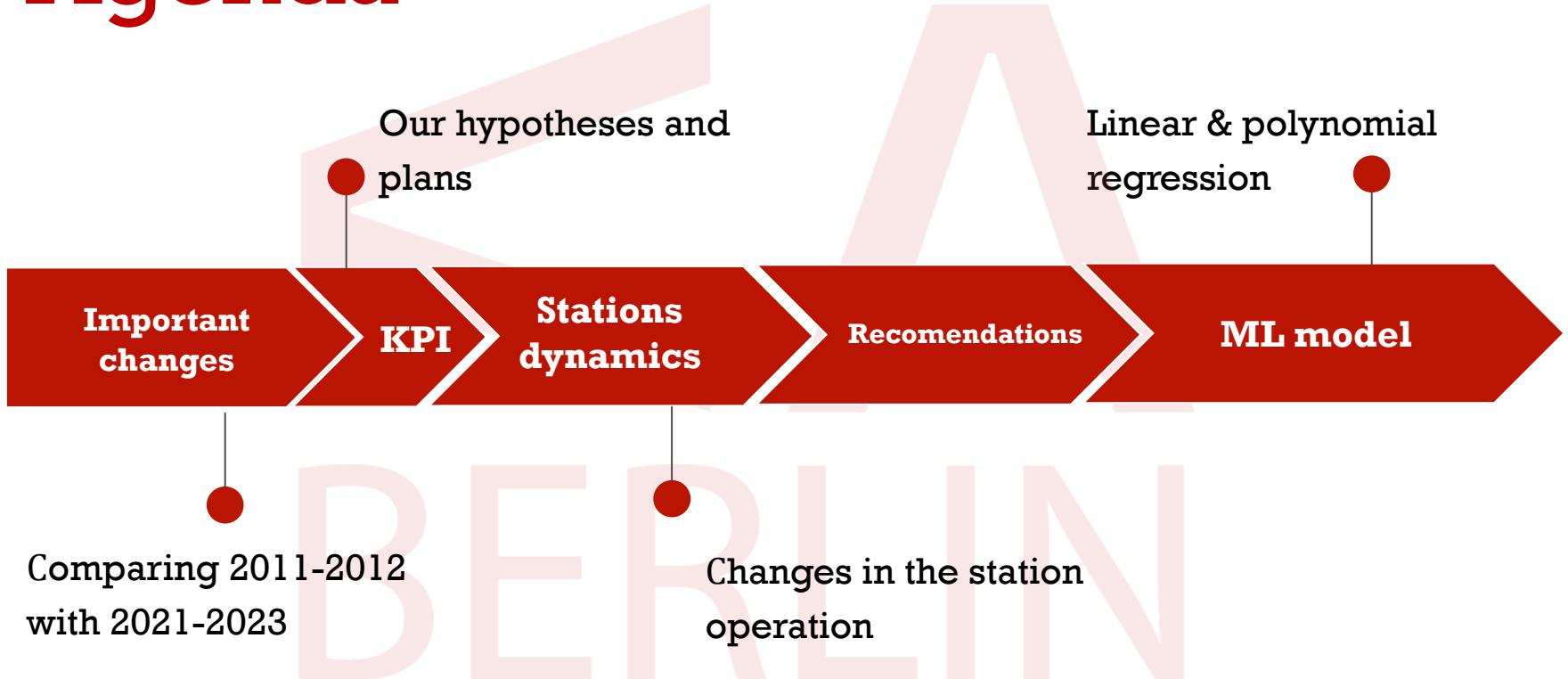


Capital Bike Share: a decade of changes and growth

Yurii Novachynskyi • 09.11.2023



Agenda



2011-2023 Important Changes



MORE bikes

Classic

(2011)



Classic
+Electro*

(2019)

*Not included in the analysis

MORE stations

144
(2011)

697
(2021)

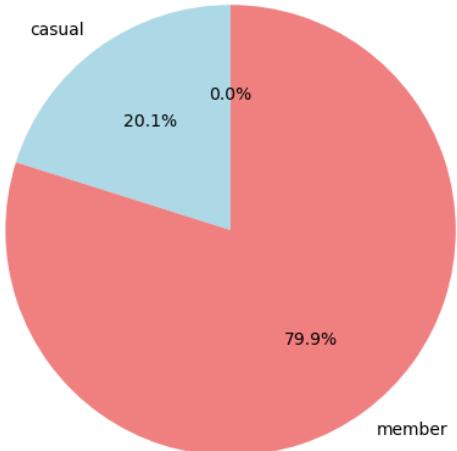
749
(2023)



MORE members

2011

Member Type Distribution for 2011



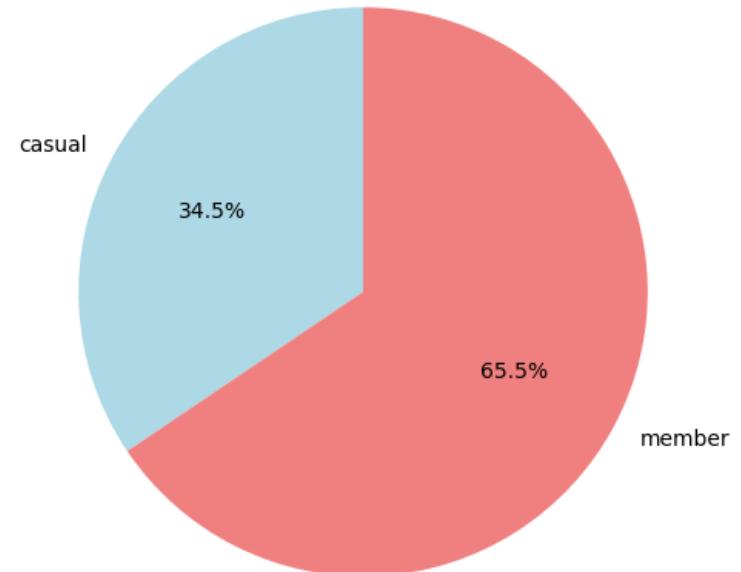
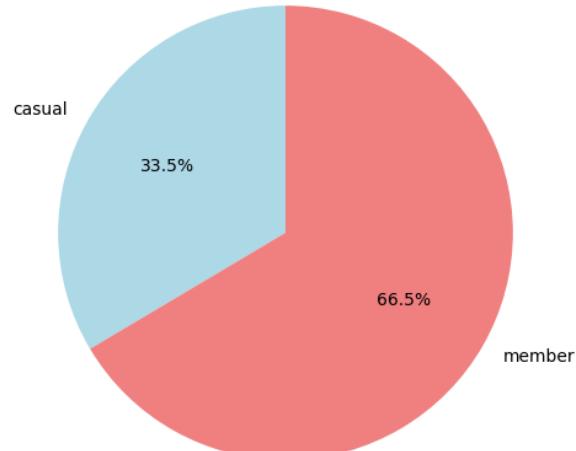
Member Type Distribution

2023

Member Type Distribution for 2023

2021

Member Type Distribution for 2021



MORE free time for members

30 min
(2011)

45 min
(2023)

60 min
(2023 for qualified residents)





Total amount of rides

1.4m

(01-06/2023)

2.7m

(2022)

2.0m

(2021)

2.0m

(2012)

1.2m

(2011)



Total amount of rides

1.4m

(01-06/2023)

\approx 2-3%

2.7m

(2022)

+34,81%

2.0m

(2021)

-0,15%

2.0m

(2012)

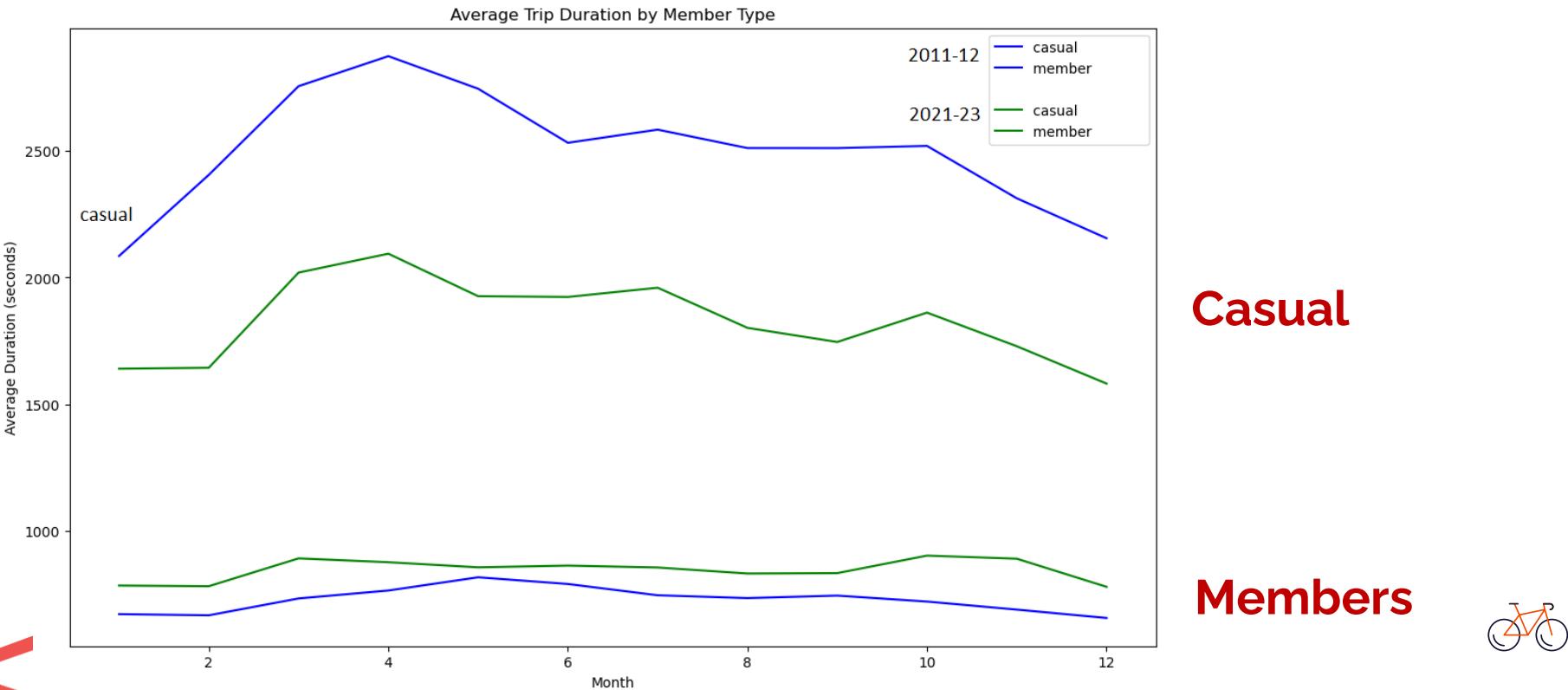
+65,39%

1.2m

(2011)

Average Ride Duration

Duration of member rides is getting longer while of casual participants is shrinking.



