

T_RTRANSRATER

PREDICT TRANSPORTATION COST EASIER

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Supported By Global Operations Science & Analytics Team

DHL Supply Chain - Excellence. Simply Delivered.

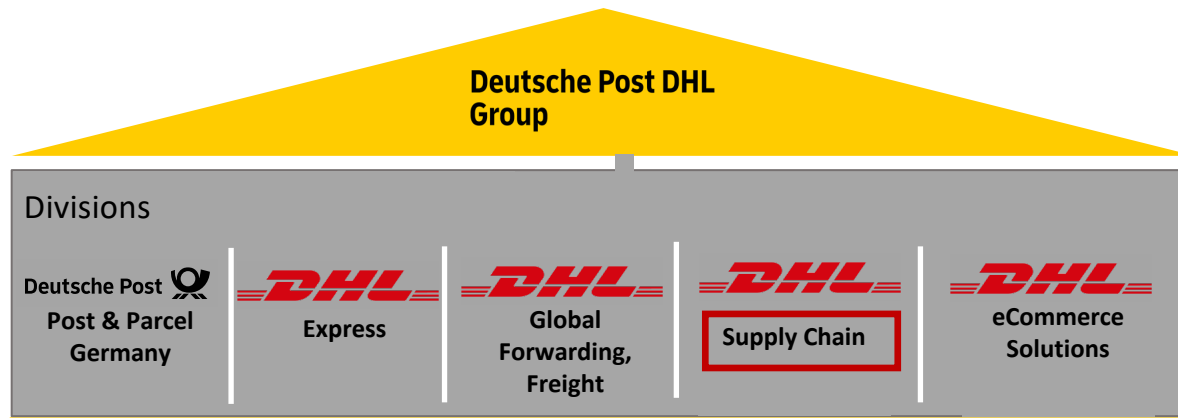
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About DHL Supply Chain



DHL Supply Chain is a division of Deutsche Post DHL Group with a global network and an extensive logistics portfolio that deals with warehousing, transport and VAS for other companies



DHL Supply Chain is the

#1

Contract Logistics Provider

Managing Supply Chains
to reduce complexities



50+
Countries Covered

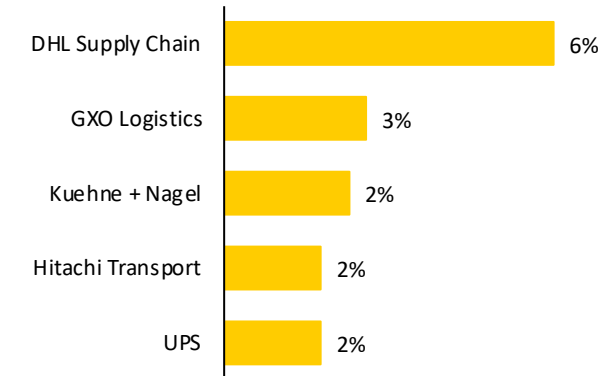


~160M
sq. footage of storage space

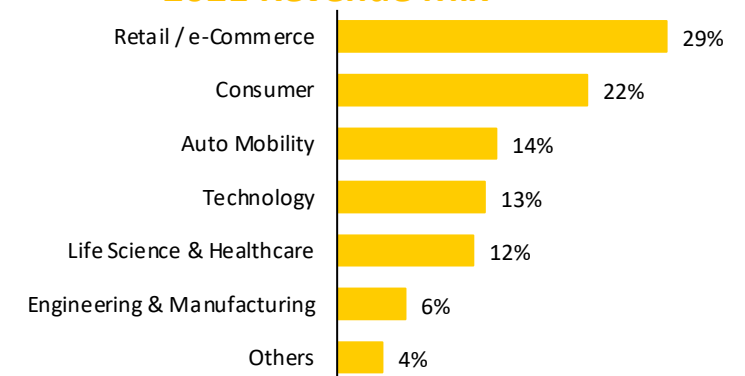


~ € 14bn
Revenue

Market Position



2021 Revenue Mix





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Project Overview



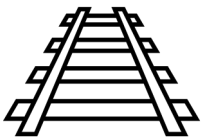
DHL Transportation Solution Team quotes third-party companies contract prices or spot prices to find feasible transportation solutions for client's shipments



TruckLoad Shipment (TL)



Less-Than-TruckLoad Shipment (LTL)



Intermodal Shipment (Railway)

DHL Transport Solution Team:

- The team considers quoted price in different data frames as a benchmark to find lowest price
- The team will save quoted price and historical shipments in the database

With available historical shipment data and quoted price data:

A Learning models is asked to **predict future transportation rates** by inputting available shipment information and quoted price.

