

Course: CS672 – Introduction to Deep Learning

Project: #2

# Deep Learning for NYC Taxi Trip Duration Prediction

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- Professor: Tassos Sarbanes
- Date: Nov 7, 2025

# Agenda

- Introduction (Problem & Goal)
- Feature Engineering (Data Processing)
- Model Design(TensorFlow)
- PyTorch Version
- Results Comparison
- Conclusion
- Code Showcase



# Introduction (Problem & Goal)

## Background

- Predict the trip duration of NYC yellow cabs
- Data sources:
  - NYC Taxi Dataset (01/01/2020~01/31/2020)
  - Meteostat Weather Data (Wall Street, NYC)

## Goal

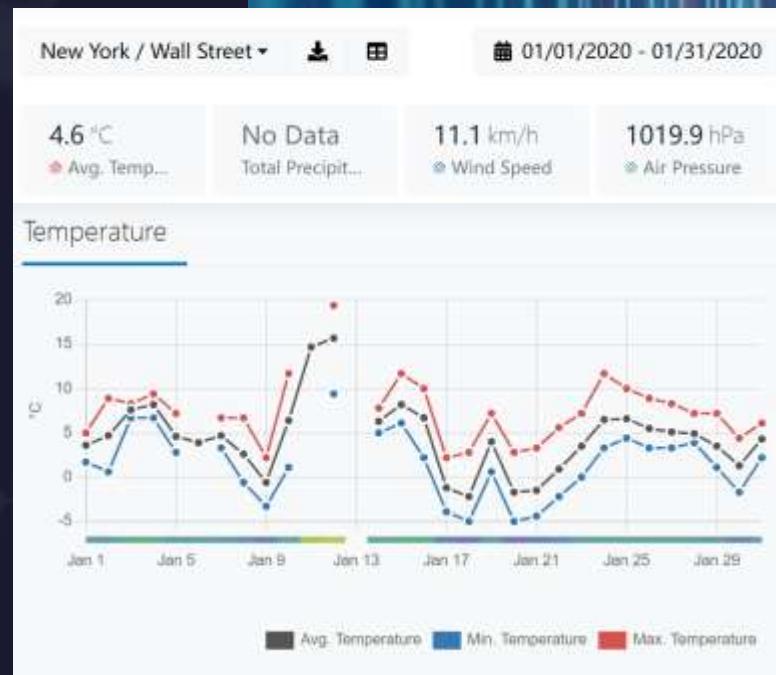
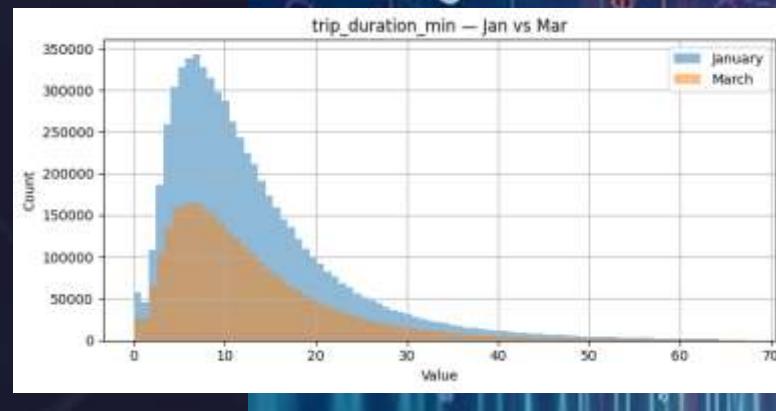
- Build a regression model based on neural networks
- Improve prediction accuracy by feature integration  
(in combination with meteorological characteristics)



# Feature Engineering (Data Processing)

## Processing Steps

1. Remove outliers (duration  $\leq 0$  or  $> 180$  minutes)
2. Extract time features: `hour`, `day_of_week`
3. Integrated weather features: `tavg`, `prcp`, `wspd`
4. Normalization: `log1p` (`trip_duration`)
5. Standardization (StandardScaler)



