

Extension Project will address Bluebikes correlation with Bus performance.

With this, we will be using Bluebikes station and trip datasets along with previous data about bus given in Bus performance project document.

Subproblem 1: Worst on-time Performance Routes correlation with Bluebikes usage

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from collections import defaultdict
import os

folder_path = "../../data/"
blue_bikes_trips = pd.read_csv(f"{folder_path}/202201-bluebikes-
tripdata.csv")
blue_bikes_stations =
pd.read_csv(f"{folder_path}/current_bluebikes_stations.csv")
census_neighbourhood = pd.read_csv(f"{folder_path}/Census-Boston-
Neighborhood.csv")
mbta_gtfs = pd.read_csv(f"{folder_path}/MBTA_Systemwide_GTFS_Map.csv")
mbta_reliability =
pd.read_csv(f"{folder_path}/MBTA_Bus_Reliability.csv")
mbta_prediction_accuracy =
pd.read_csv(f"{folder_path}/Bus_Prediction_Accuracy.csv")
mbta_bus_ridership =
pd.read_csv(f"{folder_path}/MBTA_Bus_Ridership.csv")
```

```
C:\Users\sataa\AppData\Local\Temp\ipykernel_5364\437775067.py:8:
DtypeWarning: Columns (2,3) have mixed types. Specify dtype option on
import or set low_memory=False.
```

```
    mbta_bus_ridership =
pd.read_csv(f"{folder_path}/MBTA_Bus_Ridership.csv")
```

```
def process_blue_bikes_trips(blue_bikes_trips):
    processed_blue_bikes_trips = blue_bikes_trips
    processed_blue_bikes_trips["tripduration"] =
blue_bikes_trips["tripduration"] / 60
    return processed_blue_bikes_trips
```

```
processed_blue_bikes_trips =
process_blue_bikes_trips(blue_bikes_trips)
processed_blue_bikes_trips.head()
```

	tripduration	starttime	stoptime	\
0	9.950000	2022-01-01 00:00:25.1660	2022-01-01 00:10:22.1920	
1	6.850000	2022-01-01 00:00:40.4300	2022-01-01 00:07:32.1980	
2	7.933333	2022-01-01 00:00:54.8180	2022-01-01 00:08:51.6680	
3	7.766667	2022-01-01 00:01:01.6080	2022-01-01 00:08:48.2350	

```
4      12.533333  2022-01-01 00:01:06.0520  2022-01-01 00:13:38.2300
```

```
      start station id      start station name  start station
latitude \
```

```
0      178  MIT Pacific St at Purrington St
```

```
42.359573
```

```
1      189      Kendall T
```

```
42.362428
```

```
2      94      Main St at Austin St
```

```
42.375603
```

```
3      94      Main St at Austin St
```

```
42.375603
```

```
4      19      Park Dr at Buswell St
```

```
42.347241
```

```
      start station longitude  end station id \
```

```
0      -71.101295      74
```

```
1      -71.084955      178
```

```
2      -71.064608      356
```

```
3      -71.064608      356
```

```
4      -71.105301      41
```

```
      end station name  end station
latitude \
```

```
0      Harvard Square at Mass Ave/ Dunster
```

```
42.373268
```

```
1      MIT Pacific St at Purrington St
```

```
42.359573
```

```
2      Charlestown Navy Yard
```

```
42.374125
```

```
3      Charlestown Navy Yard
```

```
42.374125
```

```
4  Packard's Corner - Commonwealth Ave at Brighto...
```

```
42.352261
```

```
      end station longitude  bikeid  usertype postal code
```

```
0      -71.118579      4923  Subscriber      02139
```

```
1      -71.101295      3112  Subscriber      02139
```

```
2      -71.054812      6901   Customer      02124
```

```
3      -71.054812      5214   Customer      02124
```

```
4      -71.123831      2214  Subscriber      02215
```

```
# print(blue_bikes_stations.head())
```

```
def process_blue_bikes_stations(blue_bikes_stations,
blue_bikes_trips):
```

```
    # Fixing first row as the column names
```

```
    new_column_names = blue_bikes_stations.iloc[0] # Get the first
row to use as column names
```

```
    blue_bikes_stations.columns = new_column_names # Set new column
```

```

names
    blue_bikes_stations =
blue_bikes_stations.iloc[1:].reset_index(drop=True)

    # Extracting the unique start station names and IDs
    start_stations = blue_bikes_trips[['start station id', 'start
station name']].drop_duplicates()
    start_stations = start_stations.rename(columns={'start station
id': 'station_id', 'start station name': 'station_name'})

    # Extracting the unique end station names and IDs.
    end_stations = blue_bikes_trips[['end station id', 'end station
name']].drop_duplicates()
    end_stations = end_stations.rename(columns={'end station id':
'station_id', 'end station name': 'station_name'})

    # Combining the start and end station information.
    combined_stations = pd.concat([start_stations,
end_stations]).drop_duplicates().set_index('station_name')
    blue_bikes_stations['station_id'] =
blue_bikes_stations['Name'].map(combined_stations['station_id'])
    blue_bikes_stations = blue_bikes_stations.dropna(subset =
["station_id"])
    return blue_bikes_stations

processed_blue_bikes_stations =
process_blue_bikes_stations(blue_bikes_stations,
processed_blue_bikes_trips)
processed_blue_bikes_stations.head()

```

0	Number	Name	Latitude
0	K32015	1200 Beacon St	42.34414899 -
71.11467361			
1	W32006	160 Arsenal	42.36466403 -
71.17569387			
2	A32019	175 N Harvard St	42.36447457 -
71.12840831			
3	S32035	191 Beacon St	42.38032335 -
71.10878613			
4	C32094	2 Hummingbird Lane at Olmsted Green	42.28887 -
71.095003			

0	District	Public	Total docks	Deployment	Year	station_id
0	Brookline	Yes	1	2021	452.0	
1	Watertown	Yes	11	2021	502.0	
2	Boston	Yes	17	2014	149.0	
3	Somerville	Yes	19	2018	378.0	
4	Boston	Yes	17	2020	493.0	

