!pip install pathway bokeh --quiet



```
--- 60.4/60.4 kB 1.2 MB/s eta 0:00:00
        ______ 26.5/26.5 MB 57.9 MB/s eta 0:00:00
45.5/45.5 kB 3.1 MB/s eta 0:00:00
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                    - 244.6/244.6 kB 16.0 MB/s eta 0:00:00
- 985.8/985.8 kB 41.1 MB/s eta 0:00:00
                     - 148.6/148.6 kB 11.2 MB/s eta 0:00:00
                    - 139.8/139.8 kB 9.4 MB/s eta 0:00:00
                     - 65.8/65.8 kB 4.2 MB/s eta 0:00:00
                    -- 55.7/55.7 kB 3.8 MB/s eta 0:00:00
                    - 118.5/118.5 kB 8.1 MB/s eta 0:00:00
                    - 196.2/196.2 kB 11.9 MB/s eta 0:00:00
                   ---- 434.9/434.9 kB 24.6 MB/s eta 0:00:00
                    -- 2.1/2.1 MB 53.2 MB/s eta 0:00:00
                   --- 2.7/2.7 MB 55.7 MB/s eta 0:00:00
                    -- 13.3/13.3 MB 63.6 MB/s eta 0:00:00
                    — 83.2/83.2 kB 1.3 MB/s eta 0:00:00
                     - 2.2/2.2 MB 60.0 MB/s eta 0:00:00
                  ---- 1.6/1.6 MB <mark>52.1 MB/s</mark> eta 0:00:00
```

ERROR: pip's dependency resolver does not currently take into account all the packages t bigframes 2.8.0 requires google-cloud-bigquery[bqstorage,pandas]>=3.31.0, but you have &

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import datetime
from datetime import datetime
import pathway as pw
import bokeh.plotting
import panel as pn



from google.colab import files
uploaded = files.upload()



Choose Files dataset.csv

• dataset.csv(text/csv) - 1595541 bytes, last modified: 7/1/2025 - 100% done Saving dataset.csv to dataset.csv

df=pd.read\_csv('dataset.csv')
df

	ID	SystemCodeNumber	Capacity	Latitude	Longitude	<b>Occupancy</b>	VehicleType
0	0	BHMBCCMKT01	577	26.144536	91.736172	61	car
1	1	BHMBCCMKT01	577	26.144536	91.736172	64	car
2	2	BHMBCCMKT01	577	26.144536	91.736172	80	car
3	3	BHMBCCMKT01	577	26.144536	91.736172	107	car
4	4	BHMBCCMKT01	577	26.144536	91.736172	150	bike
18363	18363	Shopping	1920	26.150504	91.733531	1517	truck
18364	18364	Shopping	1920	26.150504	91.733531	1487	car
18365	18365	Shopping	1920	26.150504	91.733531	1432	cycle
18366	18366	Shopping	1920	26.150504	91.733531	1321	car
18367	18367	Shopping	1920	26.150504	91.733531	1180	car
	1 2 3 4  18363 18364 18365 18366	0 0 1 1 2 2 3 3 4 4 18363 18363 18364 18364 18365 18365 18366 18366	0         0         BHMBCCMKT01           1         1         BHMBCCMKT01           2         2         BHMBCCMKT01           3         3         BHMBCCMKT01           4         4         BHMBCCMKT01                18363         18363         Shopping           18364         18364         Shopping           18365         18365         Shopping           18366         18366         Shopping	0       0       BHMBCCMKT01       577         1       1       BHMBCCMKT01       577         2       2       BHMBCCMKT01       577         3       3       BHMBCCMKT01       577         4       4       BHMBCCMKT01       577               18363       18363       Shopping       1920         18364       18364       Shopping       1920         18365       18365       Shopping       1920         18366       18366       Shopping       1920	0       0       BHMBCCMKT01       577       26.144536         1       1       BHMBCCMKT01       577       26.144536         2       2       BHMBCCMKT01       577       26.144536         3       3       BHMBCCMKT01       577       26.144536         4       4       BHMBCCMKT01       577       26.144536                 18363       18363       Shopping       1920       26.150504         18364       18364       Shopping       1920       26.150504         18365       18365       Shopping       1920       26.150504         18366       18366       Shopping       1920       26.150504	0         0         BHMBCCMKT01         577         26.144536         91.736172           1         1         BHMBCCMKT01         577         26.144536         91.736172           2         2         BHMBCCMKT01         577         26.144536         91.736172           3         3         BHMBCCMKT01         577         26.144536         91.736172           4         4         BHMBCCMKT01         577         26.144536         91.736172                   18363         18363         Shopping         1920         26.150504         91.733531           18364         18365         Shopping         1920         26.150504         91.733531           18366         18366         Shopping         1920         26.150504         91.733531	0         0         BHMBCCMKT01         577         26.144536         91.736172         61           1         1         BHMBCCMKT01         577         26.144536         91.736172         64           2         2         BHMBCCMKT01         577         26.144536         91.736172         80           3         3         BHMBCCMKT01         577         26.144536         91.736172         107           4         4         BHMBCCMKT01         577         26.144536         91.736172         150                    18363         18363         Shopping         1920         26.150504         91.733531         1517           18364         18364         Shopping         1920         26.150504         91.733531         1487           18365         18365         Shopping         1920         26.150504         91.733531         1432           18366         18366         Shopping         1920         26.150504         91.733531         1321