Data Description



Data is provided by DHL Operation and Analytics Team. All data is for study use and has been masked

Historical Shipment Data (MT_Data)

Description: The dataset includes 4 month historical shipment information

Column: 31

Rows: 4,963,508

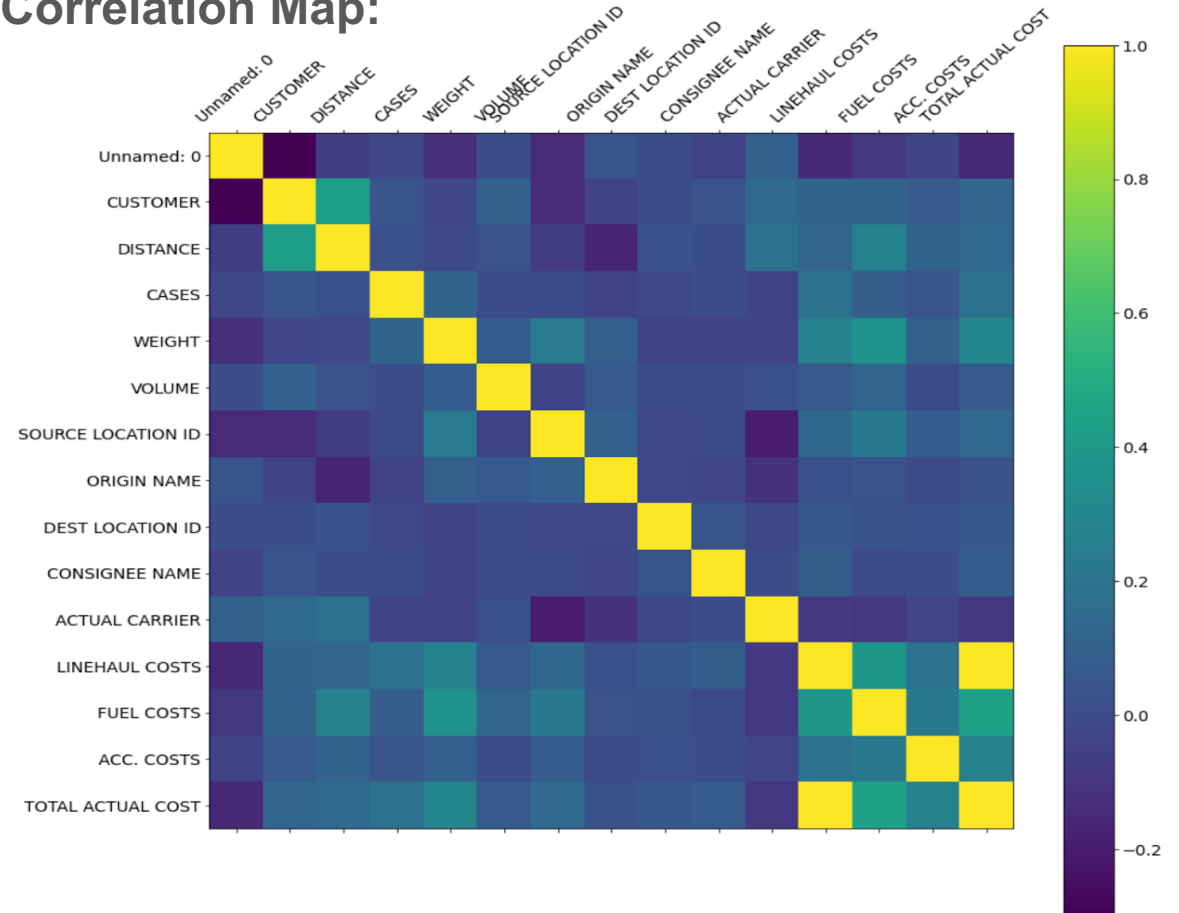
Historical Quoted Price Data (DAT_Data)

Description: The dataset includes 4 month historical quoted price data information