% of mistake
unlocks (up to 2 min)

2,95%

(01-06/2023)

2,63%

(2022)

2,55%

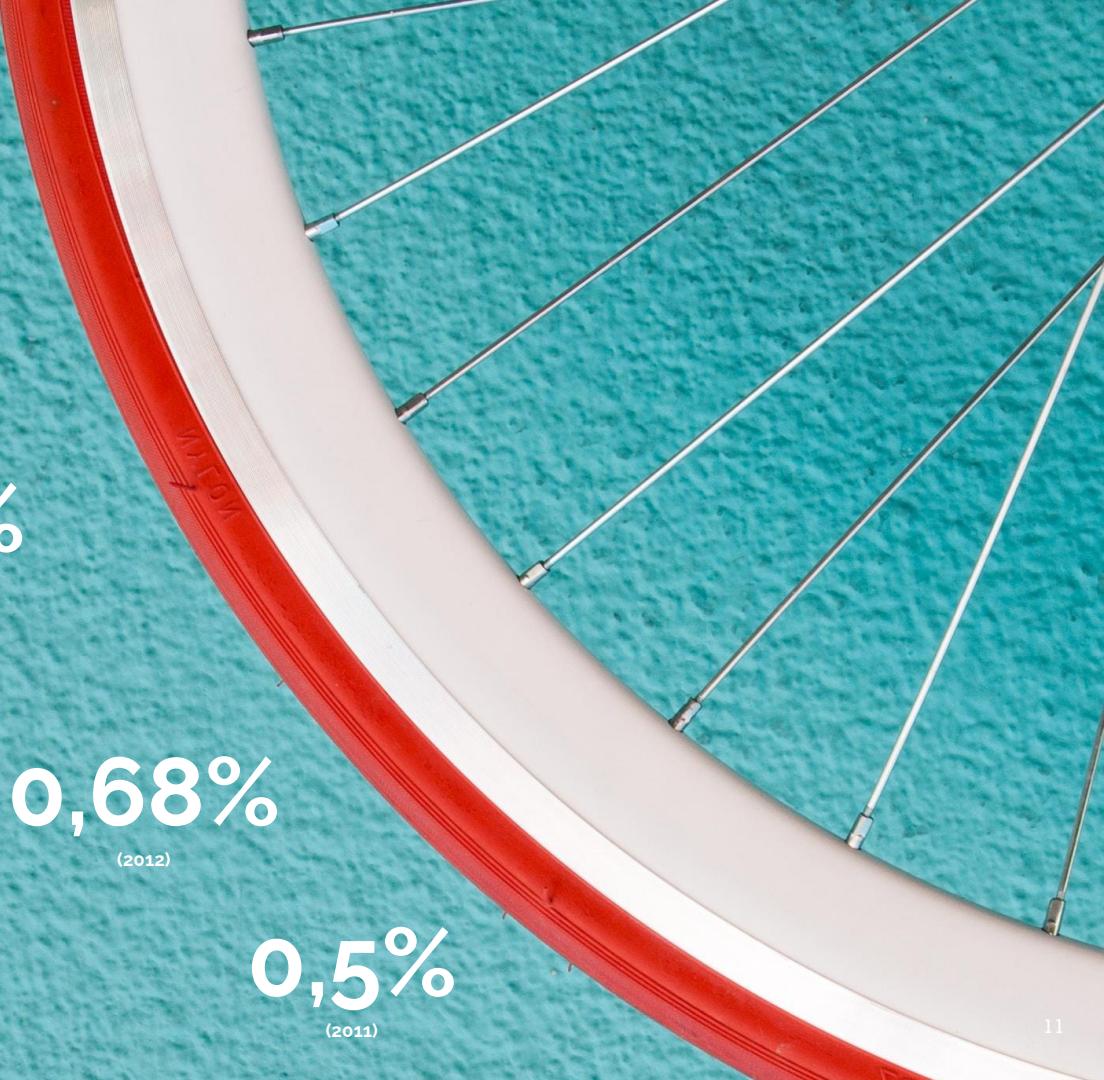
(2021)

0,68%

(2012)

0,5%

(2011)



% of mistake
unlocks (up to 2 min)

21,7% casual **2,95%**

(01-06/2023)

23,5% casual **2,63%**

(2022)

22,7% casual **2,55%**

(2021)

1,3% casual **0,68%**

(2012)

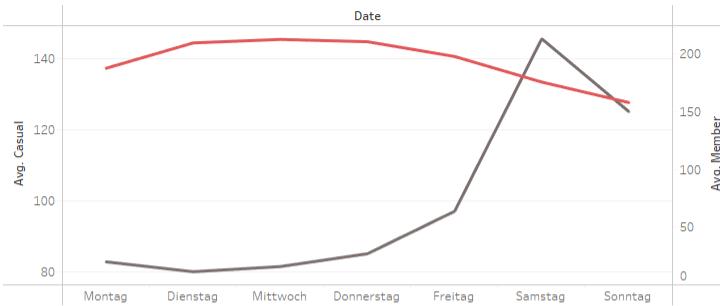
2% were casual rides

0,5%
(2011)

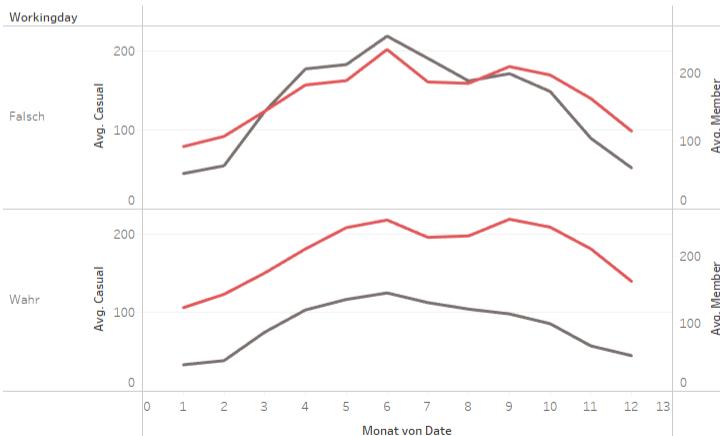


Summary

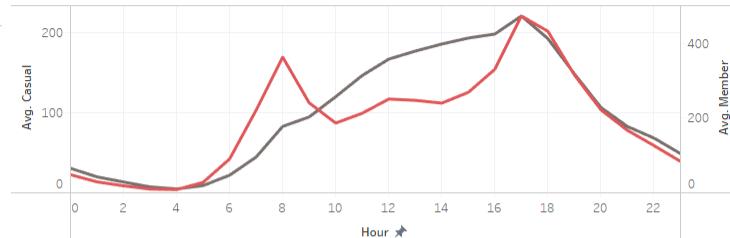
Weekday



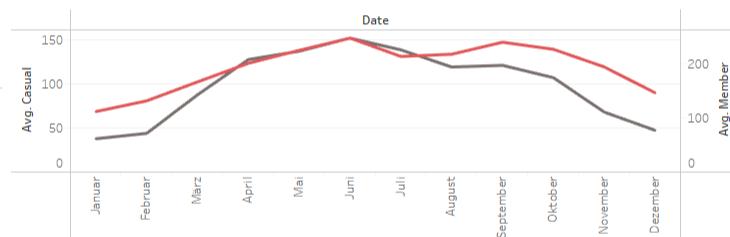
Workingday



Hours



Month_AVG



Years



Measure Names

Casual
Member

Year of Date

- (All)
- 2011
- 2012
- 2021
- 2022
- 2023

Month of Date

- (All)
- Januar
- Februar
- März
- April
- Mai
- Juni
- Juli
- August
- September
- Oktober
- November
- Dezember

Weekday of Date

- (All)
- Montag
- Dienstag
- Mittwoch
- Donnerstag
- Freitag
- Samstag
- Sonntag

Workingday

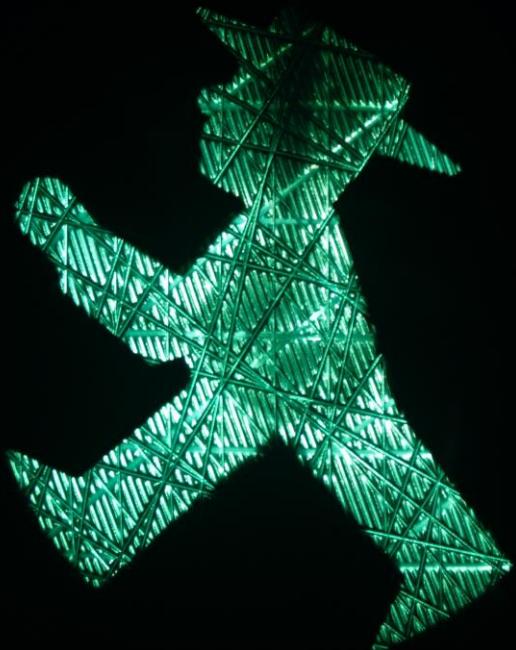
- (All)
- Falsch
- Wahr

Years



Night “wrong rides”*

*duration >2h and ended on next day



3.8k

7.5k

(01-06/2023)

5.2k

(2021)

1.2k

(2012)

1.7k

(2011)

