```
import pandas as pd
import numpy as np
import torch
import torchvision
import random
from PIL import Image
import cv2
import os
device = torch.device("cuda" if torch.cuda.is available() else "cpu")
!pip install evaluate jiwer
Collecting evaluate
  Downloading evaluate-0.4.0-py3-none-any.whl (81 kB)
                                        - 81.4/81.4 kB 2.9 MB/s eta
0:00:00
ent already satisfied: huggingface-hub>=0.7.0 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (0.10.1)
Requirement already satisfied: requests>=2.19.0 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (2.28.1)
Requirement already satisfied: xxhash in
/opt/conda/lib/python3.7/site-packages (from evaluate) (3.0.0)
Requirement already satisfied: packaging in
/opt/conda/lib/python3.7/site-packages (from evaluate) (21.3)
Requirement already satisfied: numpy>=1.17 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (1.21.6)
Requirement already satisfied: dill in /opt/conda/lib/python3.7/site-
packages (from evaluate) (0.3.5.1)
Requirement already satisfied: importlib-metadata in
/opt/conda/lib/python3.7/site-packages (from evaluate) (4.13.0)
Requirement already satisfied: responses<0.19 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (0.18.0)
Requirement already satisfied: fsspec[http]>=2021.05.0 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (2022.8.2)
Requirement already satisfied: datasets>=2.0.0 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (2.1.0)
Requirement already satisfied: multiprocess in
/opt/conda/lib/python3.7/site-packages (from evaluate) (0.70.13)
Requirement already satisfied: tqdm>=4.62.1 in
/opt/conda/lib/python3.7/site-packages (from evaluate) (4.64.0)
Requirement already satisfied: pandas in
/opt/conda/lib/python3.7/site-packages (from evaluate) (1.3.5)
Collecting levenshtein==0.20.2
  Downloading Levenshtein-0.20.2-cp37-cp37m-
manylinux 2 17 x86 64.manylinux2014 x86 64.whl (1.4 MB)
                                     — 1.4/1.4 MB 24.1 MB/s eta
0:00:00a 0:00:01
ent already satisfied: rapidfuzz<3.0.0,>=2.3.0 in
/opt/conda/lib/python3.7/site-packages (from levenshtein==0.20.2-
>iiwer) (2.11.1)
Requirement already satisfied: pyarrow>=5.0.0 in
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/opt/conda/lib/python3.7/site-packages (from datasets>=2.0.0-
>evaluate) (5.0.0)
Requirement already satisfied: aiohttp in
/opt/conda/lib/python3.7/site-packages (from datasets>=2.0.0-
>evaluate) (3.8.1)
Requirement already satisfied: typing-extensions>=3.7.4.3 in
/opt/conda/lib/python3.7/site-packages (from huggingface-hub>=0.7.0-
>evaluate) (4.1.1)
Requirement already satisfied: pyyaml>=5.1 in
/opt/conda/lib/python3.7/site-packages (from huggingface-hub>=0.7.0-
>evaluate) (6.0)
Requirement already satisfied: filelock in
/opt/conda/lib/python3.7/site-packages (from huggingface-hub>=0.7.0-
>evaluate) (3.7.1)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in
/opt/conda/lib/python3.7/site-packages (from packaging->evaluate)
(3.0.9)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in
/opt/conda/lib/python3.7/site-packages (from requests>=2.19.0-
>evaluate) (1.26.12)
Requirement already satisfied: charset-normalizer<3,>=2 in
/opt/conda/lib/python3.7/site-packages (from requests>=2.19.0-
>evaluate) (2.1.0)
Requirement already satisfied: idna<4,>=2.5 in
/opt/conda/lib/python3.7/site-packages (from requests>=2.19.0-
>evaluate) (3.3)
Requirement already satisfied: certifi>=2017.4.17 in
/opt/conda/lib/python3.7/site-packages (from requests>=2.19.0-
>evaluate) (2022.9.24)
Requirement already satisfied: zipp>=0.5 in