```
census_neighbourhood.head()
```

```
      tract20_nbhd P0020001      P0020005
P0020006 \
0 field concept      Total: White alone Black or African American
alone
1      Allston      24904      12536
1326
2      Back Bay      18190      13065
690
3      Beacon Hill      9336      7521
252
4      Brighton      52047      32694
2414
```

```
      P0020002
P002aapi \
0 Hispanic or Latino Asian, Native Hawaiian and Pacific Islander
al...
1      3259
6271
2      1208
2410
3      537
630
4      5376
8703
```

```
      P002others P0040001      P0040005 \
0 Other Races or Multiple Races, all ages      Total: White alone
1      1512      23140      11976
2      817      17042      12349
3      396      8603      6980
4      2860      47657      30752
```

```
      P0040006 ... \
0 Black or African American alone ...
1      1184 ...
2      641 ...
3      231 ...
4      2076 ...
```

```
      P0050005 \
0 Nursing facilities/Skilled-nursing facilities
1      0
2      269
3      0
4      266
```

```
P0050006
```

```
P0050007 \
```

0	Other institutional facilities	Noninstitutionalized population:
1	0	3281
2	0	1610
3	0	33
4	56	3796

	P0050008	P0050009	\
0	College/University student housing	Military quarters	
1	3214	0	
2	1487	0	
3	0	0	
4	3493	0	

	P0050010	H0010001	H0010002	H0010003	\
0	Other noninstitutional facilities	Total:	Occupied	Vacant	
1	67	10748	10027	721	
2	123	11524	10006	1518	
3	33	6037	5485	552	
4	303	23653	22535	1118	

	hhsiz
0	household size
1	2.156477511
2	1.630121927
3	1.696080219
4	2.126292434

[5 rows x 34 columns]

mbta_gtfs.head()

```
def process_gtfs(MBTA_data):
    MBTA_data = MBTA_data[MBTA_data['Neighborhood'].notnull()]
    MBTA_data = MBTA_data[MBTA_data['Routes'] != '#N/A']
    MBTA_data = MBTA_data[MBTA_data['Routes'].notnull()]

    # Split routes column to separate routes
    MBTA_data['Routes'] = MBTA_data['Routes'].str.split('|')
    MBTA_data = MBTA_data.explode('Routes')
    df = MBTA_data[["stop_id", "stop_name", "stop_lat", "stop_lon",
"Neighborhood", "Routes"]]

    return df
```

```
processed_mbta_gtfs = process_gtfs(mbta_gtfs)
processed_mbta_gtfs.head()
```

stop_id	stop_name	stop_lat	stop_lon
Neighborhood	\		
0	1 Washington St opp Ruggles St	42.330957	-71.082754

```
Roxbury
0      1 Washington St opp Ruggles St 42.330957 -71.082754
Roxbury
0      1 Washington St opp Ruggles St 42.330957 -71.082754
Roxbury
0      1 Washington St opp Ruggles St 42.330957 -71.082754
Roxbury
0      1 Washington St opp Ruggles St 42.330957 -71.082754
Roxbury
```

```
Routes
0      1
0      8
0     10
0     47
0     19
```

```
mbta_prediction_accuracy.head()
```

			weekly mode	route_id	bin	
arrival	departure	\				
0	2021/08/13 04:00:00+00	bus	NaN	0-3 min	departure	
1	2021/08/13 04:00:00+00	bus	NaN	3-6 min	departure	
2	2021/08/13 04:00:00+00	bus	NaN	6-12 min	departure	
3	2021/08/13 04:00:00+00	bus	NaN	12-30 min	departure	
4	2021/08/20 04:00:00+00	bus	NaN	0-3 min	departure	

	num_predictions	num_accurate_predictions	ObjectId
0	293039	233562	1
1	285817	229090	2
2	561098	472923	3
3	1594830	1405620	4
4	285591	228653	5

```
mbta_reliability.head()
```

```
# Code taken from Base Question 2 code
```

```
def process_reliability(df):
    new_df = df[df["mode_type"]=="Bus"] # taking only buses
    new_df = new_df.dropna(subset=['otp_denominator',
'otp_numerator','cancelled_numerator']) # No NaN / Null
    new_df['ot_rate'] =
new_df['otp_numerator']/new_df['otp_denominator']
    grouped_route = new_df.groupby('gtfs_route_id')
    grouped_rate = grouped_route['ot_rate'].mean().reset_index()
    rate_sorted = grouped_rate.sort_values(by='ot_rate',
```