Key Performance Indicators



KPI

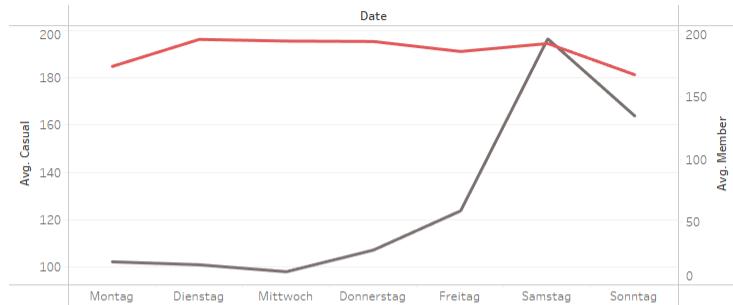


Implement 3 marketing activities
to achieve a 10% increase in the
use of bicycles by regular users
on weekends by the end of 2024

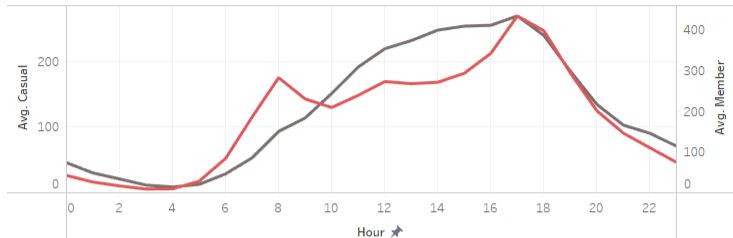
Specific Measurable Attainable Relevant Time-bound

2021

Weekday



Hours



Measure Names
 Casual
 Member

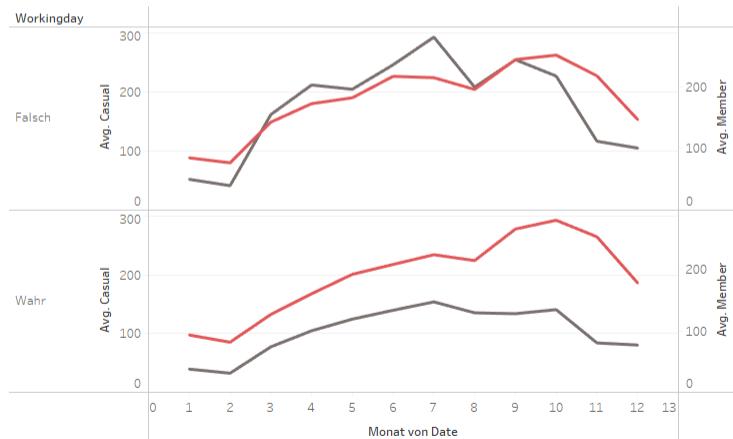
Year of Date
 (All)
 2011
 2012
 2021
 2022
 2023

Month of Date
 (All)
 Januar
 Februar
 März
 April
 Mai
 Juni
 Juli
 August
 September
 Oktober
 November
 Dezember

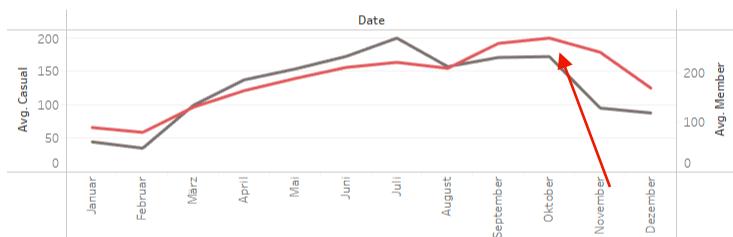
Weekday of Date
 (All)
 Montag
 Dienstag
 Mittwoch
 Donnerstag
 Freitag
 Samstag
 Sonntag

Workingday
 (All)
 Falsch
 Wahr

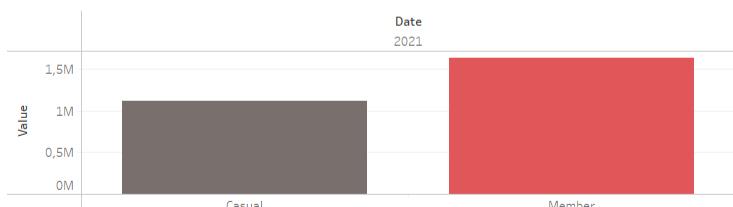
Workingday



Month_AVG



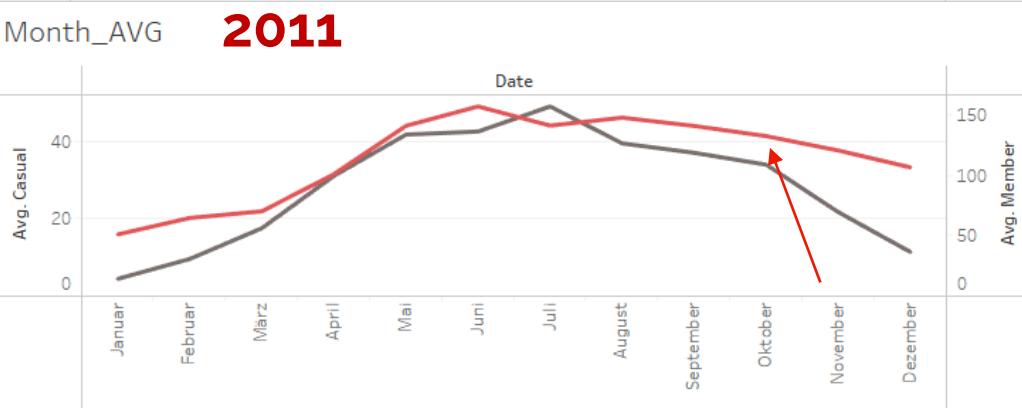
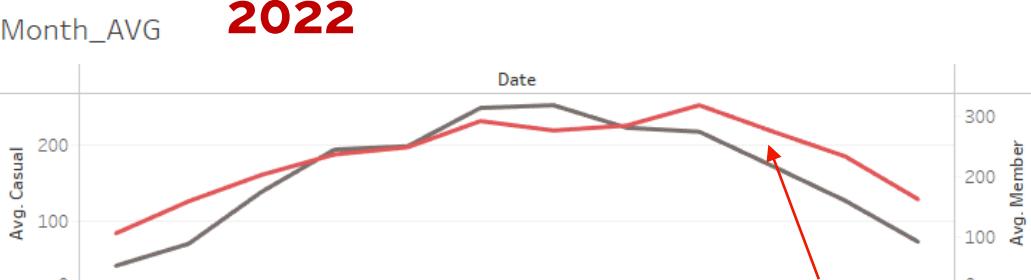
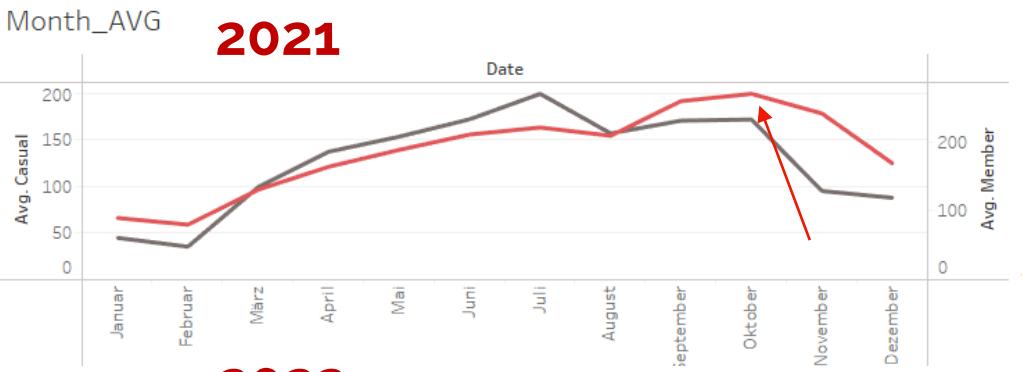
Years



Years Hours Weekday Month_AVG Month_AVG (2) Day AVG Workingday weather weather (2)

Dashboard 2 Dashboard 2 (2)







DC Residents Can Get a Free 30-Day Membership to Capital Bikeshare

The offer is meant to help alleviate some travel issues caused by reductions in Metro service

WRITTEN BY JANE RECKER [Twitter](#) [Email](#) | PUBLISHED ON OCTOBER 25, 2021

[Tweet](#) [Share](#)



Photograph by ablokhin/via iStock.

In light of significant reductions in Metro service, DC has partnered with Lyft to offer a free 30-day Capital Bikeshare membership to all DC residents. To sign up, residents should go to the "Ride Plans" section of the Capital Bikeshare or Lyft apps, or the "Pricing" section of the [Capital Bikeshare website](#).

Those who take advantage of the membership will get unlimited free 45-minute rides

Source: <https://www.washingtonian.com/2021/10/25/dc-residents-can-get-a-free-30-day-membership-to-capital-bikeshare/>

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Washingtonian Magazine

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Increase by 3% the use of bicycles by regular users on Mondays and Tuesdays in May by increasing notifications in the App.