Column: 40

Rows: 63,486

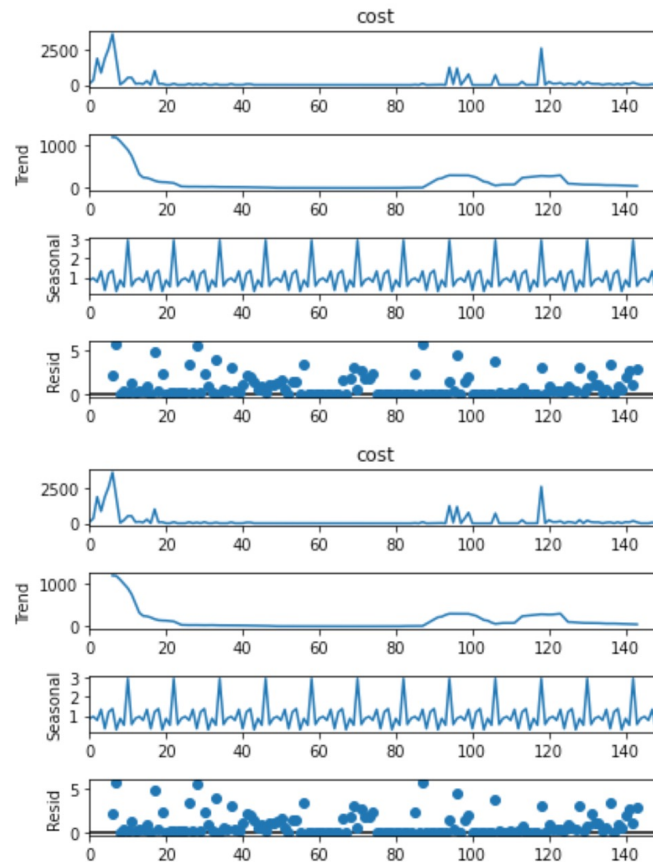
Correlation Map:



Exploratory Data Analysis



Data is provided by DHL Operation and Analytics Team. All data is for study use and has be masked



Augmented Dickey-

Fuller Results

Test Statistic -4.713

P-value 0.000

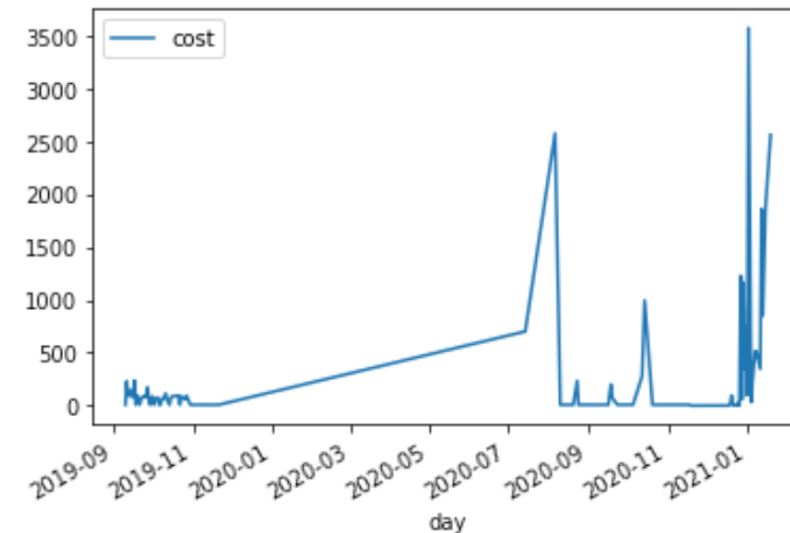
Lags 1

Trend: Constant

Critical Values: -3.48 (1%), -2.88 (5%), -2.58 (10%)

Null Hypothesis: The process contains a unit root.

Alternative Hypothesis: The process is weakly stationary.



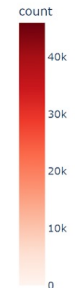
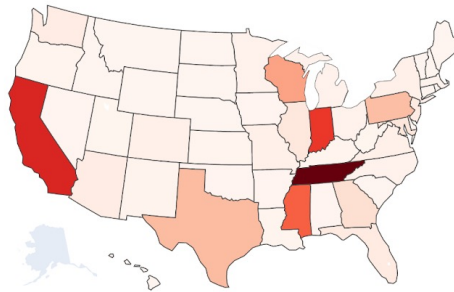
Exploratory Data Analysis