```

ascending=False)
    return rate_sorted

reliability_rate_sorted = process_reliability(mbta_reliability)

reliability_rate_sorted.head() # best ot_rate
reliability_rate_sorted.tail() # worst ot_rate

```

	gtfs_route_id	ot_rate
150	747	0.458202
106	459	0.429970
99	448	0.406302
100	449	0.402552
178	9703	0.320094

We have the best and worst on-time performance data extracted from base question 2 - Utilizes the MBTA Reliability Dataset:

Best 10:



	gtfs_route_id	ot_rate
182	CR-Shuttle003	0.925859
181	CR-Shuttle002	0.858203
180	CR-Shuttle001	0.858203
147	742	0.837185
144	73	0.820220
112	502	0.813195
65	32	0.807782
151	749	0.807251
11	111	0.803600
153	751	0.801902

Worst 10:



	gtfs_route_id	ot_rate
24	14	0.509825
140	70A	0.494182
31	19	0.493452
36	195	0.491992
82	41	0.488934
150	747	0.458202
106	459	0.429970
99	448	0.406302
100	449	0.402552
178	9703	0.320094

```
merged_data_on_routes = pd.merge(processed_mbtg_gtfs,
reliability_rate_sorted, left_on = "Routes", right_on =
"gtfs_route_id")

print(merged_data_on_routes['gtfs_route_id'].isna().sum()) # checking
no bus routes are not included in the reliability dataset.
print(merged_data_on_routes['Routes'].isna().sum()) # checking no bus
routes are not included in the GTFS dataset.

merged_data_on_routes.head()

0
0
```

	stop_id	stop_name	stop_lat	stop_lon	\
0	1	Washington St opp Ruggles St	42.330957	-71.082754	
1	10003	Albany St opp Randall St	42.331591	-71.076237	
2	10100	Albany St @ Randall St	42.331675	-71.076347	
3	10101	Melnea Cass Blvd @ Harrison Ave	42.332066	-71.079147	
4	10590	Massachusetts Ave @ Washington St	42.336621	-71.076956	

	Neighborhood	Routes	gtfs_route_id	ot_rate
0	Roxbury	1	1	0.744301
1	Roxbury	1	1	0.744301
2	Roxbury	1	1	0.744301
3	Roxbury	1	1	0.744301
4	South End	1	1	0.744301

```
# Group by 'Routes'
```

```
grouped_by_routes = merged_data_on_routes.groupby('Routes')
```

```
grouped_by_routes.head()
```

```
# # Aggregate 'ot_rate' for each route, then sort to find the worst 10
# # Assuming 'worst' means the highest values
```

```
worst_routes =
```

```
grouped_by_routes['ot_rate'].mean().sort_values(ascending=True).head(10)
```

```
# # Print the worst 10 routes based on ot_rate
```

```
print(worst_routes.head())
```

```
Routes
```

```
9703    0.320094
```

```
449     0.402552
```

```
448     0.406302
```

```
459     0.429970
```

```
747     0.458202
```

```
Name: ot_rate, dtype: float64
```

```
worst_routes_loc = pd.merge(worst_routes, merged_data_on_routes,
left_on = ["Routes", "ot_rate"], right_on = ["Routes", "ot_rate"])
```

```
worst_routes_loc.rename(columns={"Routes": "route"}, inplace=True)
```

```
worst_routes_loc.head(25) # rows are per stop, so showing more rows
ensures the visibility of other routes here beyond route 9703
```

```
# print(worst_routes_loc.shape)
```

	route	ot_rate	stop_id	stop_name	\
0	9703	0.320094	1111	Cambridge St opp Hano	
1	9703	0.320094	1112	Cambridge St @ Harvard	
2	9703	0.320094	1113	Cambridge St @ Linden	

3	9703	0.320094	1114	Cambridge St @ N Harvard
St				
4	9703	0.320094	11388	Huntington Ave @ Belvidere
St				
5	9703	0.320094	1257	Tremont St @ Prentiss
St				
6	9703	0.320094	1258	Tremont St @ Roxbury Crossing
Station				
7	9703	0.320094	1260	Columbus Ave @ New Cedar
St				
8	9703	0.320094	1262	Columbus Ave @ Heath
St				
9	9703	0.320094	1784	Ruggles St @ Huntington
Ave				
10	9703	0.320094	1785	Ruggles St @ Annunciation
Rd				
11	9703	0.320094	31391	Huntington Ave @ Gainsborough
St				
12	9703	0.320094	41391	Huntington Ave @ Opera
Pl				
13	9703	0.320094	61391	Huntington Ave @ Forsyth
Way				
14	9703	0.320094	71391	Huntington Ave @ Louis Prang
St				
15	9703	0.320094	922	Cambridge St opp Dustin
St				
16	9703	0.320094	924	Cambridge St @ Gordon
St				
17	9703	0.320094	925	Cambridge St @ Barrows
St				
18	448	0.406302	16535	Otis St @ Summer
St				
19	448	0.406302	4727	McClellan Highway @ Addison
St				
20	448	0.406302	4728	McClellan Highway @ Boardman
St				
21	448	0.406302	6535	Franklin St @ Devonshire
St				
22	448	0.406302	6564	Summer St @ South Station - Red Line
entrance				
23	448	0.406302	7094	Terminal C - Departures
Level				
24	448	0.406302	892	Summer St @ Atlantic
Ave				
	stop_lat	stop_lon	Neighborhood	gtfs_route_id
0	42.353931	-71.136365	Allston	9703
1	42.355641	-71.132361	Allston	9703
2	42.355943	-71.131448	Allston	9703

3	42.357758	-71.126505	Allston	9703
4	42.345344	-71.082045	Back Bay	9703
5	42.332930	-71.092638	Roxbury	9703
6	42.331311	-71.094831	Roxbury	9703
7	42.328067	-71.097310	Roxbury	9703
8	42.325028	-71.098483	Roxbury	9703
9	42.337416	-71.095079	Mission Hill	9703
10	42.336729	-71.093223	Mission Hill	9703
11	42.341443	-71.086788	Fenway	9703
12	42.340553	-71.088908	Fenway	9703
13	42.339219	-71.092168	Fenway	9703
14	42.337684	-71.096046	Fenway	9703
15	42.350692	-71.145688	Allston	9703
16	42.352276	-71.140761	Allston	9703
17	42.353091	-71.138430	Allston	9703
18	42.354243	-71.058557	Downtown	448
19	42.386142	-71.019171	East Boston	448
20	42.391562	-71.012888	East Boston	448
21	42.355521	-71.057253	Downtown	448
22	42.352253	-71.054774	Downtown	448
23	42.366635	-71.017167	East Boston	448
24	42.352480	-71.054849	Downtown	448

From here, we will be comparing locations of bus stations of the worst routes and the locations of bluebikes going along those routes. We will then see the average number of rides in that station.

This formula is used to take distances between locations (using longitude and latitude)