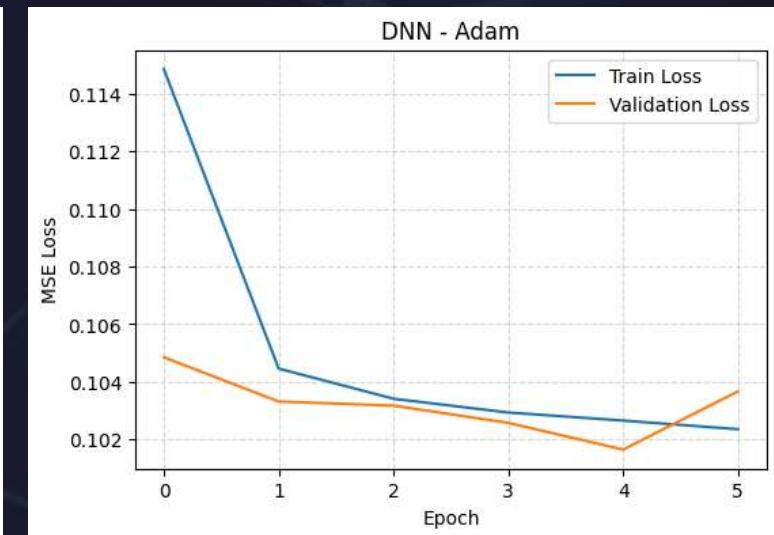
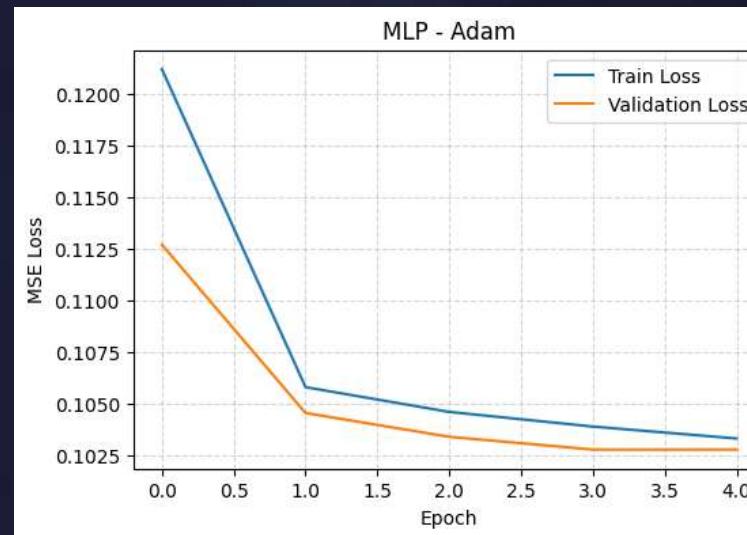
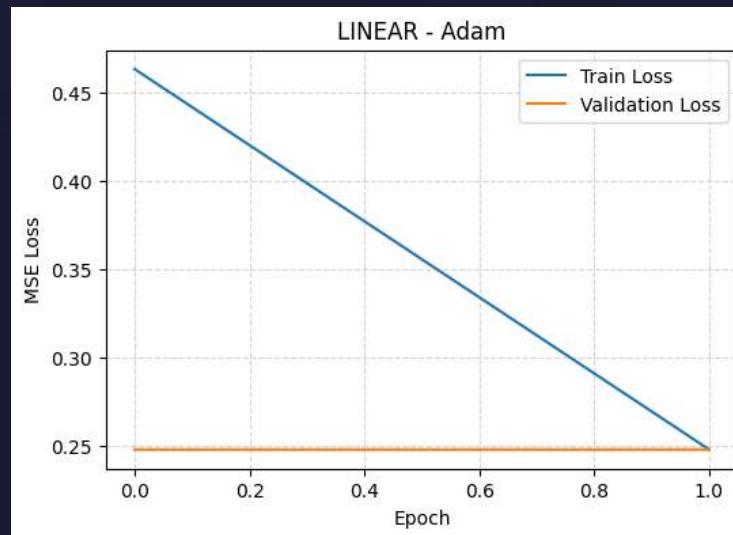
# Model Design (TensorFlow / Keras)



Types of Model	Hidden layer structure	Optimizer	Feature
Linear Regression	No hidden layer	Adam	Baseline Model
MLP (Multi-Layer Perceptron)	64–32	Adam / RMSProp	Simple nonlinearity
DNN(Deep Neural Network)	128–64–32	Adam	Optimal performance

