trainX.shape[0], 28, 28, 1))
    testX = testX.reshape((testX.shape[0], 28, 28, 1))
    # one hot encode target values
    trainY = to_categorical(trainY)
    testY = to_categorical(testY)
    return trainX, trainY, testX, testY
# scale pixels
def prep_pixels(train, test):
    # convert from integers to floats
    train_norm = train.astype('float32')
    test_norm = test.astype('float32')
    # normalize to range 0-1
    train norm = train norm / 255.0
    test_norm = test_norm / 255.0
    # return normalized images
    return train_norm, test_norm
# define cnn model
def define_model():
    model = Sequential()
 #(2,2) size of pooling area for max pooling #(3,3) convolution kernel size #32 number of convo
    model.add(Conv2D(32, (3, 3), activation='relu', kernel_initializer='he_uniform', input_shape=(2
    model.add(MaxPooling2D((2, 2)))
    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))
    model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform'))
    model.add(MaxPooling2D((2, 2)))
    model.add(Flatten())
    model.add(Dense(100, activation='relu', kernel_initializer='he_uniform'))
    model.add(Dense(62, activation='softmax'))
```

```
# compile model
    opt = SGD(learning_rate=0.01, momentum=0.9)
    model.compile(optimizer=opt, loss='categorical_crossentropy', metrics=['accuracy'])
    return model
# run the test harness for evaluating a model
def run_test_harness(data):
    # load dataset
    trainX, trainY, testX, testY = load_dataset(data)
    # prepare pixel data
    trainX, testX = prep_pixels(trainX, testX)
    # define model
    model = define_model()
    # fit model
    model.fit(trainX, trainY, epochs=10, batch_size=32, verbose=0)
    # save model
    model.save('/content/sample_data/final_model.h5')
# entry point, run the test harness
data=mnist.load_data()
run_test_harness(data)
     NameError
                                                Traceback (most recent call last)
     <ipython-input-2-61db1c596acd> in <module>()
           1 # entry point, run the test harness
     ----> 2 data=mnist.load_data()
           3 run_test_harness(data)
     NameError: name 'mnist' is not defined
      SEARCH STACK OVERFLOW
```

→ SAVE MODEL

```
# load and prepare the image
def load image(filename):
    # load the image
    img = load_img(filename, color_mode='grayscale', target_size=(28, 28))
    # convert to array
    img = img_to_array(img)
    # reshape into a single sample with 1 channel
    img = img.reshape(1, 28, 28, 1)
    # prepare pixel data
    img = img.astype('float32')
    img = img / 255.0
    return img
# load an image and predict the class
def run example():
    # load the image
    img = load_image('/content/sample_data/img.png')
```

```
# load model
model = load_model('/content/sample_data/final_model.h5')
# predict the class
predict_value = model.predict(img)
digit = argmax(predict_value)
print(digit)

# entry point, run the example

print('RESULT:')
run_example()
```

EMNIST DATASET RECOGNITION

Reference

EMNIST Classification

EMNIST Classification NB

research Paper

CNN Theory

SAVE AND RESTORE MODEL FROM WANDB

github alphabet recognition

TO FIND OUT WHAT IS TESTLOADER AND HOW TO GIVE FRAME BY FRAME IN VALIDATION DATA? or emnist classification predicts on one image

RESEARCH FOR AROUBA AND SAIMA

Points to note:

- We are using bymerge variant of EMNIST dataset. Here labels like j, o, i etc which look like J, O, I are merged.
- The EMNIST images provided here are inverted horizontally and rotated 90 anti-clockwise. For the
 ease of experimentation, we don't want to use it in this configuration. Thus we will rotate the image
 back by 90 deg anti-clockwise.
- We have a total of 814255 images. They are 28x28 pixels in resolution with only one channel.
- We have 697932 images as training data.
- We have 116323 images as testing data.
- We have 47 classes as shown:

```
# Gather EMNIST/bymerge dataset
train_ds, validation_ds = tfds.load(
    "emnist/bymerge",
    split=["train[:85%]", "train[85%:]"],
    as_supervised=True
)
```