```
def haversine(lon1, lat1, lon2, lat2):
    R = 6371 # Earth radius in km
    dlon = np.radians(lon2 - lon1)
    dlat = np.radians(lat2 - lat1)
    a = np.sin(dlat/2)**2 + np.cos(np.radians(lat1)) *
np.cos(np.radians(lat2)) * np.sin(dlon/2)**2
    c = 2 * np.arctan2(np.sqrt(a), np.sqrt(1-a))
    distance = R * c
    return distance
```

This formula will be used to test if bluebikes stations are within a 10 minute walk away from any of the worst route stops.

```
def no_more_than_x_mins(distance, x):
    max_walking_distance = x / 60 * 5 # assuming a walking speed of 5
km/h.
    return distance <= max_walking_distance
```

```

MAX_WALKING_DISTANCE = 10 # in minutes

close_blue_bikes_list = defaultdict(list)
# Comparing the locations:
for _, bus_stop in worst_routes_loc.iterrows():
    # Extract latitude and longitude for the bus stop
    bus_stop_lat, bus_stop_lon = float(bus_stop['stop_lat']),
float(bus_stop['stop_lon'])

    # Iterate through each blue bike station
    for _, bike_station in processed_blue_bikes_stations.iterrows():
        # Extract latitude and longitude for the bike station
        bike_station_lat, bike_station_lon =
float(bike_station['Latitude']), float(bike_station['Longitude'])

        # Calculate the distance between the bus stop and the bike
station
        distance = haversine(bus_stop_lon, bus_stop_lat,
bike_station_lon, bike_station_lat)

        if no_more_than_x_mins(distance, MAX_WALKING_DISTANCE):
            if (bike_station['station_id'] not in
close_blue_bikes_list[bus_stop["route"]]): # taking only the distinct
stops

close_blue_bikes_list[bus_stop["route"]].append(bike_station["station_
id"])

# Reverse the dictionary
bike_to_bus_station = {bike_station: bus_station for bus_station,
bike_stations in close_blue_bikes_list.items() for bike_station in
bike_stations}

# Filter the DataFrame
filtered_blue_bikes_stations =
processed_blue_bikes_stations[processed_blue_bikes_stations['station_i
d'].isin(bike_to_bus_station.keys())]

# Add the new column for bus_station_id
filtered_blue_bikes_stations['route_id'] =
filtered_blue_bikes_stations['station_id'].map(bike_to_bus_station)

```

C:\Users\sataa\AppData\Local\Temp\ipykernel_5364\2799738196.py:9:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation:
https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
filtered_blue_bikes_stations['route_id'] =
filtered_blue_bikes_stations['station_id'].map(bike_to_bus_station)

filtered_station_ids = set(filtered_blue_bikes_stations['station_id'])
```

```
# Filter the processed_blue_bikes_trips DataFrame
# Keep rows where either the start or end station is in the list of
filtered station IDs
```

```
filtered_trips = processed_blue_bikes_trips[
    processed_blue_bikes_trips['start station
id'].isin(filtered_station_ids) |
    processed_blue_bikes_trips['end station
id'].isin(filtered_station_ids)
]
```

```
print(filtered_trips.head())
```

	tripduration	starttime	stoptime \
4	12.533333	2022-01-01 00:01:06.0520	2022-01-01 00:13:38.2300
6	16.383333	2022-01-01 00:01:24.7490	2022-01-01 00:17:48.1850
8	13.950000	2022-01-01 00:02:30.2650	2022-01-01 00:16:27.6270
10	13.350000	2022-01-01 00:03:23.9650	2022-01-01 00:16:45.4150
11	12.483333	2022-01-01 00:03:37.2630	2022-01-01 00:16:07.1390

	start station id	start station name \
4	19	Park Dr at Buswell St
6	36	Copley Square - Dartmouth St at Boylston St
8	60	Charles Circle - Charles St at Cambridge St
10	60	Charles Circle - Charles St at Cambridge St
11	4	Tremont St at E Berkeley St

	start station latitude	start station longitude	end station id \
4	42.347241	-71.105301	41
6	42.349928	-71.077392	36
8	42.360793	-71.071190	363
10	42.360793	-71.071190	363
11	42.345392	-71.069616	4

	end station name	end station latitude \
4	Packard's Corner - Commonwealth Ave at Brighto...	42.352261