/opt/conda/lib/python3.7/site-packages (from importlib-metadata-
>evaluate) (3.8.0)
Requirement already satisfied: python-dateutil>=2.7.3 in
/opt/conda/lib/python3.7/site-packages (from pandas->evaluate) (2.8.2)
Requirement already satisfied: pytz>=2017.3 in
/opt/conda/lib/python3.7/site-packages (from pandas->evaluate)
(2022.1)
Requirement already satisfied: multidict<7.0,>=4.5 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (6.0.2)
Requirement already satisfied: async-timeout<5.0,>=4.0.0a3 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (4.0.2)
Requirement already satisfied: frozenlist>=1.1.1 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (1.3.0)
Requirement already satisfied: yarl<2.0,>=1.0 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (1.7.2)
Requirement already satisfied: asynctest==0.13.0 in
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/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (0.13.0)
Requirement already satisfied: aiosignal>=1.1.2 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (1.2.0)
Requirement already satisfied: attrs>=17.3.0 in
/opt/conda/lib/python3.7/site-packages (from aiohttp->datasets>=2.0.0-
>evaluate) (21.4.0)
Requirement already satisfied: six>=1.5 in
/opt/conda/lib/python3.7/site-packages (from python-dateutil>=2.7.3-
>pandas->evaluate) (1.15.0)
Installing collected packages: levenshtein, jiwer, evaluate
  Attempting uninstall: levenshtein
    Found existing installation: Levenshtein 0.20.7
    Uninstalling Levenshtein-0.20.7:
      Successfully uninstalled Levenshtein-0.20.7
ERROR: pip's dependency resolver does not currently take into account
all the packages that are installed. This behaviour is the source of
the following dependency conflicts.
python-levenshtein 0.20.7 requires Levenshtein==0.20.7, but you have
levenshtein 0.20.2 which is incompatible.
Successfully installed evaluate-0.4.0 jiwer-2.5.1 levenshtein-0.20.2
WARNING: Running pip as the 'root' user can result in broken
permissions and conflicting behaviour with the system package manager.
It is recommended to use a virtual environment instead:
https://pip.pypa.io/warnings/venv
"""Seed everything!"""
random.seed(42)
os.environ['PYTHONHASHSEED'] = str(42)
np.random.seed(42)
torch.manual seed(42)
torch.cuda.manual seed(42)
torch.backends.cudnn.deterministic = True
torch.backends.cudnn.benchmark = True
Load dataset
Tokenize car numbers
# Get the list of car numbers
def exec text(path):
    return path[path.find('-') + 1:path.find('.')]
input dir train = '/kaqqle/input/labtinkoff/CCPD2019-dl1/train'
car numbers = [exec text(path) for path in
os.listdir(input dir train)]
```

```
# Get the alphabet of symbols from all car numbers
seq = ''
for car_number in car numbers:
    seq += car number
alphabet = ''
for symbol in sorted(set(seq)):
    alphabet += symbol
alphabet
'0123456789ABCDEFGHJKLMNOPQRSTUVWXYZ 云京冀吉宁川新晋桂沪津浙渝湘琼甘皖粤苏蒙
藏豫贵赣辽鄂闽陕青鲁黑'
00V TOKEN = '<00V>' # out of vocabulary token
CTC_BLANK = '<BLANK>' # token for ctc matrix
PAD_TOKEN = '<PAD>' # padding token
def get char map(alphabet):
    """\overline{M}ake \overline{f}rom string alphabet character2int dict.
    Add BLANK char for CTC loss and OOV char for out of vocabulary
symbols."""
    char map = \{value: idx + 3 \text{ for } (idx, value) \text{ in } \}
enumerate(alphabet)}
    char map[CTC BLANK] = 0
    char map[00V TOKEN] = 1
    char map[PAD TOKEN] = 2
    return char map
class Tokenizer:
    """Class for encoding and decoding string word to sequence of int
    (and vice versa) using alphabet."""