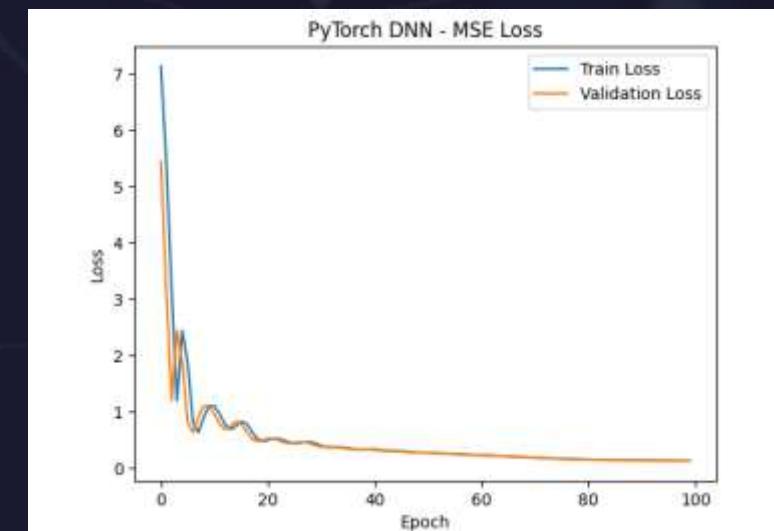


# Model Design (+Adam)

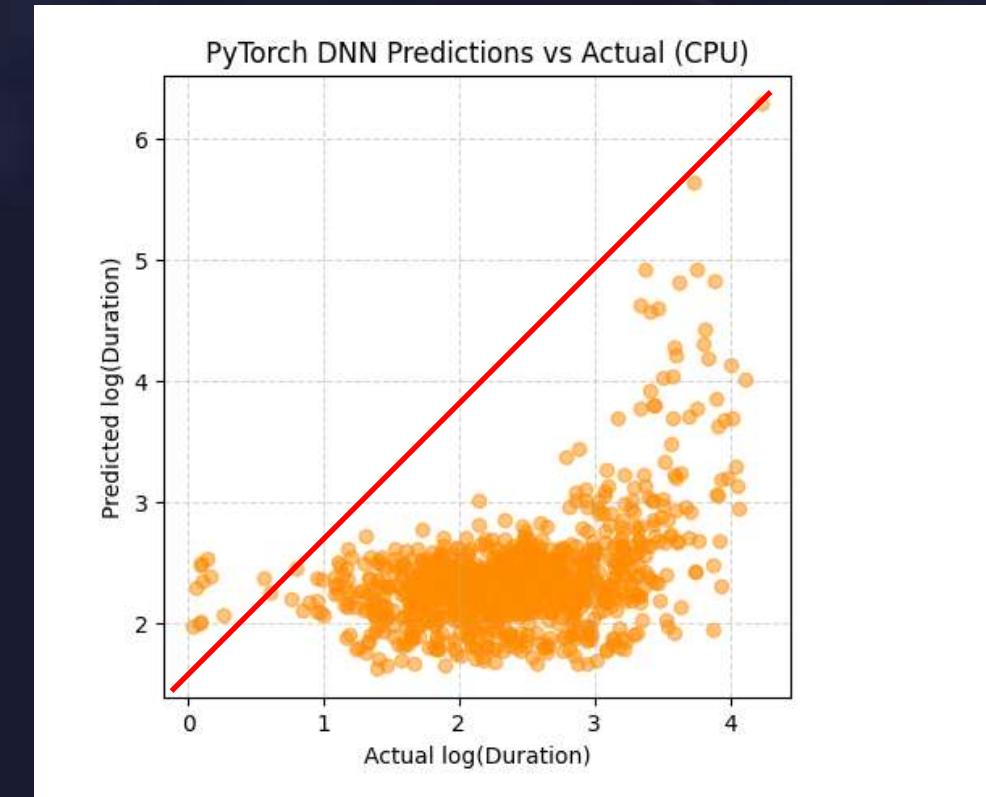
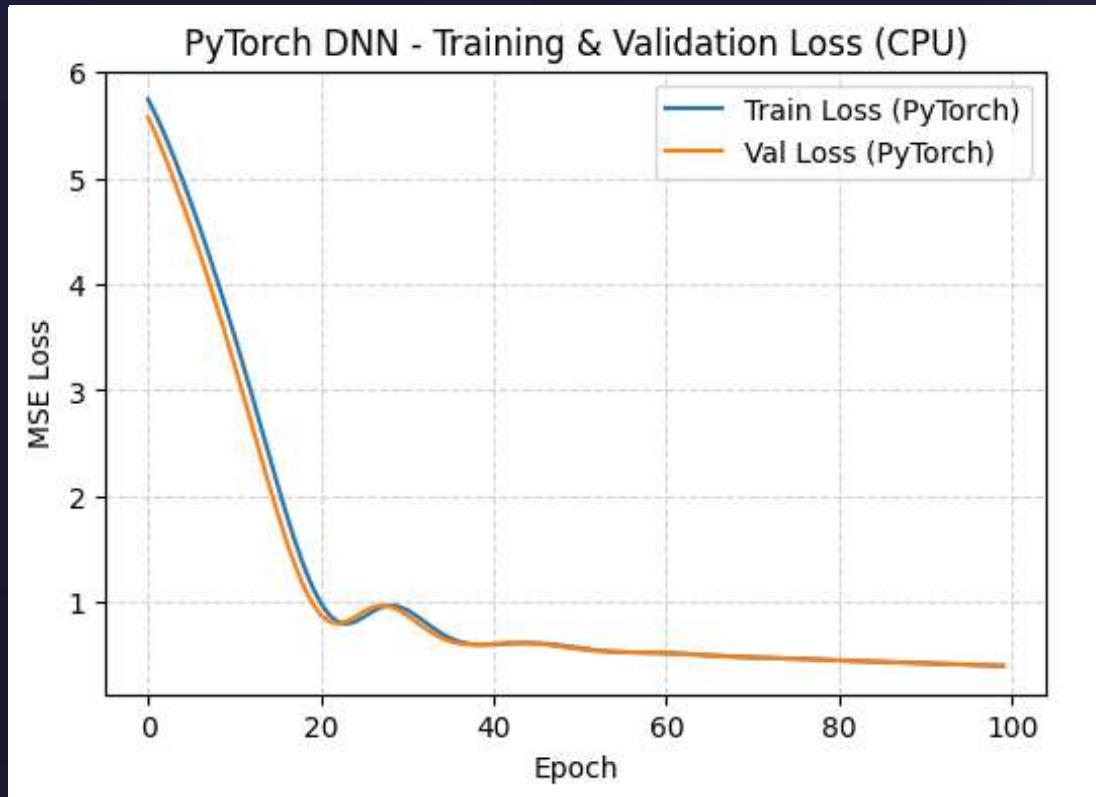