6	Copley Square - Dartmouth St at Boylston St
42.349928	
8	Harrison Ave at Mullins Way
42.345216	
10	Harrison Ave at Mullins Way
42.345216	
11	Tremont St at E Berkeley St
42.345392	

	end station longitude	bikeid	usertype	postal code
4	-71.123831	2214	Subscriber	02215
6	-71.077392	4683	Subscriber	02130
8	-71.063840	3431	Subscriber	02215
10	-71.063840	6614	Customer	02128
11	-71.069616	6588	Subscriber	02116

```
avg_trip_duration_start = filtered_trips.groupby('start station id')
['tripduration'].mean().reset_index()
avg_trip_duration_start.rename(columns={'start station id':
'station_id', 'tripduration': 'avg_start_duration'}, inplace=True)
```

Calculate average trip duration for end stations

```
avg_trip_duration_end = filtered_trips.groupby('end station id')
['tripduration'].mean().reset_index()
avg_trip_duration_end.rename(columns={'end station id': 'station_id',
'tripduration': 'avg_end_duration'}, inplace=True)
```

Merge the two dataframes on station_id

```
merged_avg_durations = pd.merge(avg_trip_duration_start,
avg_trip_duration_end, on='station_id', how='outer')
```

Calculating the mean of the two averages, handling cases where one might be NaN

```
merged_avg_durations['avg_trip_duration'] =
merged_avg_durations[['avg_start_duration',
'avg_end_duration']].mean(axis=1, skipna=True)
```

```
all_trip_averages = pd.merge(merged_avg_durations,
filtered_blue_bikes_stations)
print(all_trip_averages[["station_id", "avg_trip_duration",
"avg_start_duration", "avg_end_duration", "route_id"]])
```

	station_id	avg_trip_duration	avg_start_duration
avg_end_duration \			
0	3	17.033691	15.262563
18.804818			
1	4	13.714921	14.297755
13.132086			
2	8	26.393002	36.228563
16.557442			

3	9	13.660425	12.675271
14.645580			
4	10	12.328906	12.048364
12.609447			
..
..			
119	538	14.845833	13.483333
16.208333			
120	544	11.534584	11.609878
11.459290			
121	547	98.564286	97.273810
99.854762			
122	548	27.911111	27.911111
NaN			
123	554	13.976831	14.347641
13.606021			

	route_id
0	19
1	701
2	9703
3	747
4	747
..	...
119	14
120	747
121	14
122	14
123	701

[124 rows x 5 columns]

```
trip_duration_start = filtered_trips.groupby('start station id')
# trip_duration_start.rename(columns={'start station id':
# 'station_id', 'tripduration': 'trip_duration'}, inplace=True)

# Calculate average trip duration for end stations
trip_duration_end = filtered_trips.groupby('end station id')
# trip_duration_end.rename(columns={'end station id': 'station_id',
# 'tripduration': 'trip_duration'}, inplace=True)

print(trip_duration_start.head(), trip_duration_end.head())
```

	tripduration	starttime
stoptime \		
4	12.533333	2022-01-01 00:01:06.0520 2022-01-01
00:13:38.2300		
6	16.383333	2022-01-01 00:01:24.7490 2022-01-01
00:17:48.1850		
8	13.950000	2022-01-01 00:02:30.2650 2022-01-01


```

00:16:27.6270
10      13.350000  2022-01-01 00:03:23.9650  2022-01-01
00:16:45.4150
11      12.483333  2022-01-01 00:03:37.2630  2022-01-01
00:16:07.1390
...      ...      ...      ..
.
75059    23.150000  2022-01-28 11:15:13.7100  2022-01-28
11:38:22.8430
77193    38.716667  2022-01-28 20:07:52.7650  2022-01-28
20:46:36.4570
77749    11.733333  2022-01-28 21:33:41.1780  2022-01-28
21:45:25.6760
79611    52.766667  2022-01-29 04:27:02.8610  2022-01-29
05:19:49.4420
81251    61.066667  2022-01-31 22:33:21.6430  2022-01-31
23:34:26.1770

      start station id      start station
name \
4      19      Park Dr at Buswell St
6      36  Copley Square - Dartmouth St at Boylston St
8      60  Charles Circle - Charles St at Cambridge St
10     60  Charles Circle - Charles St at Cambridge St
11     4    Tremont St at E Berkeley St
...      ...      ...
75059    441      Sullivan Square
77193    540      Sumner St at Shirley Ave
77749    214  Airport T Stop - Bremen St at Brooks St
79611    397      Broadway at Beacham St
81251    236      Assembly Square T

      start station latitude  start station longitude  end station id
\
4      42.347241      -71.105301      41
6      42.349928      -71.077392      36
8      42.360793      -71.071190      363

```

10	42.360793	-71.071190	363	
11	42.345392	-71.069616	4	
...	
75059	42.384452	-71.075149	157	
77193	42.408337	-70.998048	217	
77749	42.374111	-71.032772	355	
79611	42.398361	-71.063738	12	
81251	42.392233	-71.077466	51	
end station name \				
4	Packard's Corner - Commonwealth Ave at Brighto...			
6	Copley Square - Dartmouth St at Boylston St			
8	Harrison Ave at Mullins Way			
10	Harrison Ave at Mullins Way			
11	Tremont St at E Berkeley St			
...	...			
75059	Seaport Blvd at Sleeper St			
77193	Orient Heights T Stop - Bennington St at Sarat...			
77749	Bennington St at Constitution Beach			
79611	Ruggles T Stop - Columbus Ave at Melnea Cass Blvd			
81251	Washington St at Lenox St			
end station latitude end station longitude bikeid usertype				
\				
4	42.352261	-71.123831	2214	Subscriber
6	42.349928	-71.077392	4683	Subscriber
8	42.345216	-71.063840	3431	Subscriber
10	42.345216	-71.063840	6614	Customer
11	42.345392	-71.069616	6588	Subscriber
...
75059	42.353178	-71.048174	2754	Subscriber
77193	42.386781	-71.006098	2328	Subscriber
77749	42.385224	-71.010631	6083	Subscriber
79611	42.336244	-71.087986	3482	Customer