    def __init__(self, alphabet):
        self.char map = get char map(alphabet)
        self.rev_char_map = {val: key for key, val in
self.char map.items()}
    def encode(self, word list):
        enc words = []
        for word in word list:
            enc_words.append(
                [self.char map[char] if char in self.char map
                 else self.char map[00V T0KEN]
                 for char in word]
        return enc words
    def get num chars(self):
        return len(self.char_map)
```

```
def decode(self, enc_word_list):
        dec words = []
        for word in enc word_list:
            word chars = ''
            for idx, char enc in enumerate(word):
                    char enc != self.char map[00V TOKEN]
                    and char enc != self.char map[CTC BLANK]
                    and not (idx > 0 \text{ and } char enc == word[idx - 1])
                    word chars += self.rev char map[char enc]
            dec words.append(word chars)
        return dec words
tokenizer = Tokenizer(alphabet)
class Laba dataset(torch.utils.data.Dataset):
    def init (self, root, tokenizer, transform=None):
        self.root = root
        self.transform = transform
        self.tokenizer = tokenizer
        self.img paths = [os.path.join(self.root, img path) for
img path in os.listdir(self.root)]
        self.text = [exec text(path) for path in
os.listdir(self.root)]
        self.enc_text = self.tokenizer.encode(self.text)
    def getitem__(self, ind):
        img = Image.open(self.img paths[ind]) # resize
        if self.transform is not None:
            img = self.transform(img) # make some augmentations
        # return image, encoded text, source text
        return (img, torch.LongTensor(self.enc text[ind]),
self.text[ind])
    def len (self):
        return len(self.img paths)
def collate fn(batch):
    images, enc texts, texts = zip(*batch)
    images = torch.stack(images, 0)
    enc_pad_texts = torch.nn.utils.rnn.pad_sequence(enc_texts,
batch_first=True, padding_value=tokenizer.char_map[PAD_TOKEN])
    return images, enc pad texts, texts
from sklearn.model selection import train test split
batch size = 128
transform = torchvision.transforms.Compose([
    torchvision.transforms.Resize((32, 128)),
    torchvision.transforms.RandomRotation(5),
```

```
torchvision.transforms.ColorJitter(),
    torchvision.transforms.GaussianBlur(3),
    torchvision.transforms.ToTensor()
        1)
dataset full = Laba dataset(input dir train, tokenizer,
transform=transform)
# split full dataset
train idx, valid idx =
train test split(list(range(len(dataset full))), train size=0.9)
dataset = {
    'train': torch.utils.data.Subset(dataset full, train idx),
    'valid': torch.utils.data.Subset(dataset full, valid idx)
}
dataset_size = {ds: len(dataset[ds]) for ds in ['train', 'valid']}
dataloader = {
    'train': torch.utils.data.DataLoader(
        dataset=dataset['train'], batch size=batch size, shuffle=True,
collate fn=collate fn
    'valid': torch.utils.data.DataLoader(
        dataset=dataset['valid'], batch_size=batch_size,
shuffle=False, collate fn=collate fn
}
input dir test = '/kaggle/input/labtinkoff/CCPD2019-dl1/test'
batch size = 128
transform test = torchvision.transforms.Compose([
    torchvision.transforms.Resize((32, 128)),
    torchvision.transforms.ToTensor()
dataset test = Laba dataset(input dir test, tokenizer,
transform=transform test)
dataloader test = torch.utils.data.DataLoader(
        dataset=dataset_test, batch_size=batch_size, shuffle=False,
collate_fn=collate fn
next(iter(dataloader['train']))[0].shape
torch.Size([128, 3, 32, 128])
img = torchvision.transforms.ToPILImage()(dataset full[173]
[0].squeeze(0))
imq
```



```
dataset test[122]
(tensor([[[0.5451, 0.5529, 0.5608,
                                    ..., 0.4157, 0.4118, 0.4706],
          [0.4431, 0.4039, 0.4039,
                                    ..., 0.4157, 0.5020, 0.5647],
