Pseudocode for 15/60/11 Sim & Ojas · Import libionies Load CSV files Handle Vallies & values with zero's Define the Enclidean distance function Calculate the distance between picking & dropoff Drop redundant tables (pickip-date time & pickup + dropoff tables) · Peorder columns · Print data franc to see seperated pickup time & distance
· Define & Manually split Train & Test sets
· Normal Te the Data Ene
· Create Function to build model Build model · Split Train & Test Sets · Train Model Extract Training a balidation loss for Epoch Evaluate model on test data

Writing Lode 15 160 Al Sim & Ojas import pandas as pd import humpy as np df=pd. read -csv (bay area - taxi.csv) # loading up the of drapha (inplace = True) # handling the missing df = df[(df['pickup-longitude']!=0)& (df[pickup-labitude']!=0)& (df['dropoff-longitude']!=0)& (df['dropoff-longitude']!=0)& (df['dropoff-labitude']!=0) & (df['dropoff-longitude']!=0) & (df['dropoff-longitude']!=0) & (df['dropoff-longitude']!=0) & (df['pickup-labitude'] # Extracting the components of pickup-date fine

of ['year'] = of ['pickup & date time']. of t. year

of ['month'] = of ['pickup date time']. of t. month

of ['day'] = of ['pickup date time']. of t. day

of ['day of week] - of ['pickup date time']. of t. olay of week

of ['hour'] = of ['pickup date time']. of t. hour

of ['minute'] = of ['pickup date time']. of t. minute

of ['second'] = of ['pickup date time']. of t. second # Defining the Euclidean distance tunctions def euc_distance (lat 1, long 1, lat 2, lat 2):
return ((lat 1-lat 2) 2+ (long 1-long 2) 2) 0.5) # Calculating the distance between pickup and dropoff

off distance = euc_distance (df_pickup_latibude],

df_pickup=longitude],

df_dropoff_latitude],

df_dropoff_latitude]

Dropping pickup-datine & pickup & dropoft since we seperated its components

of = of drop (columns = I pickup-datetime, a

df = of drop (columns = I pickup-latitude, a 0 of = of drop Columns = [pickup longitude of = of - drop (columns = [drop of - lassitude Of = of drop Columns = [dropott - longitude 0 # Reordoning columns & reassighing data frame cols = of columns bolist() cols. remove (distance) distance index = cols index (passanger)+1 Cols insert (distance index, distance) df=df[cols] Print Caf. 100 [: 24,:]) # Defining beinput features & Splitting manually into Brain-train-data = df [[passanger, distance, year, month, day, day of week, hour] train-turgets = df [fare-amount] from sklearh model - Selection imports train - test - split train-data, test-data, train targets, test tanets = train-test - Split (train-data, train-targets, test) ze 0.2, random - State = 42) train_data[: train - trangets [:

mean = brain data (mean (atis = 8) Std = train data. Std (axis=0) frain data = (atmin data - mean) / std test-data = (test-data - mean) / std # Function to build the models det build-mode (C): model = models. Sequential() model add Clayers. Dense (64, action ton = "rela"; input-Shape = Chain-data. Shape [1],))) model add (layers. Dense (64, activation = "rel")) model add Clayers. Dense (I) ["mae"]) le Coptimizer = msprop?, loss = mse', metros = return model (1) 0 4 4

#Building the model model = build-model() # Splitting training data Further into partial training of validation data partial train-data = train _ data [:int (0.8*/ent train_data)]]

partial train targets = train targets [:int (0.8*/ent train_data)]

val _data = train_data [int (0.8*/end train_data)):

val _targets = train_targets [int (0.8*/end train_data)):] #Training the mode num_epochs = 20 history = model . fit partial-train-targets, val-data, val-targets), epochs = num = epochs, batch - size = 1 Verbose =1 # Extracting the trainy of validation loss for epoch loss = history history L loss I val - loss = history history L Yab-loss] epochs = range (I) len (loss) + I) # Evaluating the model on test data

Eest-mse-score, test-mae-score = model-evaluate (test-data, test targets) print (+ "Test MAE: { test_mae-score}")