```
LABELS = ['0', '1', '2', '3', '4', '5', '6', '7', '8', '9',
          'A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'J', 'K', 'L', 'M', 'N', 'O', 'P', 'Q', 'R',
          'a', 'b', 'd', 'e', 'f', 'g', 'h', 'n', 'q', 'r', 't']
len(LABELS)
     Downloading and preparing dataset emnist/bymerge/3.0.0 (download: 535.73 MiB, generated: Unkno
     DI Completed...: 100%
                           1/1 [00:17<00:00, 12.60s/ url]
     DI Size...: 100%
                      535/535 [00:17<00:00, 51.09 MiB/s]
     Extraction completed...: 100%
                               1/1 [00:17<00:00, 17.55s/ file]
     Extraction completed...: 100%
                                 4/4 [00:12<00:00, 3.57s/ file]
     Shuffling and writing examples to /root/tensorflow_datasets/emnist/bymerge/3.0.0.incompleteGZV
                                                    697931/697932 [00:03<00:00, 288506.06 examples/s]
     100%
     Shuffling and writing examples to /root/tensorflow_datasets/emnist/bymerge/3.0.0.incompleteGZV
                                                    116322/116323 [00:00<00:00, 230190.69 examples/s]
     Dataset emnist downloaded and prepared to /root/tensorflow_datasets/emnist/bymerge/3.0.0. Subs
     47
AUTO = tf.data.experimental.AUTOTUNE
BATCH_SIZE = 256
## We are transposing to rotate the image by 90 deg clockwise making the images human friendly.
def transpose_and_flatten(image, label=None):
  image = tf.image.convert_image_dtype(image, dtype=tf.float32) # scale image pixels to [0,1]
  image = tf.transpose(image, [1,0,2]) # transpose to get human friendly image, since rotation
  image = tf.reshape(image, shape=(784,)) # permutation invariant or flatten
  label = tf.one hot(label, depth=len(LABELS)) # one hot encode label
  return image, label
trainloader = (
    train_ds
    .shuffle(1024)
    .map(transpose_and_flatten, num_parallel_calls=AUTO)
    .batch(BATCH_SIZE)
    .prefetch(AUTO)
)
testloader = (
    validation ds
    .map(transpose_and_flatten, num_parallel_calls=AUTO)
    .batch(BATCH_SIZE)
    .prefetch(AUTO)
)
```

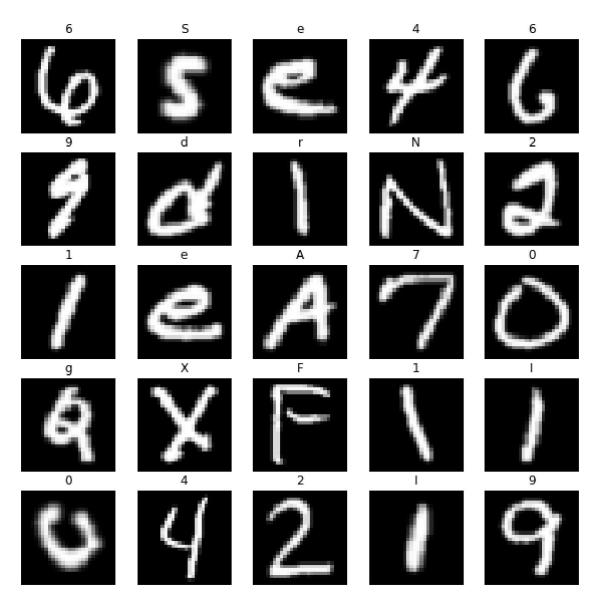
#4/ Classes

testloader

```
<PrefetchDataset element_spec=(TensorSpec(shape=(None, 784), dtype=tf.float32, name=None), Ten</pre>
```

```
imgs, labels = next(iter(trainloader))

plt.figure(figsize=(10, 10))
for n in range(25):
    ax = plt.subplot(5, 5, n+1)
    plt.imshow(tf.reshape(imgs[n], shape=(28,28)), cmap='gray')
    plt.title(LABELS[np.argmax(labels[n])])
    plt.axis('off')
```



type(imgs[0])

tensorflow.python.framework.ops.EagerTensor

```
def DenseModel():
    inputs = Input(shape=(784,))
    x = Dense(256, activation='relu')(inputs)
    x = Dense(128, activation='relu')(x)
    outputs = Dense(len(LABELS), activation='softmax')(x) #Index recieved from LABELS list
    return Model(inputs=inputs, outputs=outputs)

early_stopper = tf.keras.callbacks.EarlyStopping(
```

```
monitor='val_loss', patience=5, verbose=0, mode='auto', restore_best_weights=True
)
# initialize wandb run
# do not change entity
wandb.init(entity='iit-bhu', project='emnist')
# hyperparameters
config = wandb.config
config.epochs = 70
config.learning_rate = 0.001
# model
tf.keras.backend.clear_session()
model = DenseModel()
# optimizer
optimizer = tf.keras.optimizers.Adam(learning_rate=config.learning_rate)
# compile
model.compile(optimizer, 'categorical_crossentropy', metrics=['acc'])
# train
model.fit(trainloader,
     epochs=config.epochs,
     validation_data=testloader,
     callbacks=[WandbCallback(),
            early_stopper])
  wandb: Currently logged in as: insi29 (use `wandb login --relogin` to force relogin)
  Tracking run with wandb version 0.12.11
  Run data is saved locally in /content/wandb/run-20220304_152316-1n4moujo
  Syncing run fluent-wave-243 to Weights & Biases (docs)
  Epoch 1/70
  Epoch 2/70
  Epoch 3/70
  2318/2318 [================= ] - 157s 68ms/step - loss: 0.3550 - acc: 0.8757 - val
  Epoch 4/70
  Epoch 5/70
  Epoch 6/70
  Epoch 7/70
  Epoch 8/70
  Epoch 9/70
  Epoch 10/70
  <keras.callbacks.History at 0x7fd2f0efba10>
```

→ SAVE AND RESTORE FROM WANDB