          [0.3333, 0.2980, 0.3059,
                                     ..., 0.2353, 0.3137, 0.4353],
          [0.3333, 0.3137, 0.2667,
                                     ..., 0.1608, 0.1490, 0.1843],
          [0.3216, 0.3176, 0.3098,
                                    ..., 0.4471, 0.4706, 0.5020],
                                     ..., 0.3922, 0.4314, 0.4471]],
          [0.3255, 0.3333, 0.3529,
         [[0.5059, 0.5059, 0.5059,
                                     ..., 0.4745, 0.4706, 0.5373],
          [0.4275, 0.3686, 0.3608,
                                     \dots, 0.5216, 0.6000, 0.6627],
          [0.3569, 0.3020, 0.2980,
                                     ..., 0.3804, 0.4431, 0.5490],
          [0.4431, 0.4235, 0.3725,
                                    ..., 0.2353, 0.2392, 0.2824],
          [0.4157, 0.4157, 0.4078,
                                    ..., 0.5255, 0.5647, 0.6000],
          [0.4157, 0.4235, 0.4431,
                                     ..., 0.4784, 0.5255, 0.5412]],
                                    ..., 0.5451, 0.5176, 0.5333],
         [[0.6275, 0.6353, 0.6431,
          [0.5961, 0.5765, 0.6000,
                                    ..., 0.6000, 0.6353, 0.6471],
          [0.5490, 0.5725, 0.6471,
                                    ..., 0.5059, 0.4902, 0.5294],
          [0.4314, 0.4471, 0.4510, \ldots, 0.2980, 0.2039, 0.2000],
          [0.3882, 0.3922, 0.3922, \ldots, 0.5529, 0.5216, 0.5294],
          [0.3922, 0.3961, 0.4118, \ldots, 0.4784, 0.4863, 0.4941]])
 tensor([54, 13, 5, 8, 33, 8, 12]),
 '皖 A25V59')
Define model
from torch import nn
# To solve the problem, I used the standard CRNN structure
class ResNetBlock(nn.Module):
    def __init__(self, in_channels, out channels, kernel size=3,
stride=1, padding=0, dropout=0.15):
        super().__init__()
        self.conv = nn.Conv2d(in_channels, out_channels, kernel_size,
stride, padding, bias=False)
        self.bn = nn.BatchNorm2d(out channels)
        self.relu = nn.LeakyReLU()
        self.dropout = nn.Dropout(dropout)
        self.downsample = None
        if in channels != out channels:
            self.downsample = nn.Conv2d(in channels, out channels, 1,
stride=2)
    def forward(self, x, identity=True):
        out = self.dropout(self.bn(self.conv(x)))
        if identity:
```

```
if self.downsample is not None:
                x = self.downsample(x)
            return self.relu(out + x)
        else:
            return self.relu(out)
class CNN(nn.Module):
   def __init__(self, in_channels=1, num layers=2, dropout=0.1):
        super(). init ()
       As feature extractor i use resnet, passing through the cut the
images are
        transformed from the dimension tensor (C: 1, W: 128, H: 32) to
the
        dimension tensor (C: 1, W: 4, H: 1)
        self.start = ResNetBlock(3, 64, 7, 1, 0, 0.0)
        self.maxpool = nn.MaxPool2d(3, 2, 1)
        self.blocks1 = nn.ModuleList([ResNetBlock(64, 64, padding=1)
for _ in range(num layers)])
        self.blocks2 = nn.ModuleList([ResNetBlock(64, 128, padding=1,
stride=2)] + [ResNetBlock(128, 128, padding=1) for in
range(num layers)])
        self.blocks3 = nn.ModuleList([ResNetBlock(128, 256, padding=1,
stride=2)] + [ResNetBlock(256, 256, padding=1) for in
range(num layers)])
        self.blocks4 = nn.ModuleList([ResNetBlock(256, 512, padding=1,
stride=2)] + [ResNetBlock(512, 512, padding=1) for in
range(num layers)])
        self.blocks5 = nn.ModuleList([ResNetBlock(512, 1024,
padding=1, stride=2)] + [ResNetBlock(1024, 1024, padding=1) for in
range(num layers)])
        self.blocks = [self.blocks1, self.blocks2, self.blocks3,
self.blocks4, self.blocks5]
   def forward(self, x):
        out = self.maxpool(self.start(x, identity=False))
        for blocks in self.blocks:
            for block in blocks:
                out = block(out)
        return out
class BiLSTM(nn.Module):
   def init (self, input size, hidden size, num layers,