81251	42.335099	-71.079038	3307	Subscriber
-------	-----------	------------	------	------------

	postal code
4	02215
6	02130
8	02215
10	02128
11	02116
...	...
75059	02451
77193	NaN
77749	02128
79611	NaN
81251	02118

[1393 rows x 14 columns]		tripduration		starttime
stoptime \				
4	12.533333	2022-01-01	00:01:06.0520	2022-01-01
00:13:38.2300				
6	16.383333	2022-01-01	00:01:24.7490	2022-01-01
00:17:48.1850				
8	13.950000	2022-01-01	00:02:30.2650	2022-01-01
00:16:27.6270				
10	13.350000	2022-01-01	00:03:23.9650	2022-01-01
00:16:45.4150				
11	12.483333	2022-01-01	00:03:37.2630	2022-01-01
00:16:07.1390				
...
.				
76635	37.666667	2022-01-28	18:17:04.8630	2022-01-28
18:54:45.3170				
78188	8.216667	2022-01-28	22:27:30.0400	2022-01-28
22:35:43.6270				
79363	17.416667	2022-01-29	02:09:05.8190	2022-01-29
02:26:31.3640				
80613	54.333333	2022-01-31	19:32:57.5730	2022-01-31
20:27:18.0300				
81505	28.683333	2022-01-31	23:28:37.5810	2022-01-31
23:57:18.8170				

start station id		start station
name \		
4	19	Park Dr at Buswell St
6	36	Copley Square - Dartmouth St at Boylston St
8	60	Charles Circle - Charles St at Cambridge St

10	60	Charles Circle - Charles St at Cambridge St	
11	4	Tremont St at E Berkeley St	
...	
76635	440	Boston Landing	
78188	210	Bennington St at Byron St	
79363	210	Bennington St at Byron St	
80613	370	Dartmouth St at Newbury St	
81505	47	Cross St at Hanover St	
	start station latitude	start station longitude	end station id
\			
4	42.347241	-71.105301	41
6	42.349928	-71.077392	36
8	42.360793	-71.071190	363
10	42.360793	-71.071190	363
11	42.345392	-71.069616	4
...
76635	42.356561	-71.141675	525
78188	42.383533	-71.016191	489
79363	42.383533	-71.016191	478
80613	42.350961	-71.077828	236
81505	42.362811	-71.056067	482
		end station name	\
4	Packard's Corner - Commonwealth Ave at Brighto...		
6	Copley Square - Dartmouth St at Boylston St		
8	Harrison Ave at Mullins Way		
10	Harrison Ave at Mullins Way		
11	Tremont St at E Berkeley St		
...
76635	California at Chapel		
78188	Day Sq		

79363	Gove St at Orleans St			
80613	Assembly Square T			
81505	Kearins Playground			
	end station latitude	end station longitude	bikeid	usertype
\				
4	42.352261	-71.123831	2214	Subscriber
6	42.349928	-71.077392	4683	Subscriber
8	42.345216	-71.063840	3431	Subscriber
10	42.345216	-71.063840	6614	Customer
11	42.345392	-71.069616	6588	Subscriber
...
76635	42.365149	-71.202360	4138	Customer
78188	42.379295	-71.027733	7423	Subscriber
79363	42.370578	-71.035585	7590	Customer
80613	42.392233	-71.077466	3307	Subscriber
81505	42.406161	-71.060417	6843	Subscriber

	postal code
4	02215
6	02130
8	02215
10	02128
11	02116
...	...
76635	02458
78188	02128
79363	NaN
80613	02118
81505	02149

[1423 rows x 14 columns]

```
average_duration_by_route = all_trip_averages.groupby('route_id')
['avg_trip_duration'].mean()
print(average_duration_by_route.head())
```

