实验复现报告

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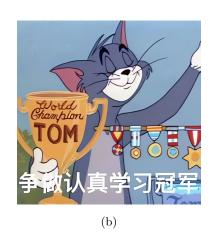
https://github.com/Game-learning/recurrent

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1 图像增强



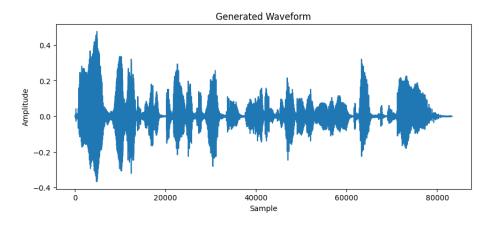


可以将一张模糊的照片增强为一张清晰的照片,提高图像的质量。

2 文本到语音转换

输入文本:

text = "Hello, this is a test of speech synthesis using Tacotron2."



可以将一段英文读成一个句子的语音。将中文文本转化为语音方面有待学习与探索。

3 文本到图片转换

输入文本:



可以根据英文提示词,生成一张对应的图片。

4 光学字符识别





图片 2a描述: a man and woman standing on a cliff overlooking a lake 图片 2b描述: a wolf is walking through a hole in the snow 可以识别图片,生成对图片的描述。

5 文本生成

当文本只有20句子时,学出来的结果可能是:

输入:

 $input_str = "Python is a good"$

生成的文本: Python is a good Python Python.. code topic science code data Python industry. industry. partsMachine partsMachine. partsMachine partsMachine.

当你的文本只有很大时,已经能够学习出句子的格式:

epochs = 10, Loss: 3.2292, 输入:

input_str = "I find it necessary"

生成的文本: I find it necessary to make a few changes to the code.

The first thing to do is to change the code to:

#include <iostream> #include <iostream> #include <iostream>

epochs = 20, Loss: 2.3805, 输入:

input_str = "I find it necessary"

生成的文本: I find it necessary to make a few changes to the code.

The first is to make the code a bit more readable.

The second is to make the code a bit more readable.

The third is to make the code a bit more readable

6 内容审核

输入文本:

"Those people are all garbage and don't deserve to live, You idiot, you can't do anything well."

类别	toxic	severe_toxic	obscene	threat	insult	identity_hate
是否存在	True	False	False	False	True	False

表 1: 内容判定情况

可以根据文本内容, 判断文本情感。

7 Python 代码

#####推荐使用虚拟环境, 安装所需的包, 避免出错。

```
#图像增强 (cmd)-----
git clone https://github.com/xinntao/Real-ESRGAN.git
cd Real-ESRGAN
# 安装 basicsr - https://github.com/xinntao/BasicSR
# 我们使用BasicSR来训练以及推断
pip install basicsr
# facexlib和gfpgan是用来增强人脸的
pip install facexlib
pip install gfpgan
pip install -r requirements.txt
python setup.py develop
wget https://github.com/xinntao/Real-ESRGAN/releases/download/v0.1.0/RealESR
\#python\ inference\_realesrgan.py\ -n\ RealESRGAN\_x4plus\ -i\ inputs\ --face\_enhance
python inference_realesrgan.py -n RealESRGAN_x4plus -i photo1.jpg ---face_enh
#文本到语音转换 —
import torch
import torchaudio
import matplotlib.pyplot as plt
from torchaudio.models import Tacotron2
from torchaudio.pipelines import TACOTRON2_WAVERNN_PHONE_LJSPEECH
#加载 Tacotron2 模型和 WaveRNN 声码器
bundle = TACOTRON2_WAVERNN_PHONE_LJSPEECH
```

```
processor = bundle.get_text_processor()
tacotron2 = bundle.get tacotron2()
vocoder = bundle.get vocoder()
# 将模型设置为评估模式
tacotron2.eval()
vocoder.eval()
# 输入文本
# 处理文本
with torch.inference_mode():
    processed, lengths = processor(text)
   # 生成梅尔频谱
   mel_spec, mel_lengths, _ = tacotron2.infer(processed, lengths)
   # 使用声码器生成波形
    waveforms, lengths = vocoder(mel_spec, mel_lengths)
# 可视化波形
plt. figure (figsize = (10, 4))
plt.plot(waveforms[0].cpu().numpy())
plt.title("Generated ⊔ Waveform")
plt.xlabel("Sample")
plt.ylabel("Amplitude")
plt.show()
# 保存音频文件
output_path = "output.wav"
torchaudio.save(output_path, waveforms.cpu(), sample_rate=22050)
\mathbf{print}(f"Audio_{\square}has_{\square}been_{\square}saved_{\square}to_{\square}\{output\_path\}")
```

```
#文本到图片转换 —
```

```
import requests
import io
from PIL import Image

#API_URL = "" # 你的API地址
#headers = {} # 你的请求头
def query(payload):
    response = requests.post(API_URL, headers=headers, json=payload)
    return response.content

image_bytes = query({
    "inputs": "Astronauturidinguauhorse", # 输入英文提示词
})

image = Image.open(io.BytesIO(image_bytes))
```