dropout=0.1):
        super().__init__()
        self.lstm = nn.LSTM(
            input size, hidden size, num layers,
            dropout=dropout, batch_first=True, bidirectional=True)
```

```
def forward(self, x):
        out, = self.lstm(x)
        return out
class CRNN(nn.Module):
    def init (
        self, number class symbols, time feature count=256,
lstm hidden=256,
        lstm len=3,
    ):
        super().__init__()
        self.feature extractor = CNN(dropout=0.15)
        self.avg pool = nn.AdaptiveAvgPool2d(
            (time_feature_count, time_feature_count))
        self.bilstm = BiLSTM(time feature count, lstm hidden,
lstm len, dropout=0.15)
        self.classifier = nn.Sequential(
            nn.Linear(lstm hidden * 2, time feature count),
            nn.GELU(),
            nn.Dropout(0.15),
            nn.Linear(time feature count, number class symbols)
        )
    def forward(self, x):
        x = self.feature extractor(x)
        b, c, h, w = x.size()
        x = x.view(b, c * h, w)
        x = self.avg pool(x)
        x = x.transpose(1, 2)
        x = self.bilstm(x)
        x = self.classifier(x)
        x = nn.functional.log softmax(x, dim=2).permute(1, 0, 2)
        return x
Define accuracy metric for evaluate validation dataset
class AverageMeter:
    def init (self):
        self.reset()
    def reset(self):
        self.avg = 0
        self.sum = 0
        self.count = 0
    def update(self, val, n=1):
        self.sum += val * n
        self.count += n
        self.avg = self.sum / self.count
```

```
def get_accuracy(y_true, y_pred):
    scores = []
    for true, pred in zip(y true, y pred):
        scores.append(true == pred)
    avg score = np.mean(scores)
    return avg score
Training loop
import pickle as pkl
def safe(obj, filename):
    with open(filename, 'wb') as outp:
        pkl.dump(obj, outp)
def read(filename):
    with open(filename, 'rb') as inp:
        return pkl.load(inp)
def weights init(m):
    classname = m.__class__.__name__
if type(m) in [nn.Linear, nn.Conv2d, nn.Conv1d]:
        torch.nn.init.xavier uniform (m.weight)
        if m.bias is not None:
            m.bias.data.fill (0.01)
    elif classname.find('BatchNorm') != -1:
        m.weight.data.normal (1.0, 0.02)
        m.bias.data.fill (0)
def val_loop(data_loader, model, tokenizer, device):
    acc avg = AverageMeter()
    for images, enc texts, texts in data loader:
        batch size = len(texts)
        text preds = predict(images, model, tokenizer, device)
        acc avg.update(get accuracy(texts, text preds), batch size)
    print(f'Validation, acc: {acc avg.avg:.4f}')
    return acc avg.avg
def predict(images, model, tokenizer, device):
    model.eval()
    images = images.to(device)
    with torch.no grad():
        output = model(images)
    pred = torch.argmax(output.detach().cpu(), -1).permute(1,
0).numpy()
    text preds = tokenizer.decode(pred)
    return text preds
def val loop ensemble(data loader, models, tokenizer, device):
```

```
acc avg = AverageMeter()
    for images, enc texts, texts in data loader:
        batch size = len(texts)
        text preds = predict ensemble(images, models, tokenizer,
device)
        acc avg.update(get accuracy(texts, text_preds), batch_size)
    print(f'Validation, acc: {acc avg.avg:.4f}')
    return acc avg.avg
def predict ensemble(images, models, tokenizer, device):
    [model.eval() for model in models]
    images = images.to(device)
    with torch.no grad():
        output = sum([model(images) for model in models]) /
len(models)
    pred = torch.argmax(output.detach().cpu(), -1).permute(1,
0).numpy()
    text preds = tokenizer.decode(pred)
    return text preds
def train loop(data loader, model, criterion, optimizer, epoch):
    loss avg = AverageMeter()
    model.train()
    for images, enc texts, texts in data loader:
        model.zero grad()
        images = images.to(device)