BERLIN

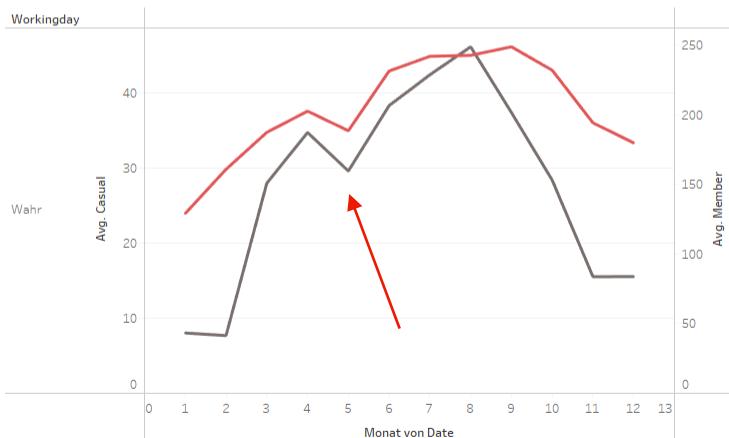
Specific Measurable Attainable Relevant Time-bound

May Weekdays

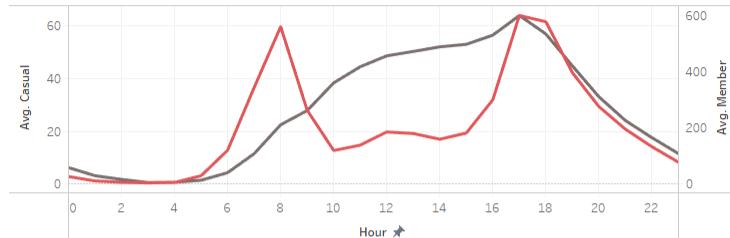
Weekday



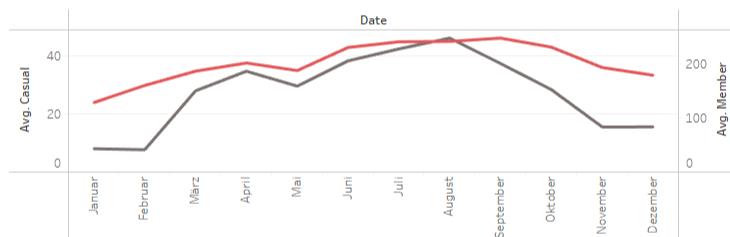
Workingday



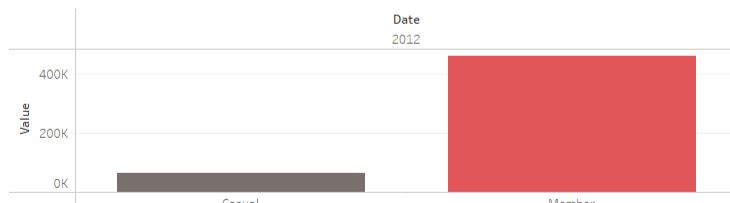
Hours



Month_AVG



Years



Measure Names

Casual

Member

Year of Date

(All)

2011

2012

2021

2022

2023

Month of Date

(All)

Januar

Februar

März

April

Mai

Juni

Juli

August

September

Oktober

November

Dezember

Weekday of Date

(All)

Montag

Dienstag

Mittwoch

Donnerstag

Freitag

Samstag

Sonntag

Workingday

(All)

Falsch

Wahr

May Weekdays



All weekdays

Only
Mondays&Tuesdays



Stations' Dynamics



Approximate stations distribution



2011 (194)

91	21
51	18

13 not identified*
(mostly from SE)

2021 (697)

234	90
147	45

181 not identified

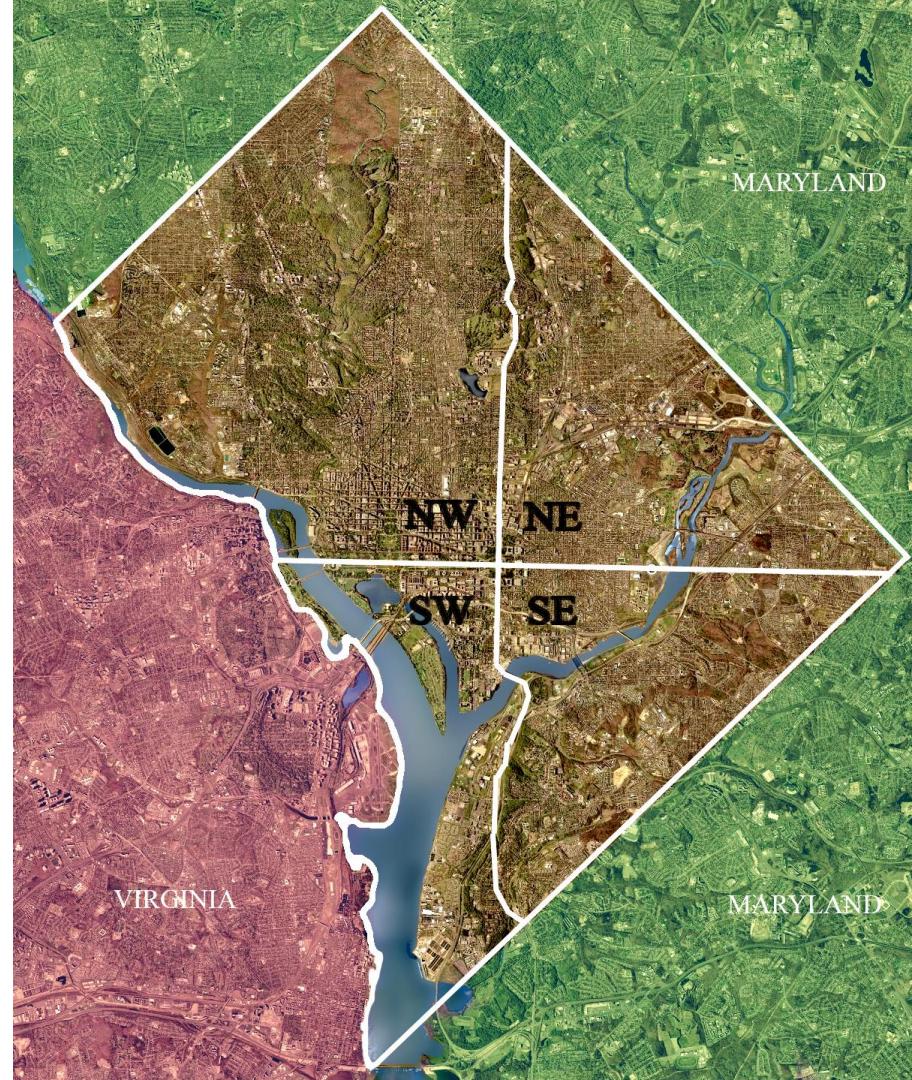
2023 (749)

240	103
154	49

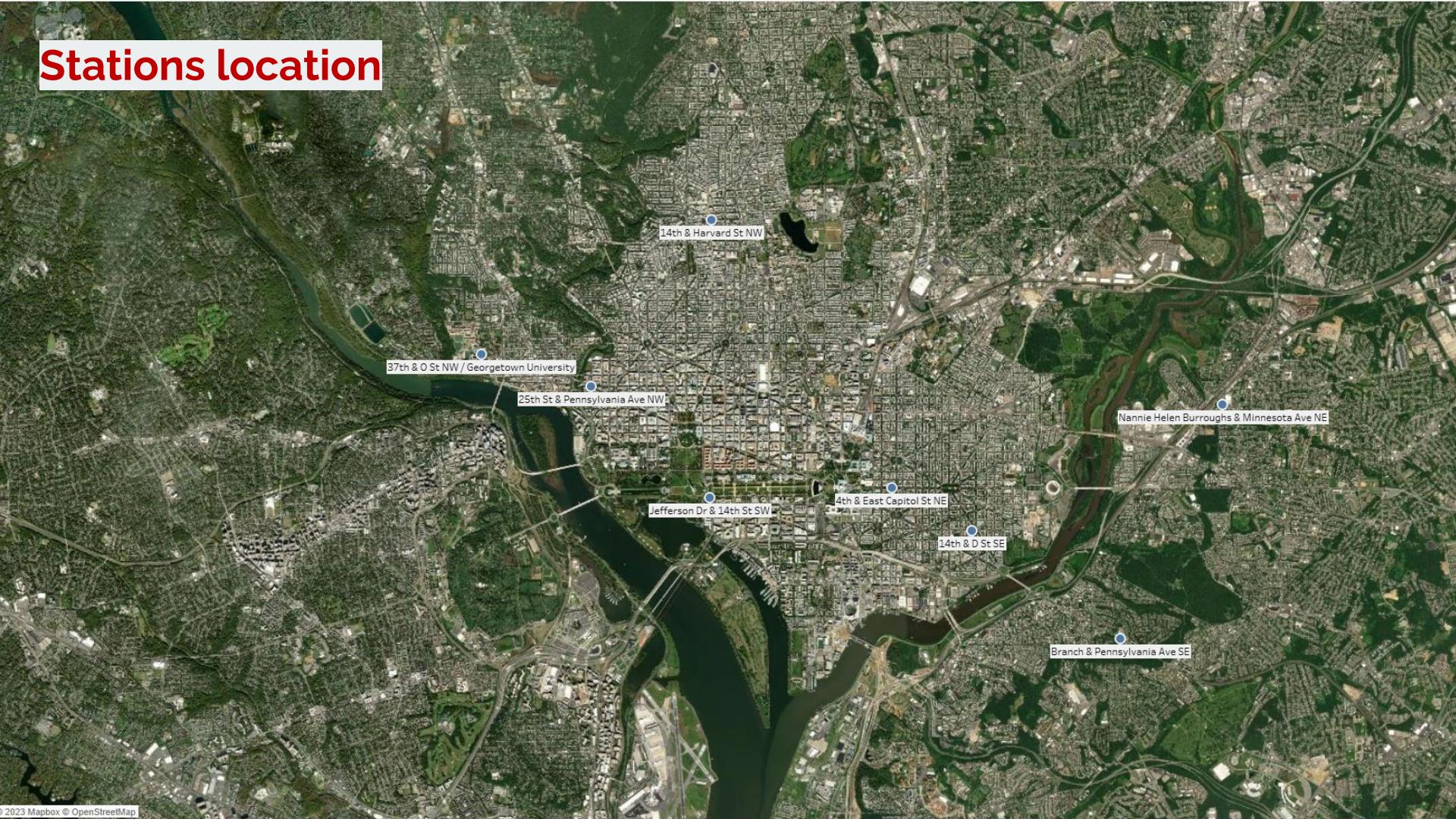
202 not identified



*Identification was automated in Python using 2 algorithms:
station name analysis and coordinate analysis



