
MUSIC LISTENING ANALYSIS

Introduction to Visualization Dashboard

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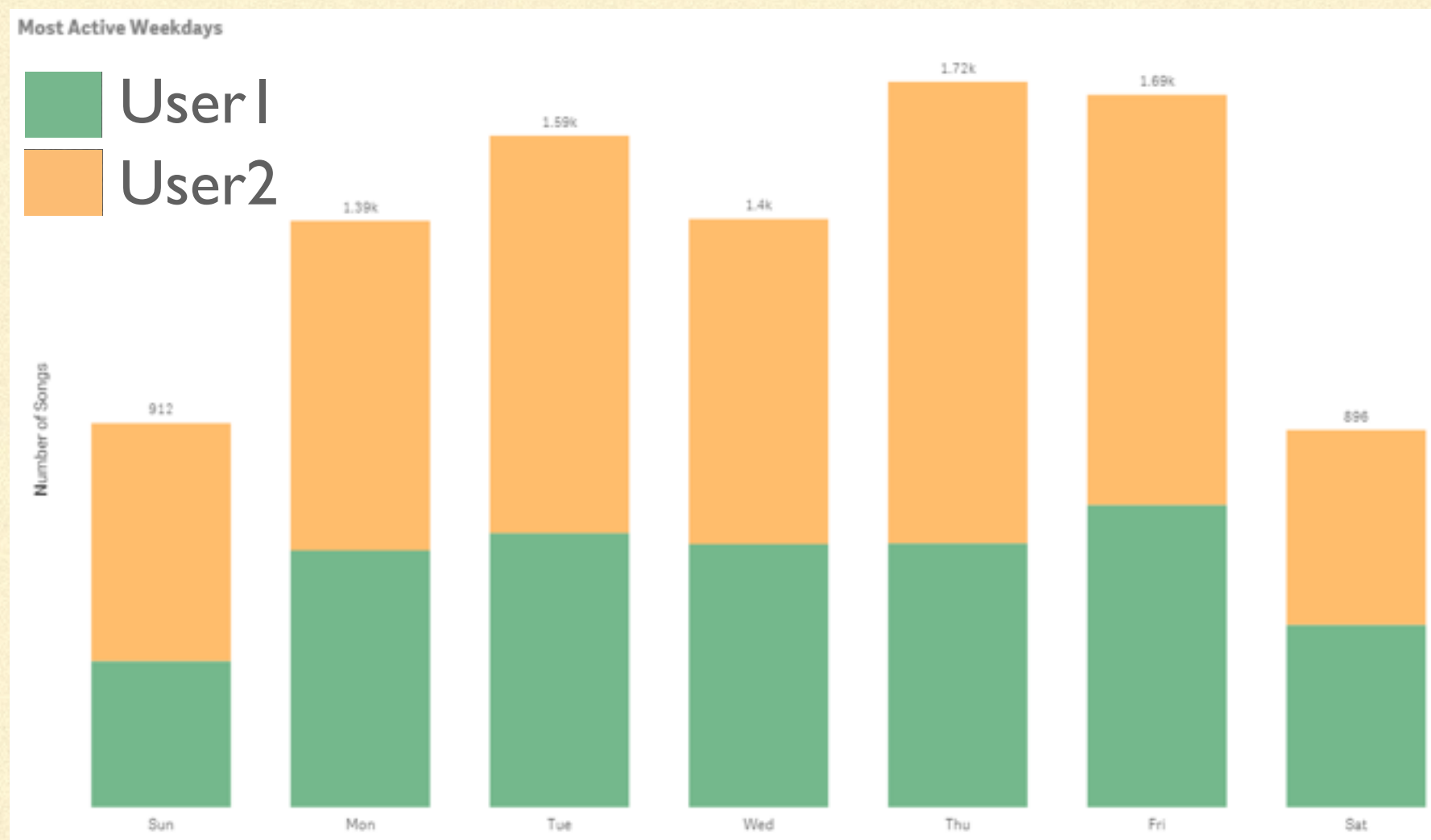
DATASET

- Listening history of **User1** and **User2** during 2015-06-30 and 2015-11-01, about 9.61k records in total
 - Contents
 - Time, devices and locations
 - Sources, artists, albums, songs and scores of songs
-