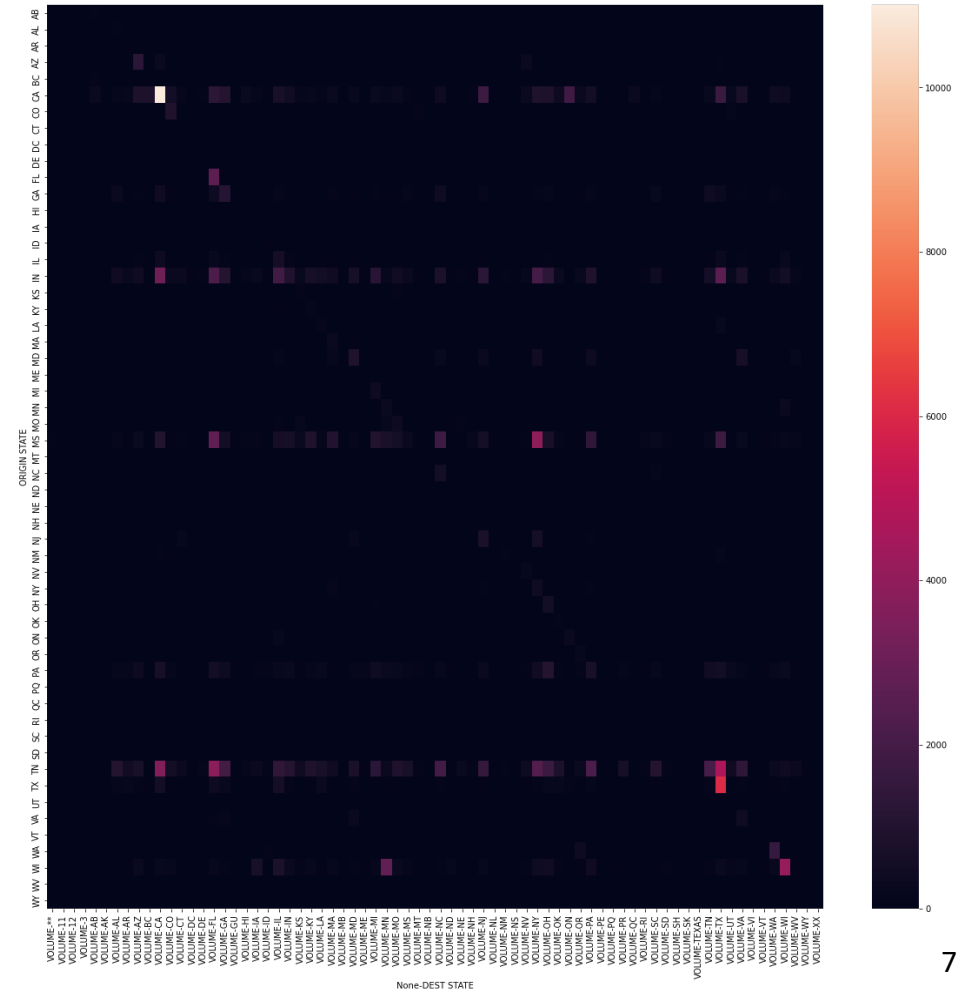
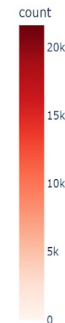
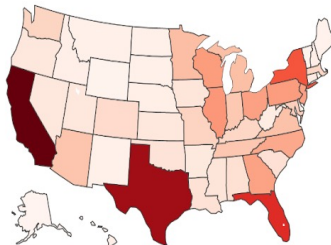
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Shipment Map:

cases count of different origin state



cases count of different destination state



Data Cleaning



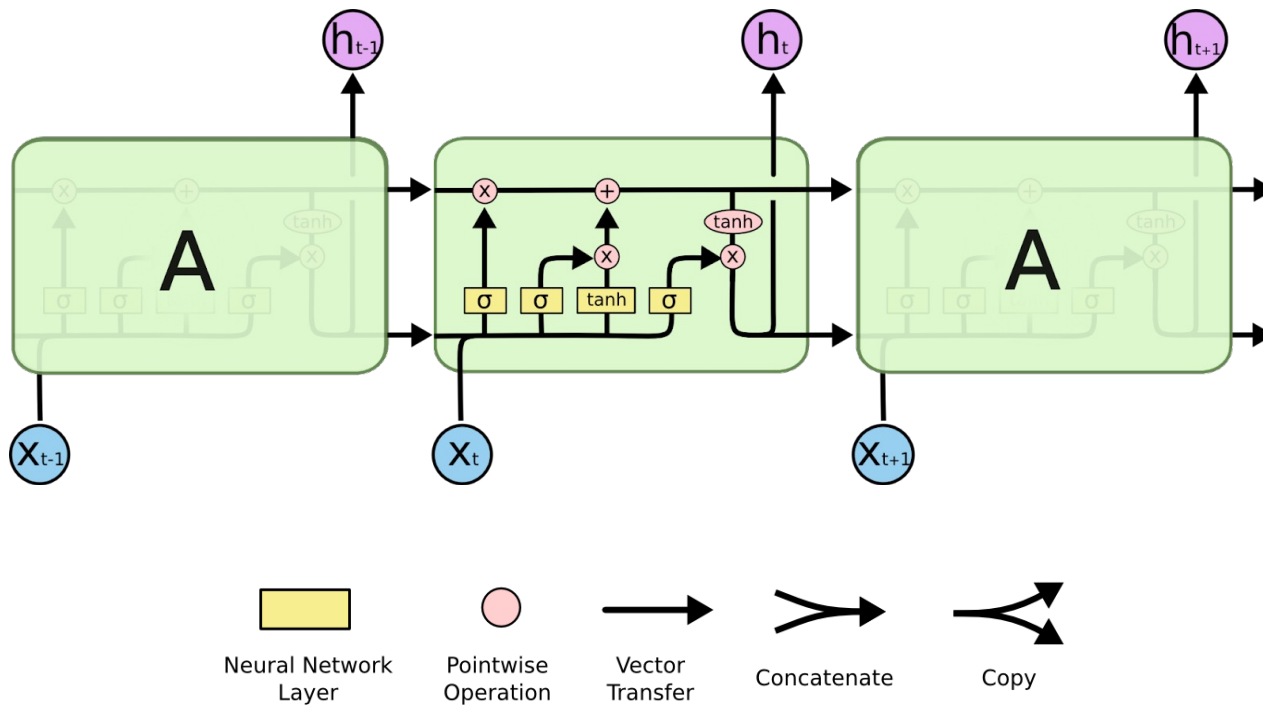
Data Cleaning is a critical working before we implementing model

| STEPS | DESCRIPTION |
|-----------------------------------|--|
| Merge two datasets | Using Zip information and Appointed Shipping time as foreigner key to connect two tables |
| Narrow dataset by location | Collect records only for US and Canada Shipment |
| Splitting Raw Data | The original data is arranged into three for LTL, TL, and Railway datasets |
| Removing NA data | If a record has missing information in LINEHAUL COSTS , FUEL COSTS, ACC. COSTS, TOTAL ACTUAL COST, the record will be removed from model training. |
| Fix Zip related feature | Correctly zip into a correct US or Canada format |
| Correct Data | Compare Only keeps records whose shipment data is not later than quotes time |
| Splitting Raw Data | The original data is arranged into three for specific model training |

Deep Learning Model Design



Based on the dataset, a Long Short-Term Memory Model is design for prediction



Good to carry both long term and short term info by:

- Cell state vector carrying memories
- Gates deciding to forget or not
- Tanh layers to process candidates

LSTM Model Building



We use PyTorch to build a class to train a sample LSTM model

Step 1

Splitting Training data and Test data

Step 2

Building a Neural Network

Step 3

Setting Accuracy Measurement for Loss Calculation

Step 4

Setting Gradient Descent as Model Optimizer

Step 5

Building a Learning Epoch

Step 6

Record Test Error

Hyperparameter and Outcome

| | |
|---------------------|--------------------------|
| Activation Function | ReLU |
| Hidden Layers | 2 |
| Hidden Layers Size | size of training row*2/3 |
| Loss Measurement | SAMPE |
| Learning Epoch | 1000 |

LSTM Model SMAPE Performance



TransRater will compare its SAMPE loss with DataRobot best model to evaluate its performance



TL MODEL

0.3378606

Wait For Test

LTL MODEL

0.25428137

Wait For Test

Railway MODEL

0.33060297

Wait For Test

Next Step



TransRater has an User Interface to help Transportation Solution Teams predict transportation quoted price

T_R

MT Data

DT Data

LTL Prediction TL Prediction Railway Prediction Predict All

DHL

Next Step



We use PowerBI to build a Dashboard for prediction visualization.



Power BI

Building...