#光学字符识别 ————

from transformers import BlipProcessor, BlipForConditionalGeneration
from PIL import Image
import requests

#加载BLIP模型和处理器

processor = BlipProcessor.from_pretrained("Salesforce/blip-image-captioning-model = BlipForConditionalGeneration.from_pretrained("Salesforce/blip-image-

```
# 读取图片
image path = 'cc2.jpg' # 替换为你的图片路径
image = Image.open(image path)
# 预处理图片
inputs = processor(images=image, return_tensors="pt")
# 生成描述
out = model.generate(**inputs)
# 解码并输出描述
description = processor.decode(out[0], skip_special_tokens=True)
print("图片描述:", description)
#文本生成 -
import requests
url = "https://www.gutenberg.org/cache/epub/74840/pg74840.txt"
response = requests.get(url)
# 检查请求是否成功
if response.status_code = 200:
   # 读取文本内容
    text_data = response.text
   # 将文本分割成行
    lines = text_data.split('\n')
   #示例:打印前几行
    for i, line in enumerate(lines[:10]):
        \mathbf{print}(f^* \text{Line}_{\sqcup}\{i+1\}:_{\sqcup}\{\text{line}\}^*)
else:
    print ("Failed uto udownload uthe ufile.")
```

```
import torch
import torch.nn as nn
import torch.optim as optim
from transformers import GPT2LMHeadModel, GPT2Tokenizer
# 检查设备
device = "cuda" if torch.cuda.is_available() else "cpu"
#使用 GPT-2 模型 (带语言建模头)和对应的分词器
model = GPT2LMHeadModel.from_pretrained("gpt2").to(device)
tokenizer = GPT2Tokenizer.from_pretrained("gpt2")
vocab_size = tokenizer.vocab_size
# 定义损失函数和优化器
criterion = nn.CrossEntropyLoss()
optimizer = optim.Adam(model.parameters(), lr = 0.0001)
epochs = 10
# 数据处理类
class Corpus:
   def ___init___(self , data , tokenizer):
       self.data = data
        self.tokenizer = tokenizer
    def make_batch(self, batch_size):
       # 使用 GPT-2 分词器编码数据
       tokenized_data = [self.tokenizer.encode(text, return_tensors="pt").s
       # 拼接所有数据
```

```
inputs = torch.cat(tokenized_data[:-1])[:batch_size].to(device)
         targets = torch.cat(tokenized data[1:])[:batch size].to(device)
         return inputs, targets
# 示例数据
data = text\_data
corpus = Corpus (data, tokenizer)
batch\_size = 256
# 训练模型
for epoch in range (epochs):
    optimizer.zero_grad()
    inputs, targets = corpus.make_batch(batch_size)
    outputs = model(inputs.unsqueeze(0)).logits # GPT2LMHeadModel 的 logits
    loss = criterion(outputs.view(-1, vocab\_size), targets.view(-1))
# 计算损失
    loss.backward()
    optimizer.step()
    if (epoch + 1) \% 2 == 0:
         \mathbf{print}(f^*\operatorname{Epoch}_{\sqcup} \{\operatorname{epoch}_{\sqcup} + _{\sqcup} 1\}, _{\sqcup}\operatorname{Loss} : _{\sqcup} \{\operatorname{loss} . \operatorname{item}() : .4 f\}^*)
# 文本生成函数
def generate_text(model, input_str, max_len=50):
    model.eval()
    input_ids = tokenizer.encode(input_str, return_tensors="pt").to(device)
    output_ids = input_ids.tolist()[0]
    with torch.no grad():
         for _ in range(max_len):
```

```