18368 rows x 12 columns

Next steps: ( Go

Generate code with df

▼ View recommended plots

New interactive sheet

```
array(['BHMBCCMKT01', 'BHMBCCTHL01', 'BHMEURBRD01', 'BHMMBMMBX01', 'BHMNCPHST01', 'BHMNCPNST01', 'Broad Street', 'Others-CCCPS105a', 'Others-CCCPS119a', 'Others-CCCPS135a', 'Others-CCCPS202', 'Others-CCCPS8', 'Others-CCCPS98', 'Shopping'], dtype=object)
```

```
df["TrafficCondition_Code"] = df["TrafficConditionNearby"].astype("category").cat.codes # ]
df["vehicle_Code"] = df["VehicleType"].astype("category").cat.codes #car:1,bike:0,truck:3:c
df
```

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18368 rows × 14 columns

Next steps: Generate code with df View recommended plots New interactive sheet plots=[] for lot in lots: import datetime df['Timestamp'] = pd.to\_datetime(df['LastUpdatedDate'] + ' ' + df['LastUpdatedTime'], #c format='%d-%m-%Y %H:%M:%S') df\_lot=df[df['SystemCodeNumber']==lot] #making dataframe for one lot at a time df\_lot = df\_lot.sort\_values('Timestamp').reset\_index(drop=True) #sorting time class ParkingSchema(pw.Schema): Capacity:int Occupancy:int Timestamp:str SystemCodeNumber:str VehicleType:str TrafficConditionNearby:str QueueLength:int IsSpecialDay:int TrafficCondition\_Code:int vehicle\_Code:int filename = f"parking\_stream\_{lot}.csv" df\_lot[["Timestamp", "Occupancy", "Capacity", "SystemCodeNumber","VehicleType","TrafficCor data = pw.demo.replay\_csv(filename, schema=ParkingSchema, input\_rate=1000) # Load the da fmt = "%Y-%m-%d %H:%M:%S"

data\_with\_time = data.with\_columns(

```
t = data.Timestamp.dt.strptime(fmt), #containes full datetime
    day = data.Timestamp.dt.strptime(fmt).dt.strftime("%Y-%m-%dT00:00:00"), #contains onl
    hour = data.Timestamp.dt.strptime(fmt).dt.hour(),
                                                       #contains hour
    day of week = data.Timestamp.dt.strptime(fmt).dt.weekday(),
                                                                  #assigns monday: 0,...,s
    occupancy_rate = data.Occupancy / data.Capacity
)
def time_of_day_weight(hour): #the reason for choosing these categories is explained in
  if 11 <= hour < 14:
      return 1.0
                    # Midday
  elif 14 <= hour < 17:
      return 0.7 # Evening
  else:
      return 0.4 # Morning
def weekday_weight(day_of_week):
  if day_of_week < 5:
      return 1.0 # Weekday
  else:
      return 0.7 # Weekend
def queue_time_of_day_weight(hour):
  if 12<= hour < 14:
      return 1.0
  else:
      return 0.7
def traffic_time_of_day_weight(hour):
  if 12<= hour < 14:
      return 1.0
  else:
      return 0.5
def hour effect(hour):
 a = 1.0
 mu = 12
  sigma = 3
  return a * np.exp(-((hour - mu) ** 2) / (2 * sigma ** 2))
beta=1.0
gamma=1.0
delta=1.0
epsilon=1.0
zeta=1.0
vehicle_weights = {'car': 1.0, 'bike': 0.8, 'truck': 1.3, 'cycle': 0.7}
def demand_fn(occ_rate,tod_weight,wd_weight,adj_queue,adj_traffic,hour_term,veh_term,spec_
  return (
     2.0 * occ_rate * tod_weight * wd_weight+beta * np.log1p(adj_queue)+gamma * np.log1p(adj_queue)
import datetime
```