Linear Regression vs. MLP vs. DNN vs. PyTorch DNN

Train Loss Curve vs. Validate the Loss curve



# PyTorch Version DNN



Reproduce the DNN model in PyTorch

Use GPU acceleration for training

The verification result is close to TensorFlow

Prediction vs. Actual scatter plot

# Findings and analysis

Model	Val MSE	Val MAE	Framework
Linear	0.248185	0.381835	TF
MLP	0.102788	0.226832	TF
DNN	0.101652	0.224575	TF
DNN (PyTorch-CPU)	0.10296	0.225242	PT

Best Model: DNN | Optimizer: Adam | MAE: 0.2246

## Key points of the conclusion

- Improve the accuracy of predictions
- DNN has the best generalization and avoids overfitting
- Comparable to that of TF, which verifies the stability of PyTorch

# Conclusion & Future Work

- Deep learning can effectively model the nonlinear relationship of travel time
- The integration of weather data significantly enhances the prediction effect
- Future Expectations:
  - Time series modeling is conducted using Transformer
  - Add real-time traffic data
  - Predict multi-day trends



```
private int calculateLocationScore(User user) {
    double distance = calculateDistance(user.getLocation(), user2.getLocation());
    if (distance < MAX_DISTANCE) {
        return (int) Math.round((MAX_DISTANCE - distance) / MAX_DISTANCE * 100);
    }
    return 0;
}
```

# Code Showcase



# Q&A / Thank you

Project #2: Deep Learning for NYC Taxi Trip Duration Prediction

Team #4: Krishna Pothana & Xianrong Liang