TTI Prediction in Urban Road Network Using Computer Vision Techniques

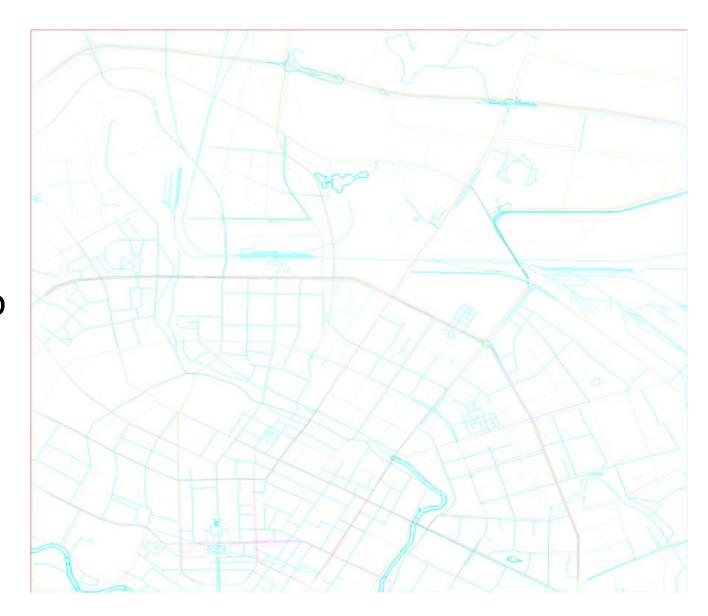
LI Xiahan

Modification

Renew the picture to white background

Renew the output from 1 to 21

Apply K-fold cross validation(now k=5)



Training Detail

- Dataset: Training and validation dataset: 20181013-20181017, 5 days.
 Len(Training):Len(Validation) = 4:1. Testing dataset: 20181018, 1 days
- Prediction: Predict 20 randomly selected individual road and 1 area
 TTI after 10 minutes
- Converge: because of the restriction of GPU usage of Kaggle (9 hours in maximum), none of neural network converge
- Neural Network: CNN, CNN+LSTM(series connection),
 CNN+LSTM(parallel connection), Historical Average(as baseline)

Deep Learning Result

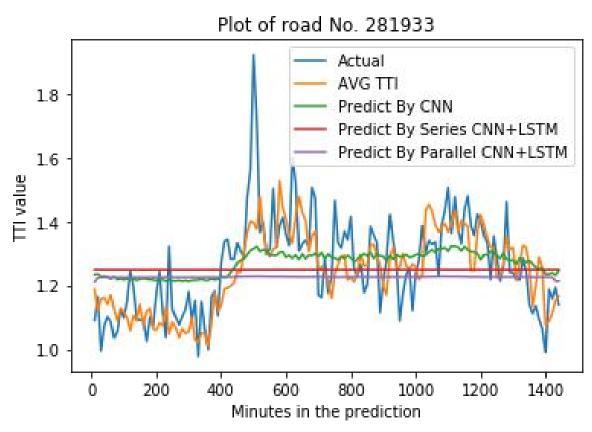
Actual: Actual TTI Value of the road

AVG TTI: Historical Average

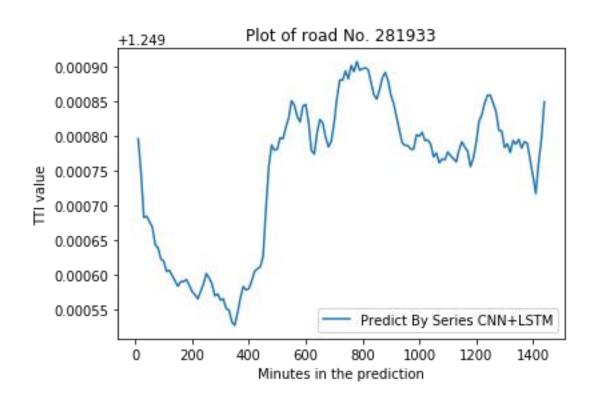
CNN: Resnet34

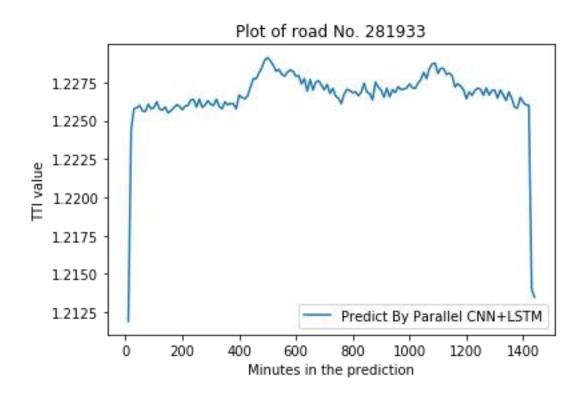
Series CNN+LSTM: Input of LSTM is the output of Resnet34, similar to the series connection in circult

Parallel CNN+LSTM: Combine the output of LSTM and output of Resnet34, similar to the parallel connection in circult



Deep Learning Result





Possible reason: LSTM is good for continuous data, but our data have 10 minutes interval, which is discrete.

Statistical Result

Goodness of prediction:
 CNN > CNN+LSTM(Series) =
 CNN+LSTM(Parallel) >
 Historical Average

According to MSELoss

```
In [3]: def Loss(df1, df2):
              total loss = 0.0
              for item in dfl. columns. tolist()[1:]:
                   for i in range (144):
                       total_loss += abs(np. power(dfl. at[i, item]-df2. at[i, item], 2))
              print('Total Loss in Average: ', total_loss)
In [4]: cnndf = pd. read_csv(os. path. join(path, 'CNNresult. csv'))
    [5]: actualdf = pd. read_csv(os. path. join(path, 'Actual. csv'))
    [6]: cnnlstmdf = pd. read csv(os. path. join(path, 'CNN+LSTM result. csv'))
    [7]: cnnlstmattdf = pd. read_csv(os. path. join(path, 'CNN+LSTM+Parallel_result. csv'))
In [8]: avgresultdf = pd. read_csv(os. path. join(path, 'Avg_TTI. csv'))
In [9]: Loss (cnndf, actualdf) #MSE Loss of results predicted by CNN
          Total Loss in Average: 291.94913895356495
   [10]: Loss (cnnlstmdf, actualdf) #MSE Loss of results predicted by CNN+LSTM series connection
          Total Loss in Average: 316.96470999038377
In [11]: Loss (avgresultdf, actualdf) #MSE Loss of results predicted by Historical Average
          Total Loss in Average: 453.12626849645585
   [12]: Loss(cnn1stmattdf, actualdf) #MSE Loss of results predicted by CNN+LSTM parallel connection
          Total Loss in Average: 316.9975012693749
```

Further Plan

- Begin to write the thesis
- Add exploration data analysis part, we need to statistically explain the data to reader.