route_id	
14	762.125907
19	55.925891

```

41      74.703294
459     31.483486
701     17.881840
Name: avg_trip_duration, dtype: float64

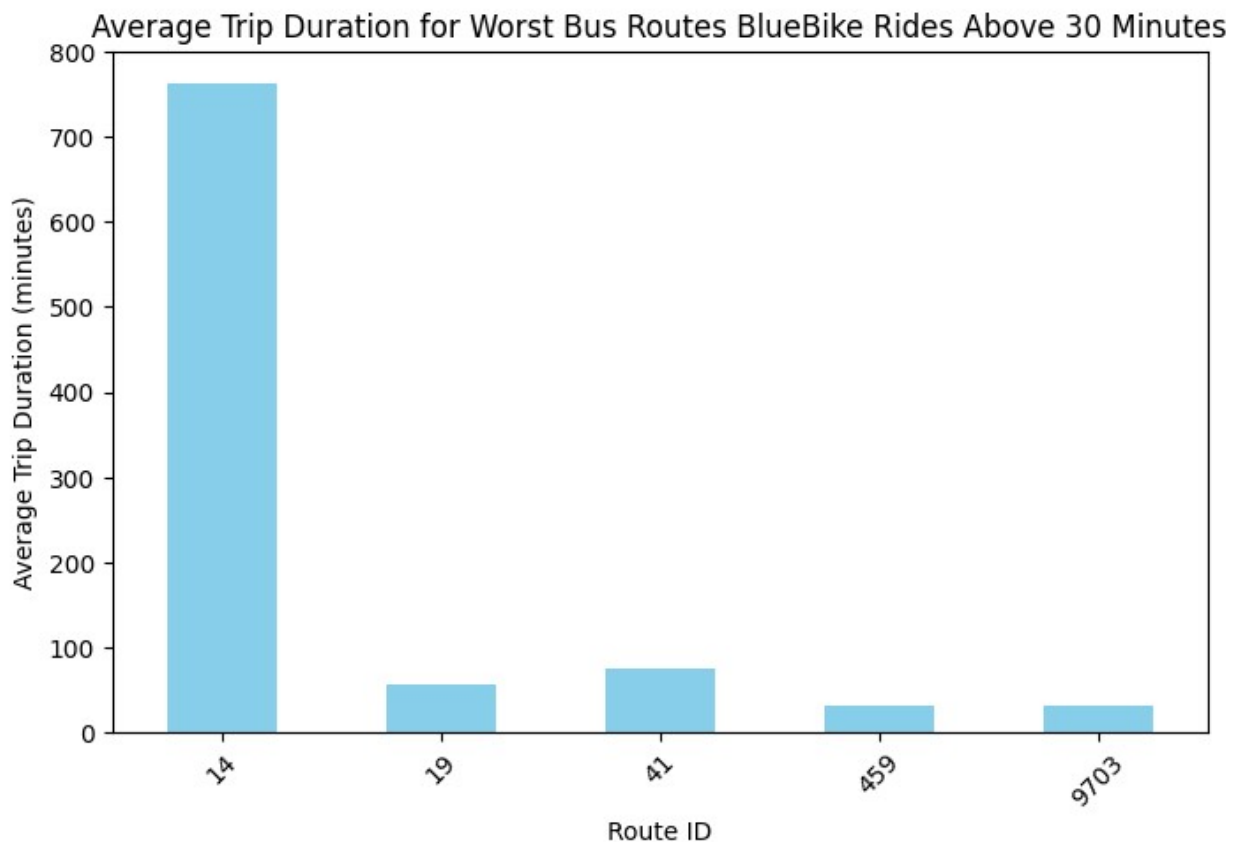
# Filter to include only routes with an average duration above 60
minutes
routes_above_30 = average_duration_by_route[average_duration_by_route
> 30]

# Creating a bar graph for routes with average duration above 60
minutes
plt.figure(figsize=(8, 5))
routes_above_30.plot(kind='bar', color='skyblue')

plt.xlabel('Route ID')
plt.ylabel('Average Trip Duration (minutes)')
plt.title('Average Trip Duration for Worst Bus Routes BlueBike Rides
Above 30 Minutes')
plt.xticks(rotation=45)

plt.show()

```



```
print(len(close_blue_bikes_list["14"]),
len(close_blue_bikes_list["41"]))
```

29 21

Despite the similarity in number of bluebike stations close by those two stations 41, and 14, we see that the average biking duration for route 14 is much much higher. This may be an indication of either of two things: Either the bus route servicing is subpar in the area, or there may be an abundance of explorers there.

```
start_station_trip_count = blue_bikes_trips.groupby('start station
id')['start station
id'].count().reset_index(name='start_station_trip_count')
end_station_trip_count = blue_bikes_trips.groupby('end station id')
["end station id"].count().reset_index(name='end_station_trip_count')

start_station_trip_count.rename(columns={'start station id':
'station_id'}, inplace=True)
end_station_trip_count.rename(columns={'end station id':
'station_id'}, inplace=True)

# Merge the two dataframes on 'station_id'
trip_counts = pd.merge(start_station_trip_count,
end_station_trip_count)
trip_counts.columns = ['start_station_trip_count',
'end_station_trip_count']
trip_counts["difference"] = trip_counts["end_station_trip_count"] -
trip_counts["start_station_trip_count"] # Negative means more stations
that people pick up bikes from.
print(trip_counts.head())
```

```
-----
-----
ValueError                                Traceback (most recent call
last)
```

```
c:\Users\sataa\OneDrive\Documents\Fall 2023\CS 506\Boston Transit
Performance Project\ds-boston-transit-performance\fa23-team-a\
Deliverables\deliverable-3\extension-project.ipynb Cell 20 line 9
<a
```

```
href='vscode-notebook-cell:/c%3A/Users/sataa/OneDrive/Documents/Fall
%202023/CS%20506/Boston%20Transit%20Performance%20Project/ds-boston-
transit-performance/fa23-team-a/Deliverables/deliverable-3/extension-
project.ipynb#X3lsZmlsZQ%3D%3D?line=6'>7</a> # Merge the two
dataframes on 'station_id'
```

```
<a
href='vscode-notebook-cell:/c%3A/Users/sataa/OneDrive/Documents/Fall
%202023/CS%20506/Boston%20Transit%20Performance%20Project/ds-boston-
transit-performance/fa23-team-a/Deliverables/deliverable-3/extension-
project.ipynb#X3lsZmlsZQ%3D%3D?line=7'>8</a> trip_counts =
```

```

pd.merge(start_station_trip_count, end_station_trip_count)
----> <a
href='vscode-notebook-cell:/c%3A/Users/sataa/OneDrive/Documents/Fall
%202023/CS%20506/Boston%20Transit%20Performance%20Project/ds-boston-
transit-performance/fa23-team-a/Deliverables/deliverable-3/extension-
project.ipynb#X31sZmlsZQ%3D%3D?line=8'>9</a> trip_counts.columns =
['start_station_trip_count', 'end_station_trip_count']
<a
href='vscode-notebook-cell:/c%3A/Users/sataa/OneDrive/Documents/Fall
%202023/CS%20506/Boston%20Transit%20Performance%20Project/ds-boston-
transit-performance/fa23-team-a/Deliverables/deliverable-3/extension-
project.ipynb#X31sZmlsZQ%3D%3D?line=9'>10</a>
trip_counts["difference"] = trip_counts["end_station_trip_count"] -
trip_counts["start_station_trip_count"] # Negative means more stations
that people pick up bikes from.
<a
href='vscode-notebook-cell:/c%3A/Users/sataa/OneDrive/Documents/Fall
%202023/CS%20506/Boston%20Transit%20Performance%20Project/ds-boston-
transit-performance/fa23-team-a/Deliverables/deliverable-3/extension-
project.ipynb#X31sZmlsZQ%3D%3D?line=10'>11</a>
print(trip_counts.head())