# 获取 logits 并预测下一个 token
           outputs = model(input ids).logits
           next token id = torch.argmax(outputs[:, -1, :], dim=-1).item()
            if next_token_id == tokenizer.eos_token_id:
               break
            output_ids.append(next_token_id)
           input_ids = torch.tensor([output_ids], dtype=torch.long).to(devi
    return tokenizer.decode(output_ids, skip_special_tokens=True)
# 测试文本生成
input_str = "Kike_arrived_at"
generated_text = generate_text(model, input_str)
print("生成的文本:", generated_text)
# 测试文本生成
input_str = "Iufinduitunecessary"
generated_text = generate_text(model, input_str)
print("生成的文本:", generated_text)
#内容审核 ———
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.multioutput import MultiOutputClassifier
from sklearn.metrics import classification_report
# 加载数据集
train df = pd.read csv('train.csv')
```

```
test_df = pd.read_csv('test.csv')
test labels df = pd.read csv('test labels.csv')
sample submission df = pd.read csv('sample submission.csv')
# 随机提取100000个样本
train_sample = train_df.sample(n=100000, random_state=12)
test_sample = test_df.sample(n=100000, random_state=12)
#注意, test_labels中的ID需要与test_sample中的ID匹配
test_labels_sample = test_labels_df[test_labels_df['id'].isin(test_sample['i
# sample_submission中的ID也需要与test_sample中的ID匹配
sample_submission_sample = sample_submission_df[sample_submission_df['id'].i
# 保存样本数据
train_sample.to_csv('train_sample.csv', index=False)
test_sample.to_csv('test_sample.csv', index=False)
test_labels_sample.to_csv('test_labels_sample.csv', index=False)
sample_submission_sample.to_csv('sample_submission_sample.csv', index=False)
# 加载样本数据集
train_sample = pd.read_csv('train_sample.csv')
# 提取特征和标签
X = train_sample['comment_text']
y = train_sample[['toxic', 'severe_toxic', 'obscene', 'threat', 'insult', 'i
# 分割数据集
X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.2, rando
```

```
# 特征提取
vectorizer = TfidfVectorizer (max features=10000)
X train tfidf = vectorizer.fit transform(X train)
X_val_tfidf = vectorizer.transform(X_val)
# 训练多标签分类模型
model = MultiOutputClassifier(LogisticRegression(max_iter=1000))
model.fit(X_train_tfidf, y_train)
# 预测和评估
y_pred = model.predict(X_val_tfidf)
print(classification_report(y_val, y_pred, target_names=y.columns))
def predict_labels(comment):
    comment_tfidf = vectorizer.transform([comment])
    prediction = model.predict(comment_tfidf)
    labels = y.columns
    return {label: bool(pred) for label, pred in zip(labels, prediction[0])}
    # 示例句子
    comment = "Those \sqcup people \sqcup are \sqcup all \sqcup garbage \sqcup and \sqcup don't \sqcup deserve \sqcup to \sqcup live"
    prediction = predict_labels(comment)
    print(prediction)
    # 示例句子
    comment = "You \sqcup idiot, \sqcup you \sqcup can't \sqcup do \sqcup anything \sqcup well."
    prediction = predict_labels(comment)
    print(prediction)
```