Stations location



4th & East Capitol St NE

14th & D St SE

14th & Harvard St NW

25th St & Pennsylvania Ave NW

37th & O St NW / Georgetown University

Jefferson Dr & 14th St SW

Branch & Pennsylvania Ave SE

Nannie Helen Burroughs & Minnesota Ave NE

touristic spot

center

residential area

between residential area and center

Students

touristic spot

residential area

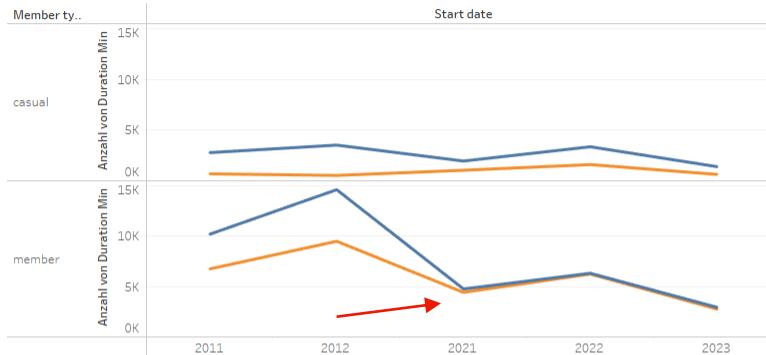
transport hub



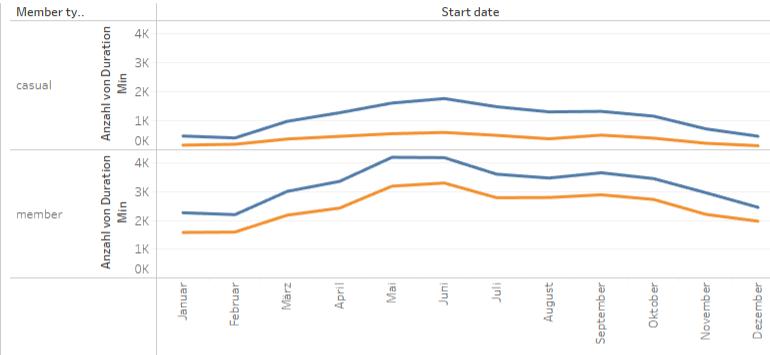
4th & East Capitol St NE

Start stati.. 4th & East Capi.. 14th & D St SE

Year changes



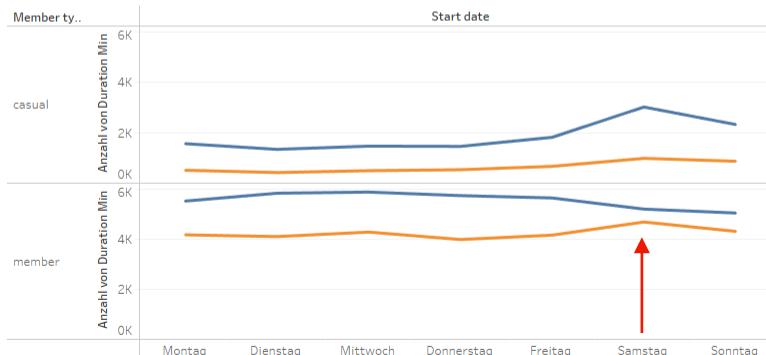
Months



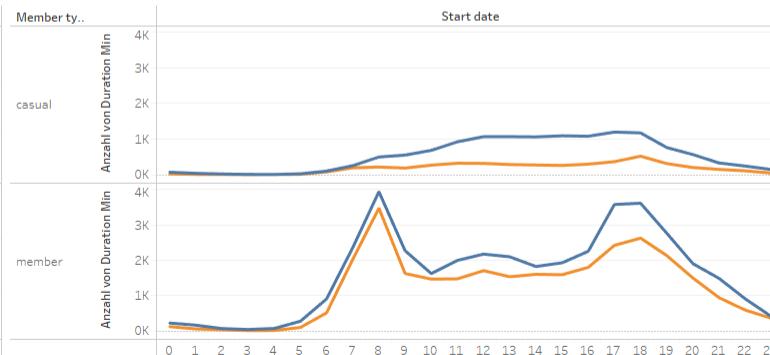
Start station

- (All)
- 4th & East Capi..
- 14th & D St SE
- 14th & Harvard ...
- 25th St & Penns...
- 37th & O St NW ...
- Branch & Penns...
- Jefferson Dr & 1...
- Nannie Helen B...

Week dynamic



Daily dynamic



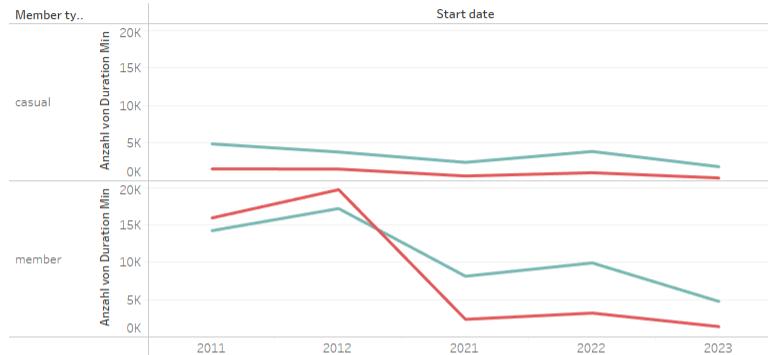
14th & Harvard St NW

25th St & Pennsylvania Ave NW

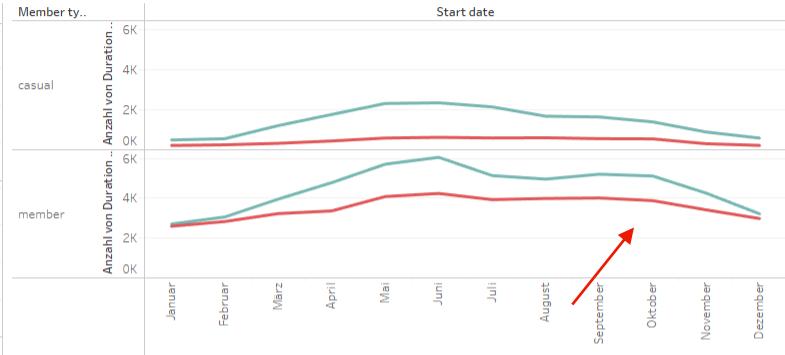
RB

Start stati.. ■ 14th & Harvard .. ■ 25th St & Penn..

Year changes



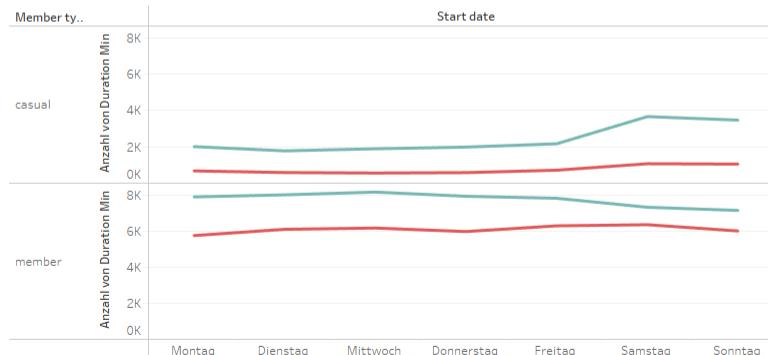
Months



Start station

- (All)
- 4th & East Capi...
- 14th & D St SE
- 14th & Harvard ...
- 25th St & Penns...
- 37th & O St NW ...
- Branch & Penns...
- Jefferson Dr & 1...
- Nannie Helen B...

Week dynamic



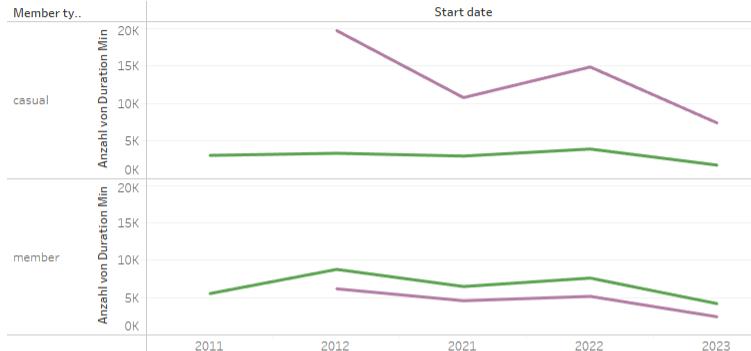
Daily dynamic



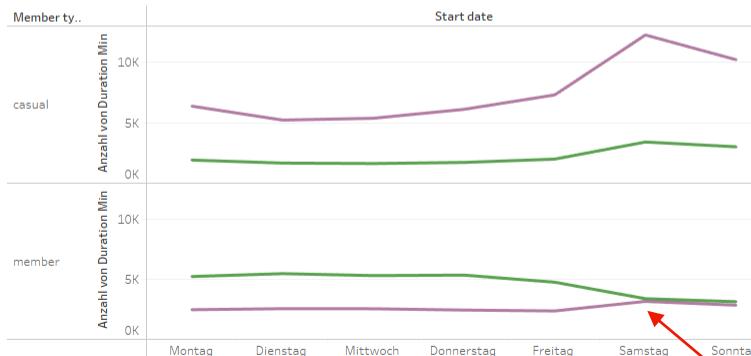
37th & O St NW / Georgetown University

Start stati.. 37th & O St NW.. Jefferson Dr & ..

Year changes



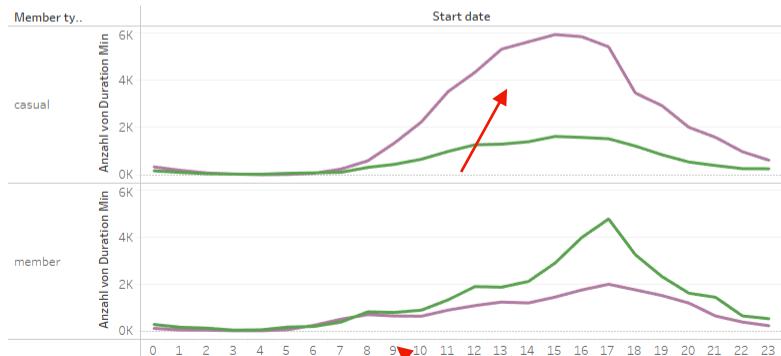
Week dynamic



Months



Daily dynamic



ST

Jefferson Dr & 14th St SW

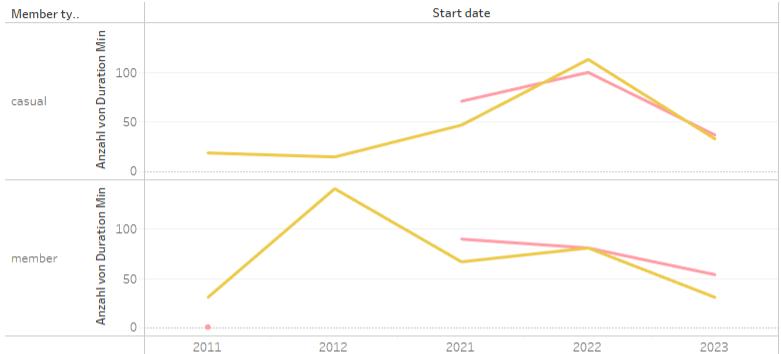
Start station
 (All)
 4th & East Capi...
 14th & D St SE
 14th & Harvard ...
 25th St & Penns...
 37th & O St NW ...
 Branch & Penns...
 Jefferson Dr & 1...
 Nannie Helen B...