```
api = wandb.Api()
   run = api.run("iit-bhu/emnist/1n4moujo")
  model.save(os.path.join(wandb.run.dir, "model.h5"))
  # Save a model file manually from the current directory:
  wandb.save('model.h5')
        ['/content/wandb/run-20220304_152316-1n4moujo/files/model.h5']
  # restore the model file "model.h5" from a specific run by user "lavanyashukla"
  # in project "save_and_restore" from run "10pr4joa"
  best_model = wandb.restore('model.h5', run_path="iit-bhu/emnist/1n4moujo")
  # use the "name" attribute of the returned object
  # if your framework expects a filename, e.g. as in Keras
  model.load_weights(best_model.name)

→ SAVE AND LOAD MODEL

  model.save('/content/sample_data/final_model.h5')
  model = load_model('/content/sample_data/final_model.h5')

→ SAVE AND LOAD MODEL FROM DRIVE

   !pip install pyyaml h5py
  from google.colab import drive
  drive.mount('/content/gdrive')
       Requirement already satisfied: pyyaml in /usr/local/lib/python3.7/dist-packages (3.13)
       Requirement already satisfied: h5py in /usr/local/lib/python3.7/dist-packages (3.1.0)
       Requirement already satisfied: cached-property in /usr/local/lib/python3.7/dist-packages (from
       Requirement already satisfied: numpy>=1.14.5 in /usr/local/lib/python3.7/dist-packages (from h
       Mounted at /content/gdrive
  model.save('/content/gdrive/MyDrive/FYP@10Pearls/final.h5')
```

→ Prediciton of Test Data

model_best=load_model('/content/gdrive/MyDrive/FYP@10Pearls/final.h5')

Convert Image to Tensor

```
import cv2
test=[]
def infer_prec(img, img_size):
    img = tf.expand_dims(img, -1)
                                          # from 28 x 28 to 28 x 28 x 1
    img = tf.divide(img, 255)
                                           # normalize
    img = tf.image.resize(img,
                                           # resize acc to the input
             [img_size, img_size])
                                           # reshape to add batch dimension
    img = tf.reshape(img,
           [784])
    return img
img = cv2.imread('/content/sample_data/sample_image.png', 0) # read image as gray scale
print(img.shape) # (720, 1280)
img = infer_prec(img, 28) # call preprocess function
print(img.shape)
img=tf.image.convert_image_dtype(
    img, dtype=tf.float32, name=None
test.append(img)
test=np.array(test)
     (1480, 1490)
     (784,)
plt.imshow(tf.reshape(img, shape=(28,28)), cmap='gray')
```

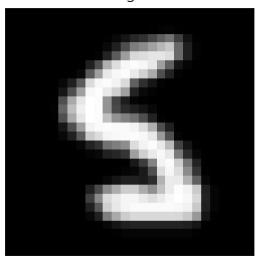
```
<matplotlib.image.AxesImage at 0x7fd2ef57c190>
type(test)
    numpy.ndarray
         100
tested=tf.convert_to_tensor(test, dtype=tf.float32)
         tested2=tf.convert_to_tensor(img_1, dtype=tf.float32)
           E 10 15 00 05
type(tested)
    tensorflow.python.framework.ops.EagerTensor
y_pred = model_best.predict(tested)
#y_pred # probabilities
# get predicted label
pred=tf.argmax(y_pred, axis=-1).numpy() # array([8], dtype=int64)
pred
    array([7])
LABELS[pred[0]]
    '7"
df=pd.DataFrame({'Expected Label':y_test,'Predicted Label':y_preds,'LABELS':data_p})
df.head(10)
        Expected Lahel Dredicted Lahel LABELS
```

	Expected Label	Predicted Label	LABELS
0	28	28	S
1	6	6	6
2	1	1	1
3	25	33	X
4	2	2	2
5	1	1	1
6	0	0	0
7	9	9	9
8	12	12	С
9	5	5	5