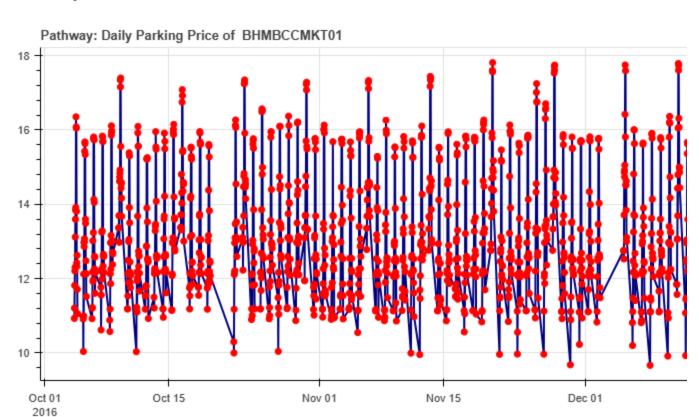
```
data with price=(
    data_with_time.with_columns(
        tod weight = pw.apply(time of day weight, data with time.hour),
        wd_weight = pw.apply(weekday_weight, data_with_time.day_of_week),
        queue_weight = pw.apply(queue_time_of_day_weight,data_with_time.hour),
        traffic_weight = pw.apply(traffic_time_of_day_weight,data_with_time.hour),
    )
    .with columns(
        adj_queue = pw.apply(lambda q, w: q * w, data_with_time.QueueLength, pw.this.queue_
        adj_traffic = pw.apply(lambda t, w: t * w, data_with_time.TrafficCondition_Code, pv
        hour term = pw.apply(hour effect,data with time.hour),
        veh_term = vehicle_weights.get(data_with_time.VehicleType, 1.0),
        spec_term = data_with_time.IsSpecialDay
    )
    .with_columns(
     demand=pw.apply(demand_fn,
                     data_with_time.occupancy_rate,
                     pw.this.tod weight,
                      pw.this.wd_weight,
                     pw.this.adj_queue,
                     pw.this.adj_traffic,
                     pw.this.hour_term,
                     pw.this.veh term,
                     pw.this.spec_term
     )
 LAMBDA=1.0
def normalize(d):
   return (d - 2) / (8 - 2)
def price_fn(demand_norm, LAMBDA=LAMBDA):
   return 10 * (1 + LAMBDA * demand_norm)
# aligned data = data with price.demand.with universe of(data with price)
 plot table=(data with price.with columns(
    # with_price=data_with_price
     demand_norm = pw.apply(normalize,data_with_price.demand)
 )
 .with columns(
     price = pw.apply(price_fn, pw.this.demand_norm)
 )
def price_plotter(source): #writing bokeh plot fn
```

```
# Create a Bokeh figure with datetime x-axis
   fig = bokeh.plotting.figure(
       height=400,
       width=800,
       title=f"Pathway: Daily Parking Price of {lot}",
       x_axis_type="datetime",
   )
   # Plot a line graph showing how the price evolves over time
   fig.line("t", "price", source=source, line_width=2, color="navy")
   # Overlay red circles at each data point for better visibility
   fig.scatter("t", "price", source=source, size=6, color="red")
   return fig
 viz = plot_table.plot(price_plotter, sorting_col="t")
 plots.append(pn.Column(f"Lot: {lot}", viz.servable()))
dashboard = pn.Column(*plots)
dashboard.servable()
```