Branch & Pennsylvania Ave SE

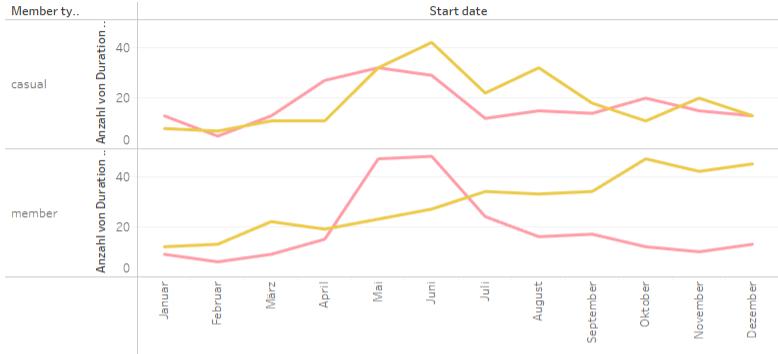
Nannie Helen Burroughs & Minnesota Ave NE

Start stati.. █ Branch & Penns.. █ Nannie Helen B..

Year changes



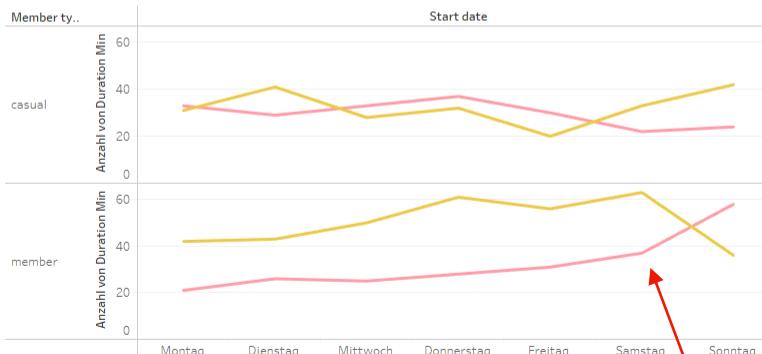
Months



- Start station
- (All)
 - 4th & East Capi...
 - 14th & D St SE
 - 14th & Harvard ...
 - 25th St & Penns...
 - 37th & O St NW ...
 - Branch & Penns...
 - Jefferson Dr & 1...
 - Nannie Helen B...

RH

Week dynamic



Daily dynamic



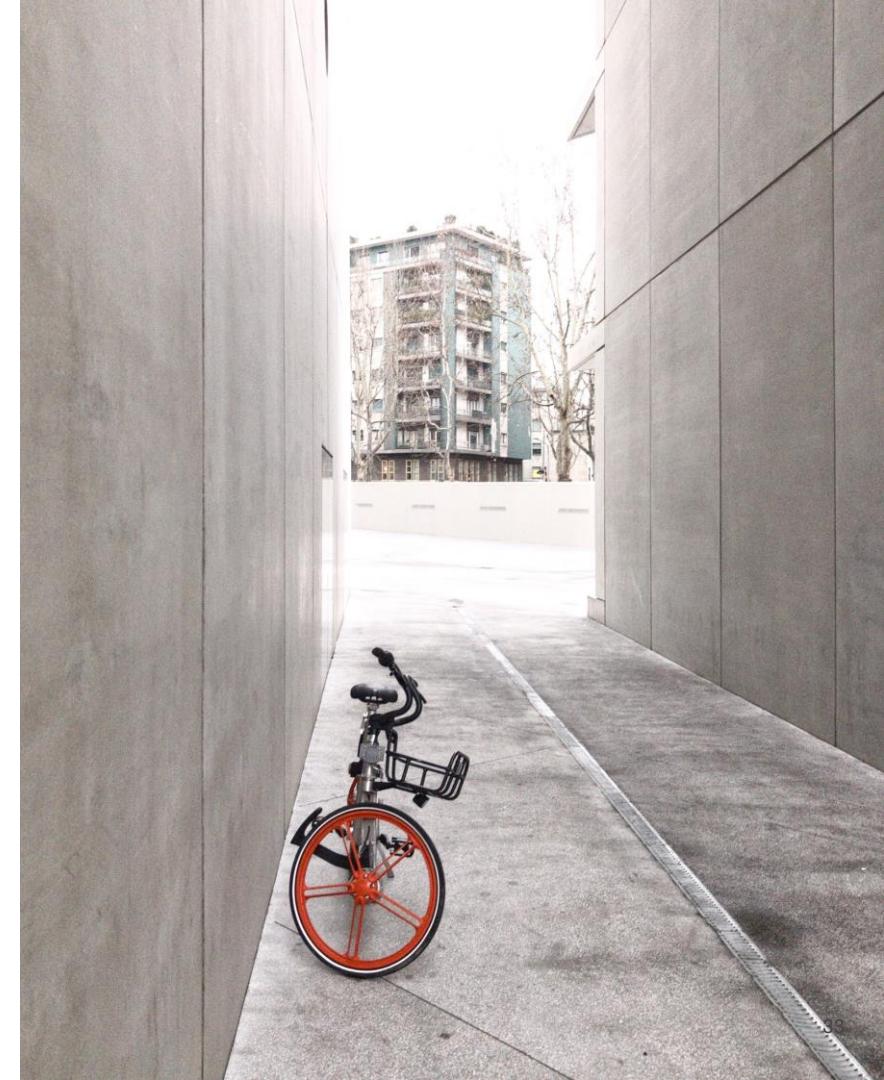
Recommendations



Recommendations

Changing the formats of user interaction with the service (**through the application**) creates opportunities for personalising offers for regular users.

Increasing the number of **regular users** has the potential to improve service quality monitoring.



Recommendations

Taking into account not only general seasonal fluctuations in demand, but also individual station data offers the potential to optimise bicycle services.



Recommendations

The blocking system needs to be improved.

The statistics of false unlocks and incorrect locks are growing faster than overall growth.



BEN

Recommendations

Data to analyse the hourly load of each station would be useful.
How many bicycles or free spots are available etc.