```

```

File c:\Users\sataa\.virtualenvs\cs506\Lib\site-packages\pandas\core\
generic.py:6218, in NDFrame.__setattr__(self, name, value)
    6216 try:
    6217     object.__getattr__(self, name)
-> 6218     return object.__setattr__(self, name, value)
    6219 except AttributeError:
    6220     pass

```

```

File properties.pyx:69, in
pandas._libs.properties.AxisProperty.__set__()

```

```

File c:\Users\sataa\.virtualenvs\cs506\Lib\site-packages\pandas\core\
generic.py:767, in NDFrame._set_axis(self, axis, labels)
    762 """
    763 This is called from the cython code when we set the `index`
attribute
    764 directly, e.g. `series.index = [1, 2, 3]`.
    765 """
    766 labels = ensure_index(labels)
--> 767 self._mgr.set_axis(axis, labels)
    768 self._clear_item_cache()

```

```

File c:\Users\sataa\.virtualenvs\cs506\Lib\site-packages\pandas\core\
internals\managers.py:227, in BaseBlockManager.set_axis(self, axis,
new_labels)
    225 def set_axis(self, axis: AxisInt, new_labels: Index) -> None:
    226     # Caller is responsible for ensuring we have an Index
object.

```



```
--> 227     self._validate_set_axis(axis, new_labels)
      228     self.axes[axis] = new_labels
```

File c:\Users\sataa\.virtualenvs\cs506\Lib\site-packages\pandas\core\internals\base.py:85, in DataManager._validate_set_axis(self, axis, new_labels)

```
      82     pass
      84 elif new_len != old_len:
--> 85     raise ValueError(
      86         f"Length mismatch: Expected axis has {old_len}
elements, new "
      87         f"values have {new_len} elements"
      88     )
```

ValueError: Length mismatch: Expected axis has 3 elements, new values have 2 elements

Reverse the dictionary

```
station_to_bus_route = {}
for route, stations in close_blue_bikes_list.items():
    for station in stations:
        if station not in station_to_bus_route:
            station_to_bus_route[station] = route
```

```
trip_counts['route_id'] =
trip_counts['station_id'].map(station_to_bus_route)
print(trip_counts.head())
```

```
df = pd.DataFrame(processed_blue_bikes_trips)
```

```
df = df[df['tripduration'] <= 240]
```

```
print(processed_blue_bikes_trips.head())
```

Create a box plot for the 'tripduration' column

```
plt.figure(figsize=(10, 6))
plt.boxplot(df['tripduration'], vert=False) # 'vert=False' makes the
box plot horizontal
plt.title('Box plot of Trip Durations')
plt.xlabel('Duration (in minutes)')
plt.show()
```

```
(81449, 14)
```

	tripduration	starttime	stoptime \
0	9.950000	2022-01-01 00:00:25.1660	2022-01-01 00:10:22.1920
1	6.850000	2022-01-01 00:00:40.4300	2022-01-01 00:07:32.1980
2	7.933333	2022-01-01 00:00:54.8180	2022-01-01 00:08:51.6680
3	7.766667	2022-01-01 00:01:01.6080	2022-01-01 00:08:48.2350
4	12.533333	2022-01-01 00:01:06.0520	2022-01-01 00:13:38.2300

start station id	start station name	start station
latitude \		

0	178	MIT Pacific St at Purrington St
42.359573		
1	189	Kendall T
42.362428		
2	94	Main St at Austin St
42.375603		
3	94	Main St at Austin St
42.375603		
4	19	Park Dr at Buswell St
42.347241		

	start station longitude	end station id \
0	-71.101295	74
1	-71.084955	178
2	-71.064608	356
3	-71.064608	356
4	-71.105301	41

	end station name	end station latitude \
0	Harvard Square at Mass Ave/ Dunster	
42.373268		
1	MIT Pacific St at Purrington St	
42.359573		
2	Charlestown Navy Yard	
42.374125		
3	Charlestown Navy Yard	
42.374125		
4	Packard's Corner - Commonwealth Ave at Brighto...	
42.352261		

	end station longitude	bikeid	usertype	postal code
0	-71.118579	4923	Subscriber	02139
1	-71.101295	3112	Subscriber	02139
2	-71.054812	6901	Customer	02124
3	-71.054812	5214	Customer	02124
4	-71.123831	2214	Subscriber	02215

Box plot of Trip Durations