Lot: BHMBCCMKT01

### Streaming mode



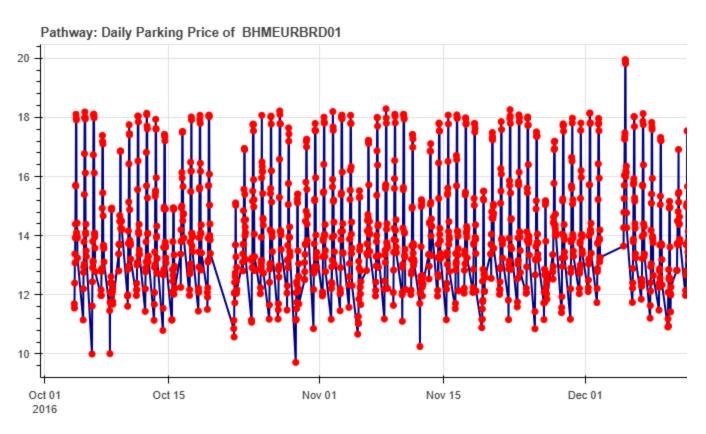
Lot: BHMBCCTHL01



Oct 01 Oct 15 Nov 01 Nov 15 Dec 01 2016

Lot: BHMEURBRD01

### Streaming mode



Lot: BHMMBMMBX01

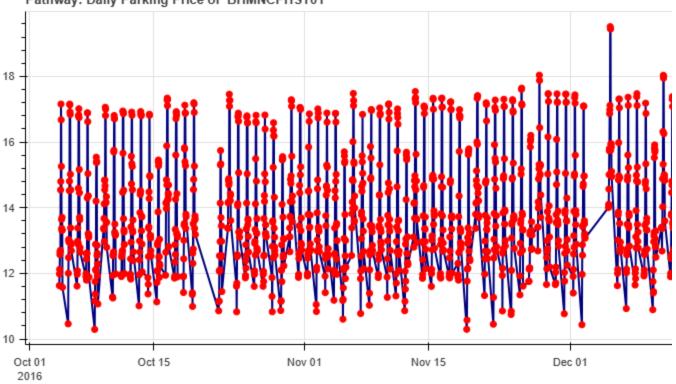




Lot: BHMNCPHST01

### Streaming mode

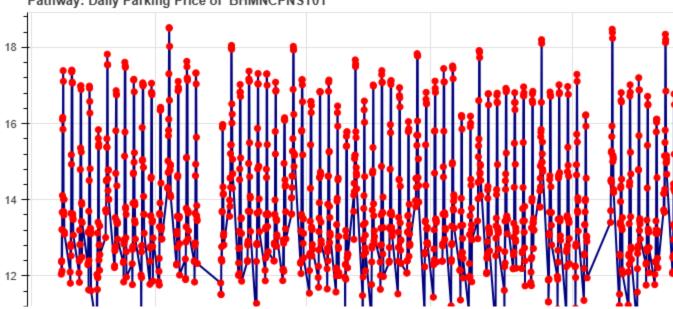




Lot: BHMNCPNST01

# Streaming mode

# Pathway: Daily Parking Price of BHMNCPNST01

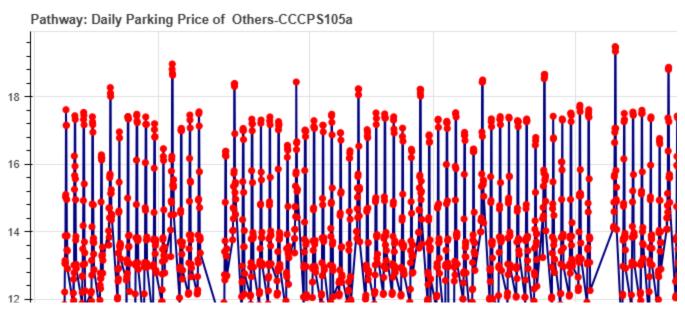


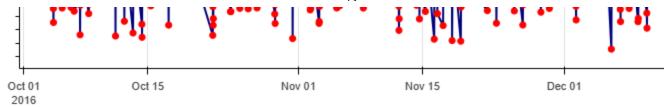
Lot: Broad Street

Streaming mode



Lot: Others-CCCPS105a