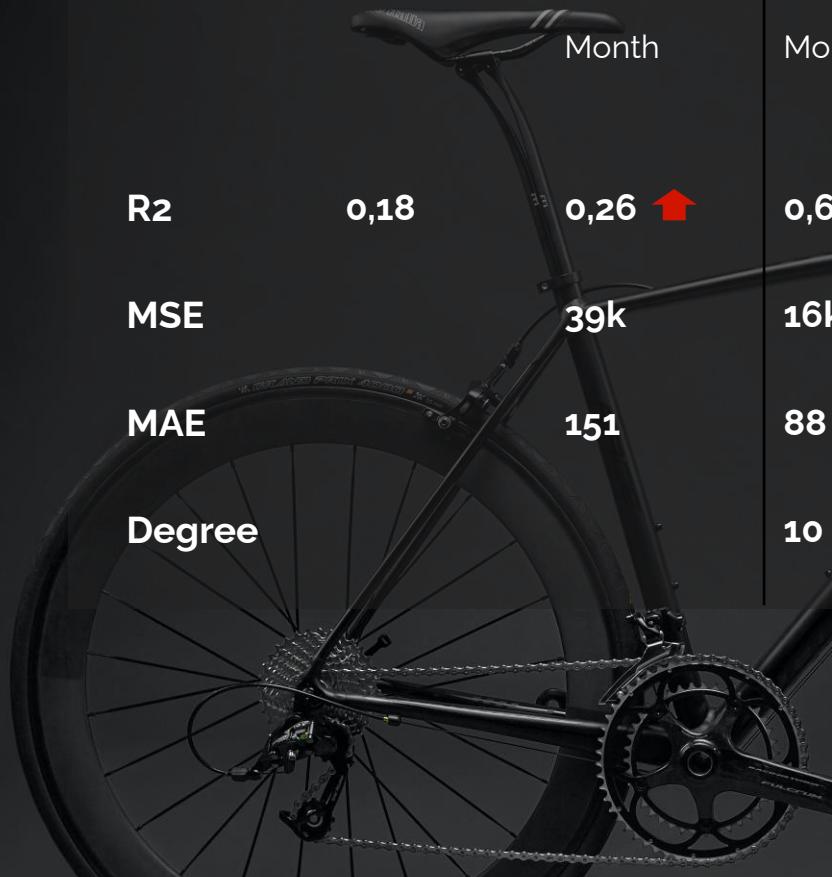


Maschine Learning model

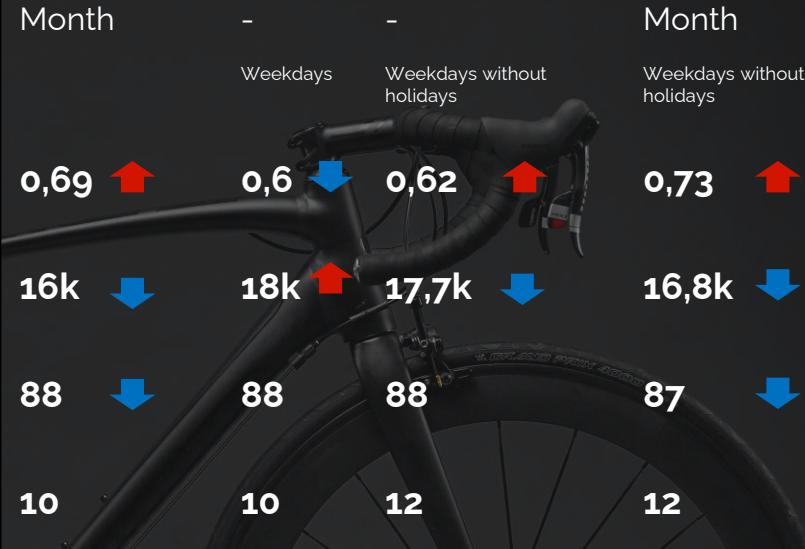


Comparison of results

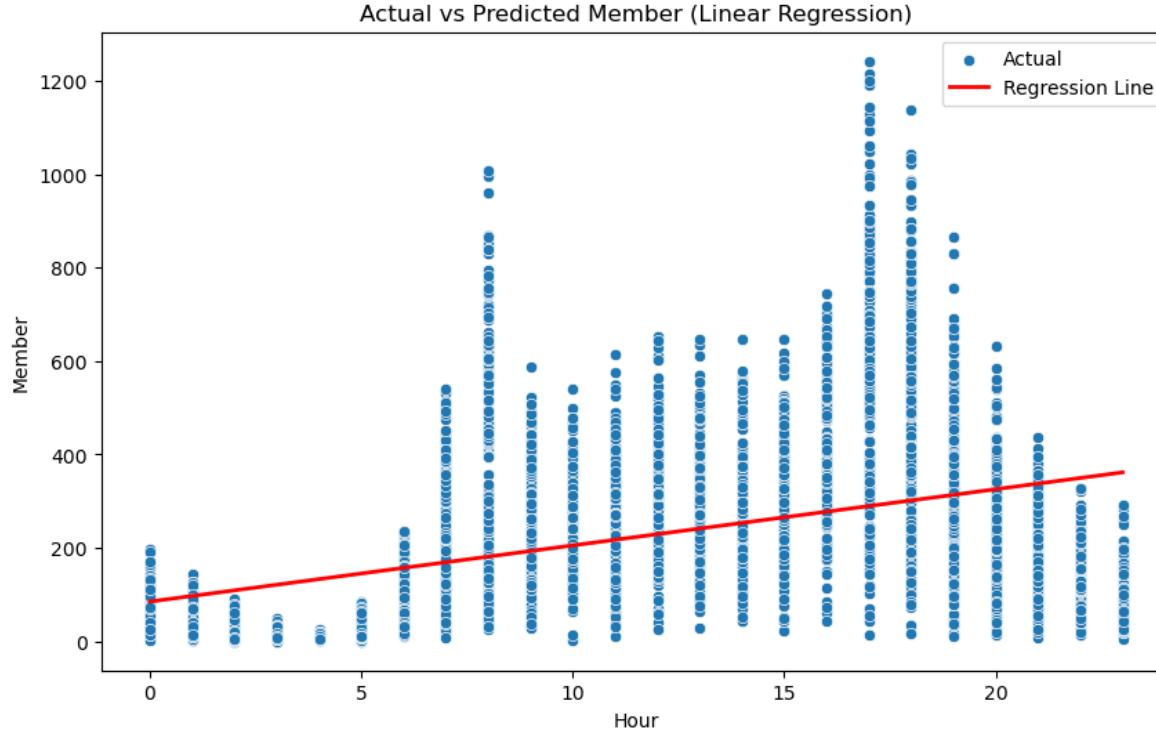
Linear Regression



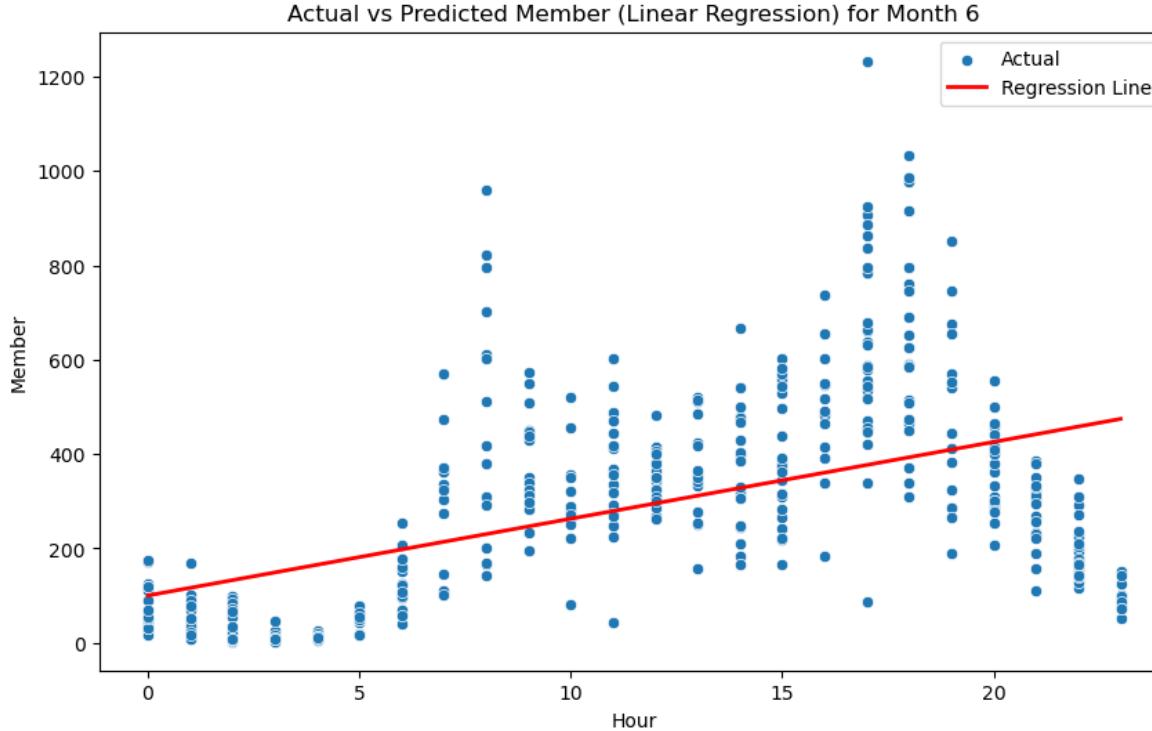
Polynomial Regression



Linear Regression forecasting member rides demand by hours

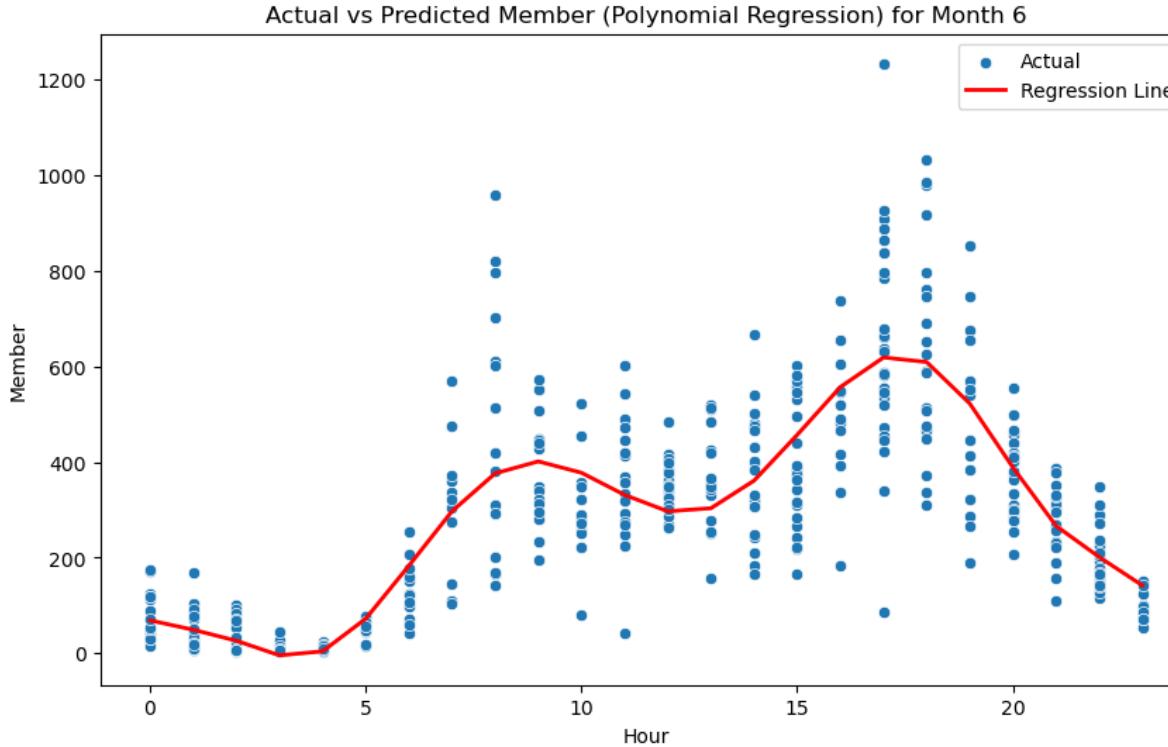


Linear Regression forecasting member rides demand by hours and month (in this case June)