```
plt.figure(figsize=(10, 10))
ax = plt.subplot(2, 2, 1)
plt.imshow(tf.reshape(imgs2[0], shape=(28,28)), cmap='gray')
plt.title(LABELS[y_preds[0]])
plt.axis('off')
```

(-0.5, 27.5, 27.5, -0.5)

ς



DF=pd.DataFrame({'Labels':LABELS})
DF

	Labels			
0	0			
1	1			
2	2			
3	3			
4	4			
5	5			
6	6			
7	7			
8	8			
9	9			
10	Α			
11	В			
12	С			
13	D			
14	E			
15	F			
16	G			
17	Н			
18	1			
19	J			
20	K			
21	L			
22	M			
23	N			
24	0			
25	Р			
26	Q			
27	R			
LABELS[28]				
'S'				
3U	U 1			
y_preds[0]				
28 33	X			
y_test[0]				

__

Updates in Code

```
def cv2_imshow(img):
    plt.imshow(img,cmap='gray')
def predict_image(img):
  image = cv2.imread(img)
  gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
  blurred = cv2.GaussianBlur(gray, (5, 5), 0)
  edged = cv2.Canny(blurred, 30, 150)
  cnts = cv2.findContours(edged.copy(), cv2.RETR_EXTERNAL,
    cv2.CHAIN_APPROX_SIMPLE)
  cnts = imutils.grab_contours(cnts)
  cnts = sort_contours(cnts, method="left-to-right")[0]
  chars = []
   print(cnts)
 #cv2_imshow(edged)
  for c in cnts:
    # compute the bounding box of the contour
    (x, y, w, h) = cv2.boundingRect(c)
    # filter out bounding boxes, ensuring they are neither too small
    # nor too large
    if (w >= 5 \text{ and } w <= 150) and (h >= 15 \text{ and } h <= 120):
      # extract the character and threshold it to make the character
      # appear as *white* (foreground) on a *black* background, then
      # grab the width and height of the thresholded image
      roi = gray[y:y + h, x:x + w]
      thresh = cv2.threshold(roi, 0, 255,
        cv2.THRESH BINARY INV | cv2.THRESH OTSU)[1]
      (tH, tW) = thresh.shape
      # if the width is greater than the height, resize along the
      # width dimension
      if tW > tH:
        thresh = imutils.resize(thresh, width=32)
      # otherwise, resize along the height
      else:
        thresh = imutils.resize(thresh, height=32)
      (tH, tW) = thresh.shape
      dX = int(max(0, 32 - tW) / 2.0)
      dY = int(max(0, 32 - tH) / 2.0)
      # pad the image and force 32x32 dimensions
      padded = cv2.copyMakeBorder(thresh, top=dY, bottom=dY,
        left=dX, right=dX, borderType=cv2.BORDER_CONSTANT,
        value=(0, 0, 0))
      padded = cv2.resize(padded, (28, 28))
      padded = padded.astype("float32")
      padded = np.expand_dims(padded, axis=-1)
                                                   #originally -1
      chars.append((padded, (x, y, w, h)))
        cv2_imshow(padded)
    #print("Padded: -")
    #cv2_imshow(padded)
```

```
.....
  padded = cv2.copyMakeBorder(blurred, top=dY, bottom=dY,
           left=dX, right=dX, borderType=cv2.BORDER_CONSTANT,
           value=(0, 0, 0))
  padded = cv2.resize(padded, (32, 32))
# final_image=np.expand_dims(padded,axis=2)
  print("characters ",len(chars))
  boxes = [b[1] for b in chars]
  #cv2_imshow(chars)
  chars = np.array([c[0] for c in chars], dtype="float32")
 # OCR the characters using our handwriting recognition model
   for c in chars:
  preds = model.predict(chars)
 # define the list of label names
  labelNames = "0123456789"
  labelNames += "ABCDEFGHIJKLMNOPQRSTUVWXYZabcdefghijklmnopqrstuvwxyz"
  labelNames = [1 for 1 in labelNames]
  output=""
  fig=plt.figure()
  for c in range(len(chars)):
   print(c)
  # ax=fig.add_subplot(1,len(chars),c+1)
  #fig.show()
  for (pred, (x, y, w, h)) in zip(preds, boxes):
    i = np.argmax(pred)
  # print("i is ",i)
  # print("label length ",len(labelNames))
   #print("prediction is ",len(pred))
   prob = pred[i]
    label = labelNames[i]
   output+=label
  return output
predict image('/content/drive/MyDrive/Kaggle/aro.jpg')
```

NANO NETS

→ TRAINED DATA WITH MODEL

NANO NET MODEL

```
import requests
import json
url = 'https://app.nanonets.com/api/v2/OCR/Model/866d66dc-69d6-433c-82c8-6123cb2db3b6/LabelFile/'

data = {'file': open('/content/sample_data/opencv_frame_0.png', 'rb')}

response = requests.post(url, auth=requests.auth.HTTPBasicAuth('N1E3KM3daTBYb6TrHNUOJdHLMe1tSNLB',
    res=json.loads(response.text)
res["result"][0]["prediction"][0]["ocr_text"]
```