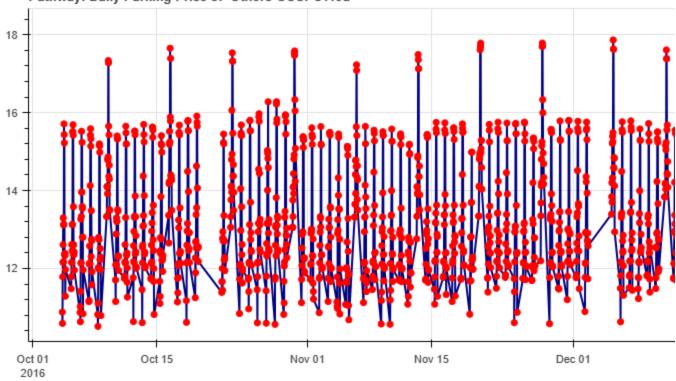




Lot: Others-CCCPS119a

# Streaming mode

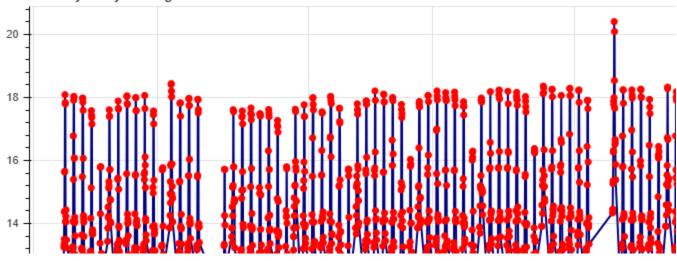


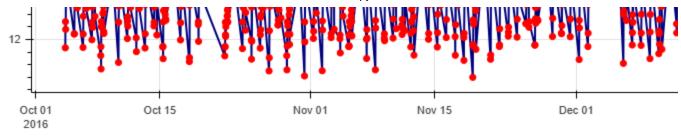


Lot: Others-CCCPS135a

# Streaming mode

# Pathway: Daily Parking Price of Others-CCCP\$135a

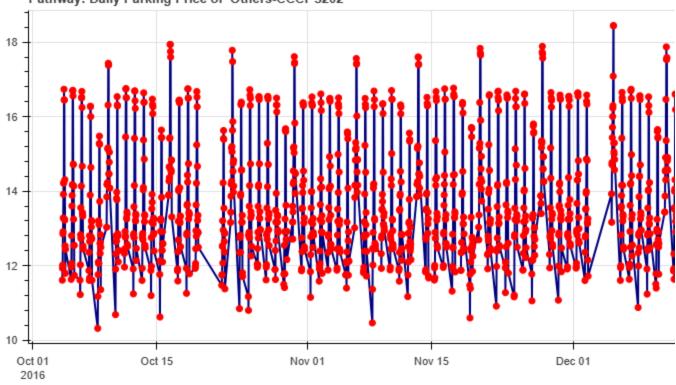




Lot: Others-CCCPS202

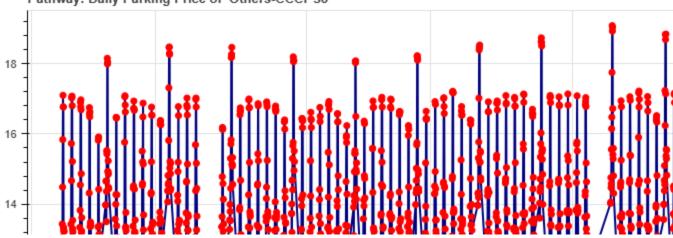
### Streaming mode

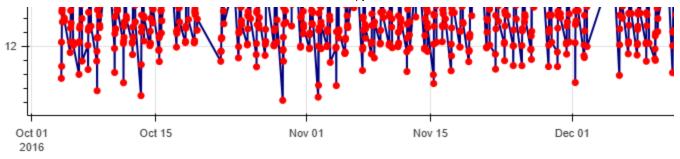




Lot: Others-CCCPS8

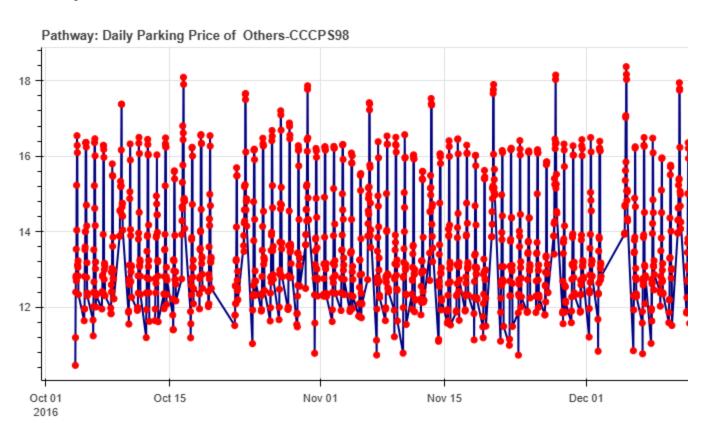






Lot: Others-CCCPS98

### Streaming mode



Lot: Shopping