USER I

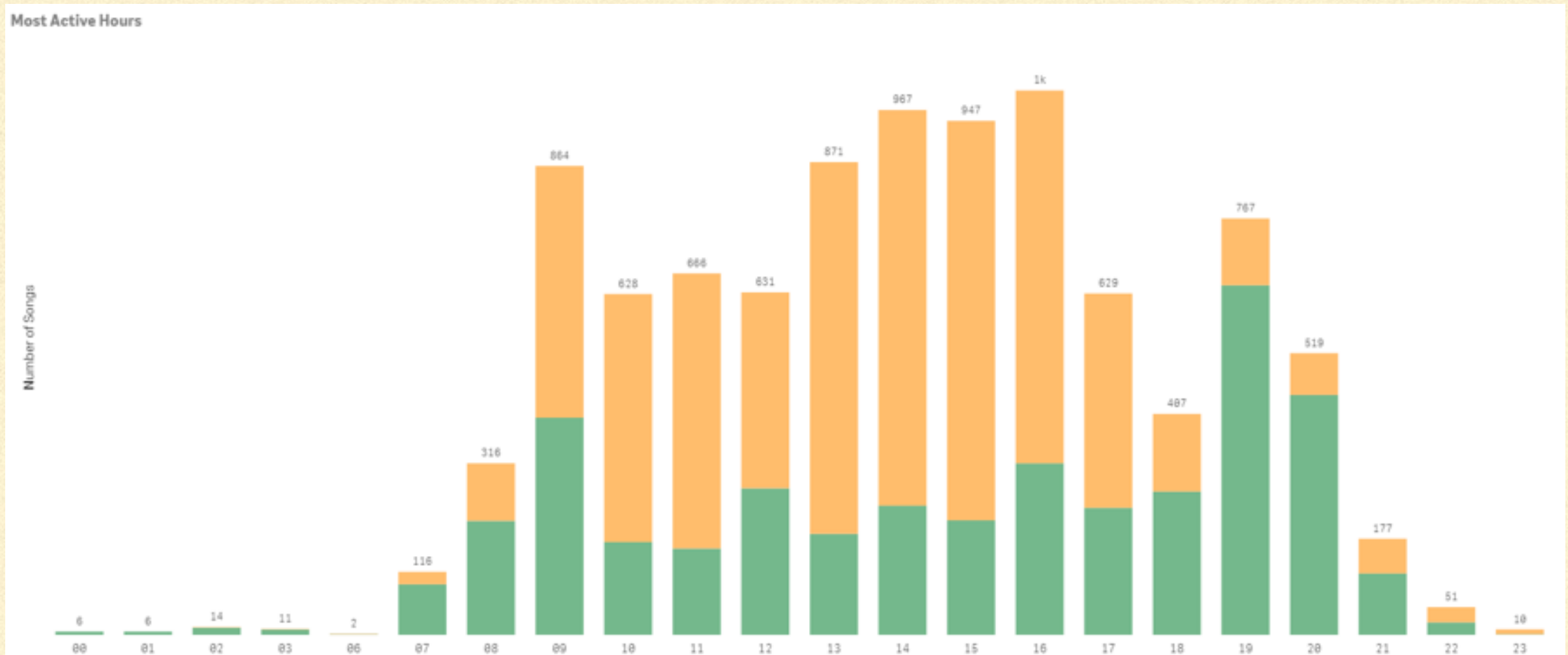
WHEN

- Both User1 and User2 like listening on workdays



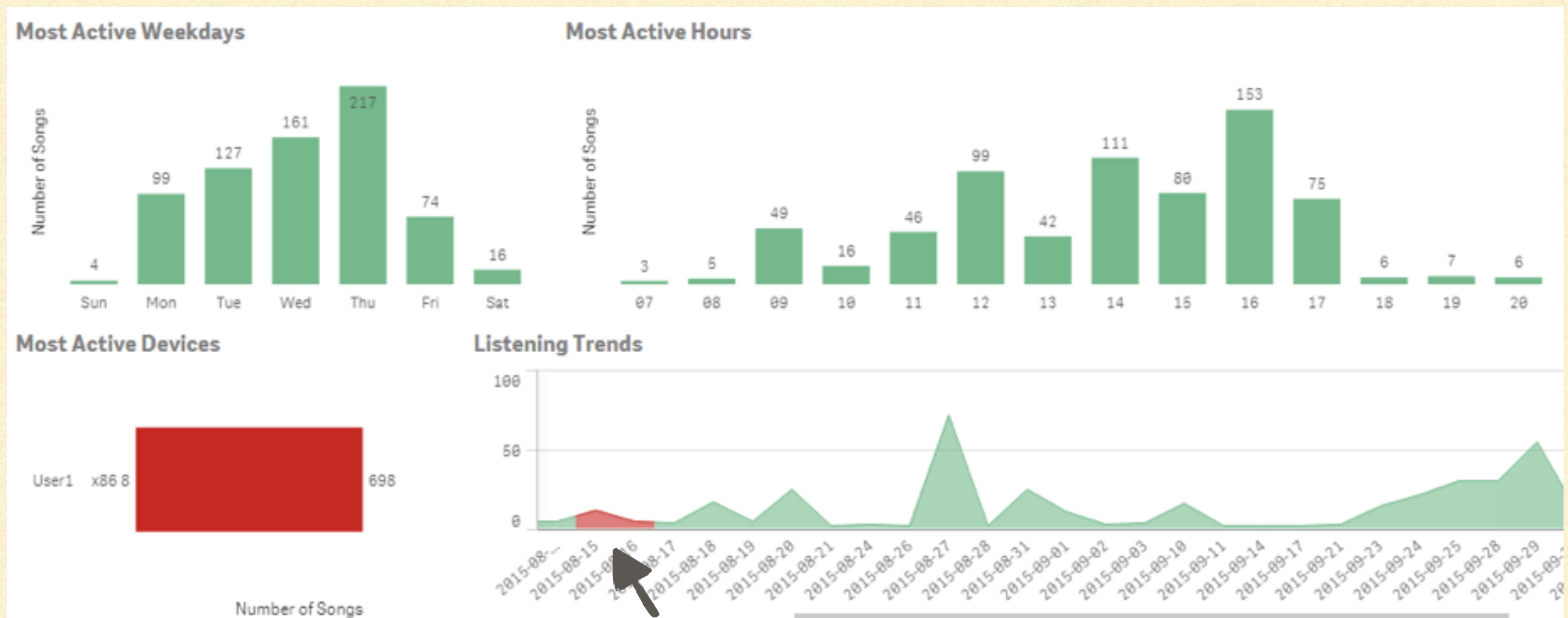
WHEN

- User1 listens more music on the way to work or after coming back to home; while User2 like listening during work