        batch size = len(texts)
        output = model(images)
        output lenghts = torch.full(
            size=(output.size(1),),
            fill value=output.size(0),
            dtype=torch.long
        )
        text lens = torch.LongTensor([len(text) for text in texts])
        loss = criterion(output, enc_texts, output_lenghts, text_lens)
        loss avg.update(loss.item(), batch size)
        loss.backward()
        torch.nn.utils.clip_grad_norm_(model.parameters(), 2)
        optimizer.step()
    for param group in optimizer.param groups:
        lr = param group['lr']
    print(f'\nEpoch {epoch}, Loss: {loss avg.avg:.5f}, LR: {lr:.7f}')
    return loss avg.avg
def train(dataloader, epochs):
    train loader, val loader = dataloader['train'],
dataloader['valid']
    model = CRNN(number class symbols=tokenizer.get num chars())
    model.apply(weights init)
    model.to(device)
```

```
criterion = torch.nn.CTCLoss(blank=0, reduction='mean',
zero infinity=True)
    optimizer = torch.optim.AdamW(model.parameters(), lr=0.001,
                                  weight decay=0.01)
    scheduler = torch.optim.lr scheduler.ReduceLROnPlateau(
        optimizer=optimizer, mode='max', factor=0.5, patience=5)
    best acc = -np.inf
    acc avg = val loop(val loader, model, tokenizer, device)
    for epoch in range(epochs):
        loss avg = train loop(train loader, model, criterion,
optimizer, epoch)
        acc avg = val loop(val loader, model, tokenizer, device)
        scheduler.step(acc avg)
        if acc avg > best acc:
            best acc = acc_avg
        safe(model, f'model {epoch}')
train(dataloader, 8)
Validation, acc: 0.0000
Epoch 0, Loss: 1.52695, LR: 0.0010000
Validation, acc: 0.8252
Epoch 1, Loss: 0.08662, LR: 0.0010000
Validation, acc: 0.9335
Epoch 2, Loss: 0.05117, LR: 0.0010000
Validation, acc: 0.9433
Epoch 3, Loss: 0.03628, LR: 0.0010000
Validation, acc: 0.9570
Epoch 4, Loss: 0.03567, LR: 0.0010000
Validation, acc: 0.9253
Epoch 5, Loss: 0.03618, LR: 0.0010000
Validation, acc: 0.9701
Epoch 6, Loss: 0.02495, LR: 0.0010000
Validation, acc: 0.9591
Epoch 7, Loss: 0.02684, LR: 0.0010000
Validation, acc: 0.9676
model = read('/kaggle/working/model_5') # load model with best acc on
validation
img, enc label, label = dataset full[2001]
```

```
pred = predict(img.unsqueeze(0).to(device), model, tokenizer, device)
# sample pred
pred
['皖KLJ029']
real img = torchvision.transforms.ToPILImage()(img)
real img
Compute metrics
from evaluate import load
cer = load("cer")
{"version_major":2,"version_minor":0,"model_id":"15cd030f4c6147509acee
f3bae3b6bf2"}
references = dataset test.text
predictions = []
for imgs, enc text, text in dataloader test:
    predictions += predict(imgs, model, tokenizer, device)
cer.compute(predictions=predictions, references=references)
0.008486562942008486
len(references) == len(predictions)
True
errors = {} # dict of errors {predictions: references}
for pred, refer in zip(predictions, references):
    if cer.compute(predictions=[pred], references=[refer]) != 0.0:
        errors[pred] = refer
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'皖 A26': '皖 ADD226',
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'鄂 A0450D6': '苏 D750D6',
'皖 AX83DR': '皖 AX830R',
'苏DB809J': '苏DR809J',
```

```
'皖 AAG490': '皖 AAR490', '京 N89N13': '豫 N89N13', '苏 E48F88': '赣 E48F88', '皖 A952U2', '苏 DB7208': '苏 DR7208', '浙 DQ7222': '渝 DQ7222', '皖 CJ0001': '渐 C06BS7'}
```

As we can see, the model is most wrong on Chinese characters, I tried to fix it with augmentations, but still there are numbers on which the model is wrong

```
key_err_0 = list(errors.values())[0]
ind_err_0 = dataset_test.text.index(key_err_0)
torchvision.transforms.ToPILImage()(dataset_test[ind_err_0][0])
```



```
list(errors.keys())[0]
```

'苏 AYC335'

key\_err\_0

'皖 AYC335'

acc\_avg = val\_loop(dataloader\_test, model, tokenizer, device)

Validation, acc: 0.9533

acc\_avg

0.9532953295329533