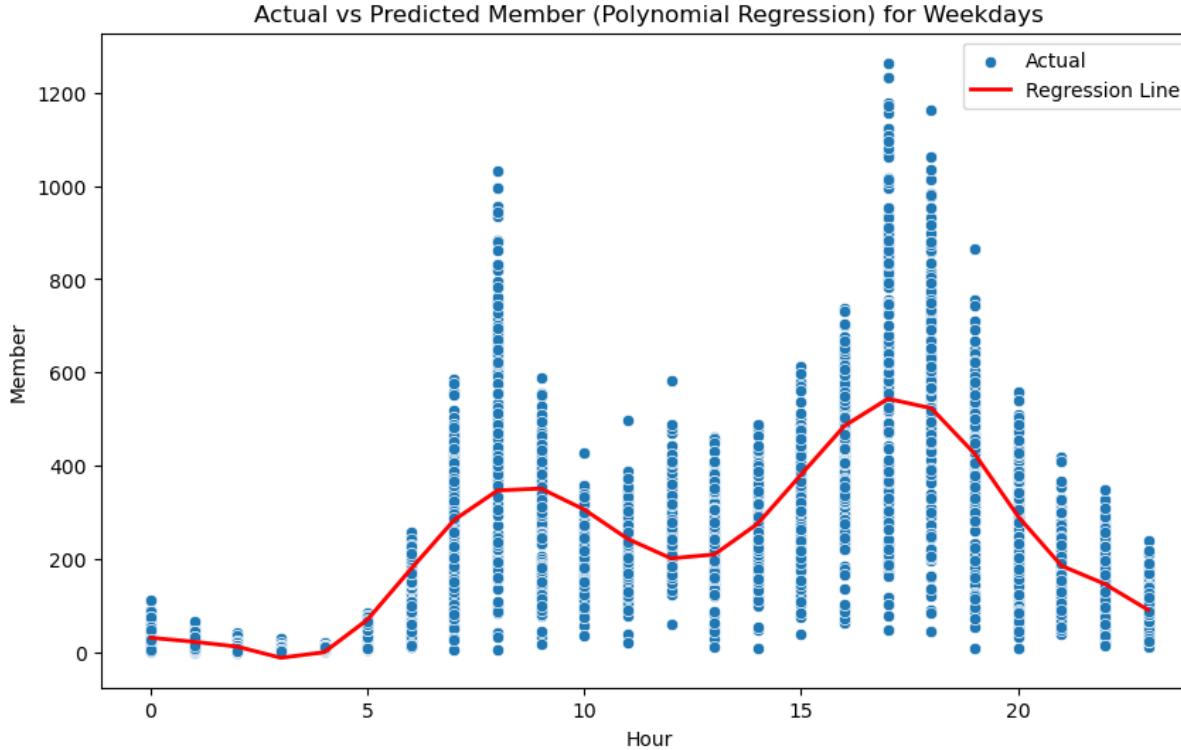


Polynomial Regression

forecasting member rides demand by hours and month (in this case June)

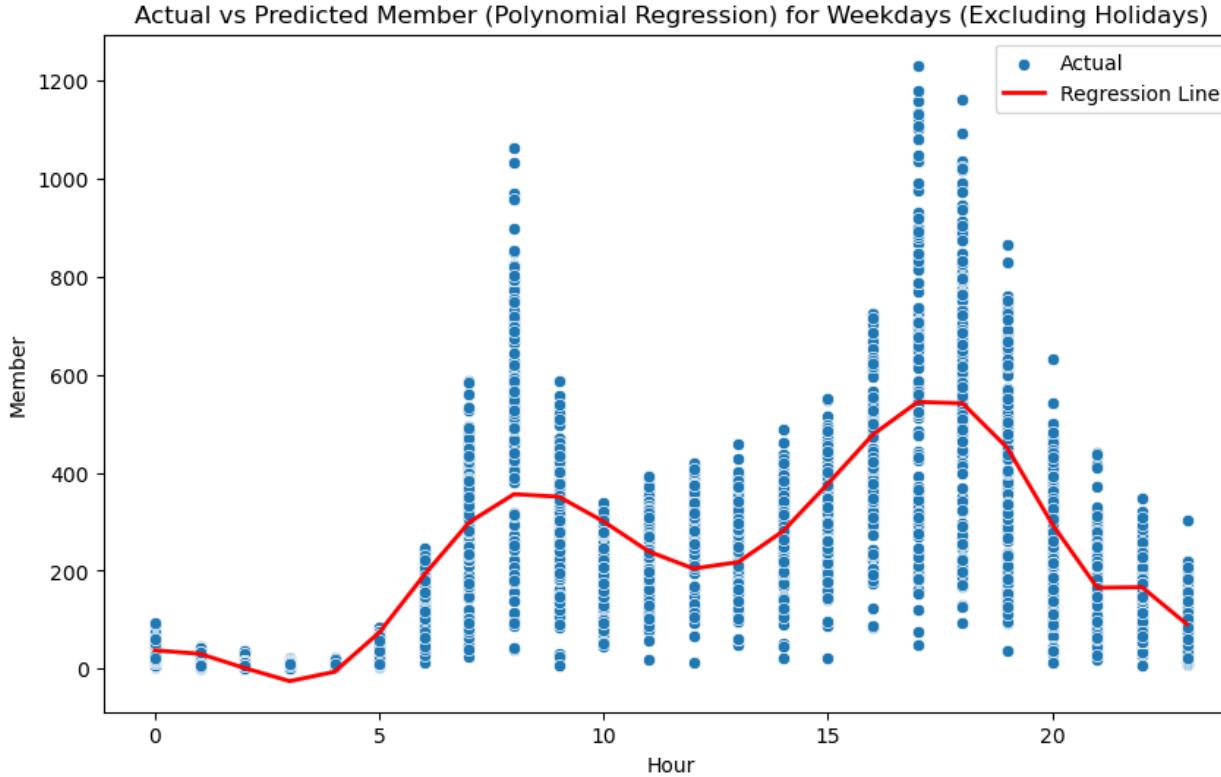


Polynomial Regression forecasting member rides demand by hours and weekdays



Polynomial Regression

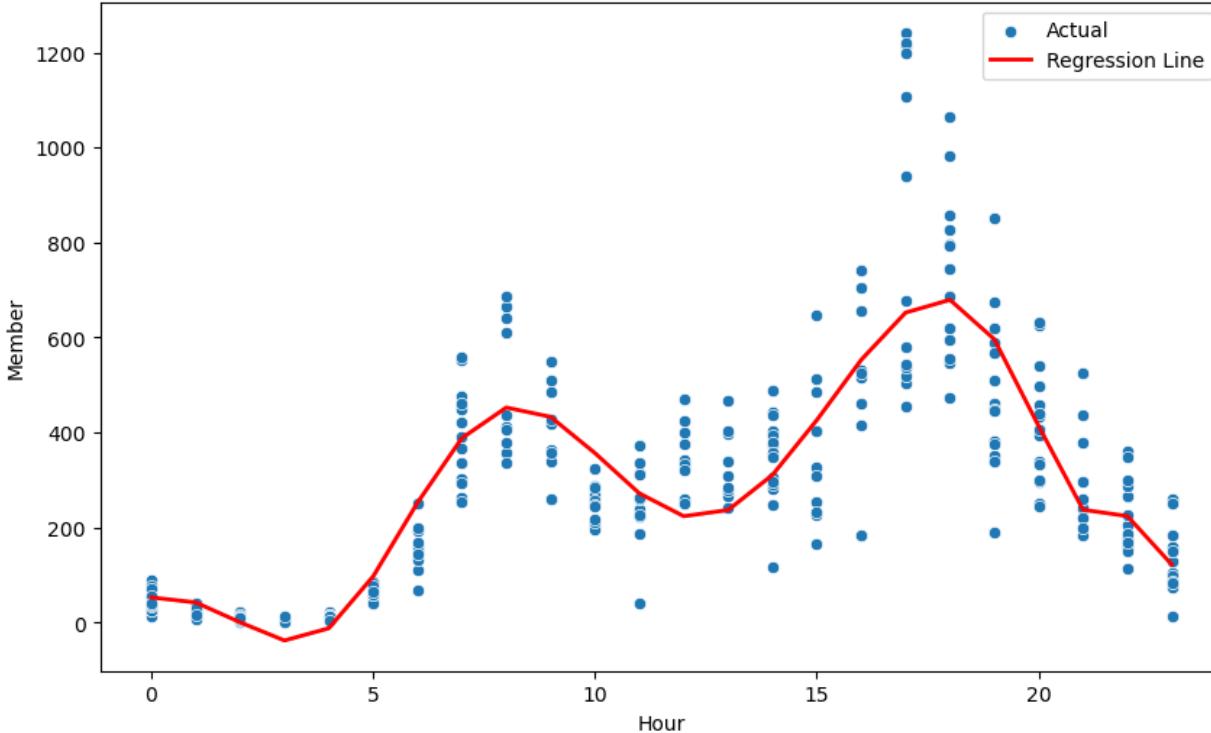
forecasting member rides demand by hours and weekdays (without holidays)



Polynomial Regression

forecasting member rides demand by hours and weekdays (without holidays) for June

Actual vs Predicted Member (Polynomial Regression) for Weekdays (Excluding Holidays) - Month 6



R²

0,73

MSE

16,8k

MAE

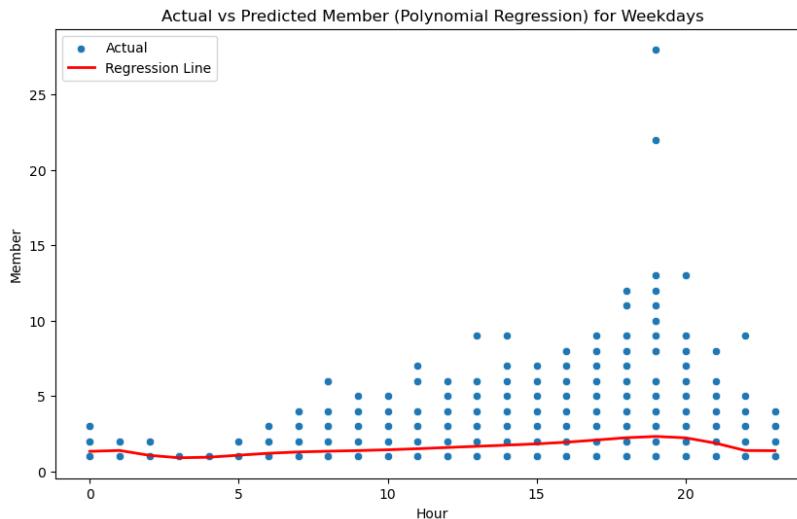
87

Degree

12

Jefferson Dr & 14th St SW

Unsuccessful attempt to implement polynomial regression for forecasting member rides demand by hours for 1 station



- not enough data
- grouping stations patterns can show better results
- this situation doesn't need ML model

A close-up photograph of a bicycle's rear wheel and drivetrain. The red frame of the bike is visible on the left and top right. The central focus is the silver-colored multi-speed cassette and the black chain. The background is blurred.

Thank you for your attention
BERLIN

The project was implemented as part of the Datanalysis & Machine Learning Bootcamp at Code Academy Berlin
10/21 week of the program

Pictures source: <https://unsplash.com/> <https://www.wikipedia.org/>