USER1 - HOW

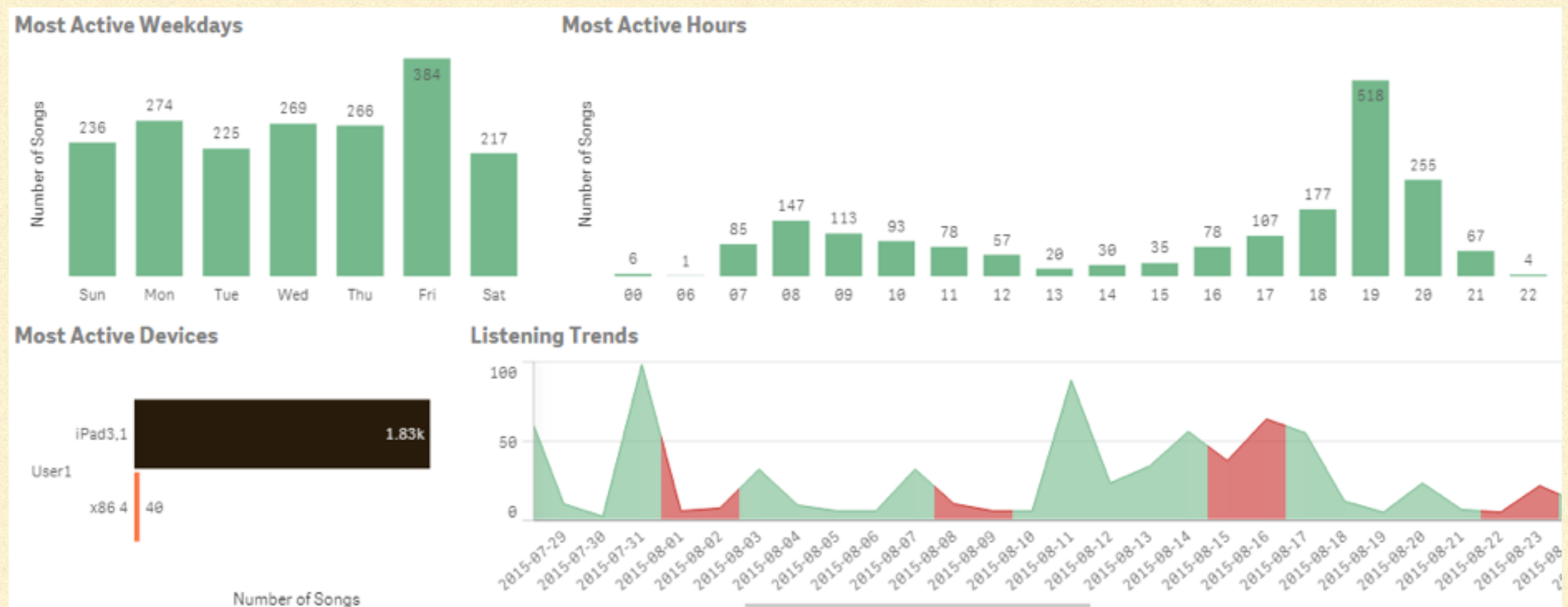
- x86 8 is User1's working device



Red represents weekends

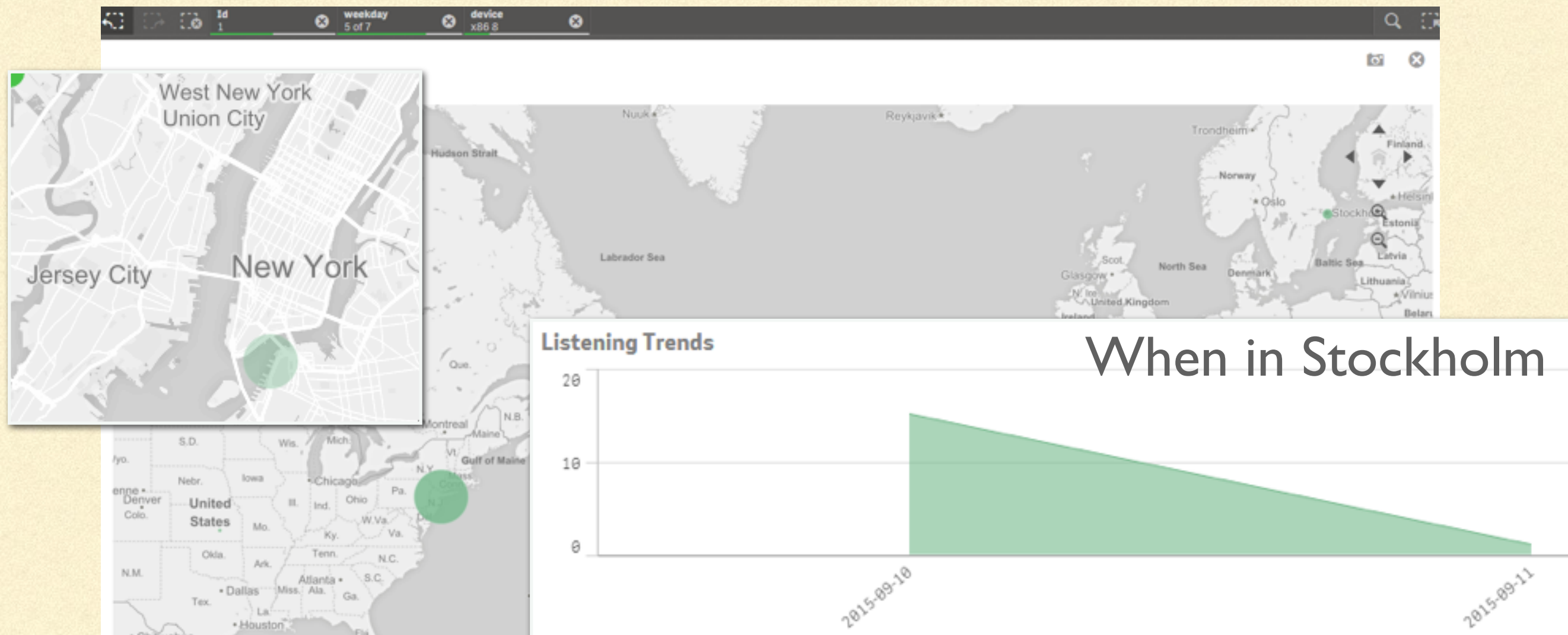
USER1 - HOW

- iPad3.1 and x86 4 are User1's entertainment devices



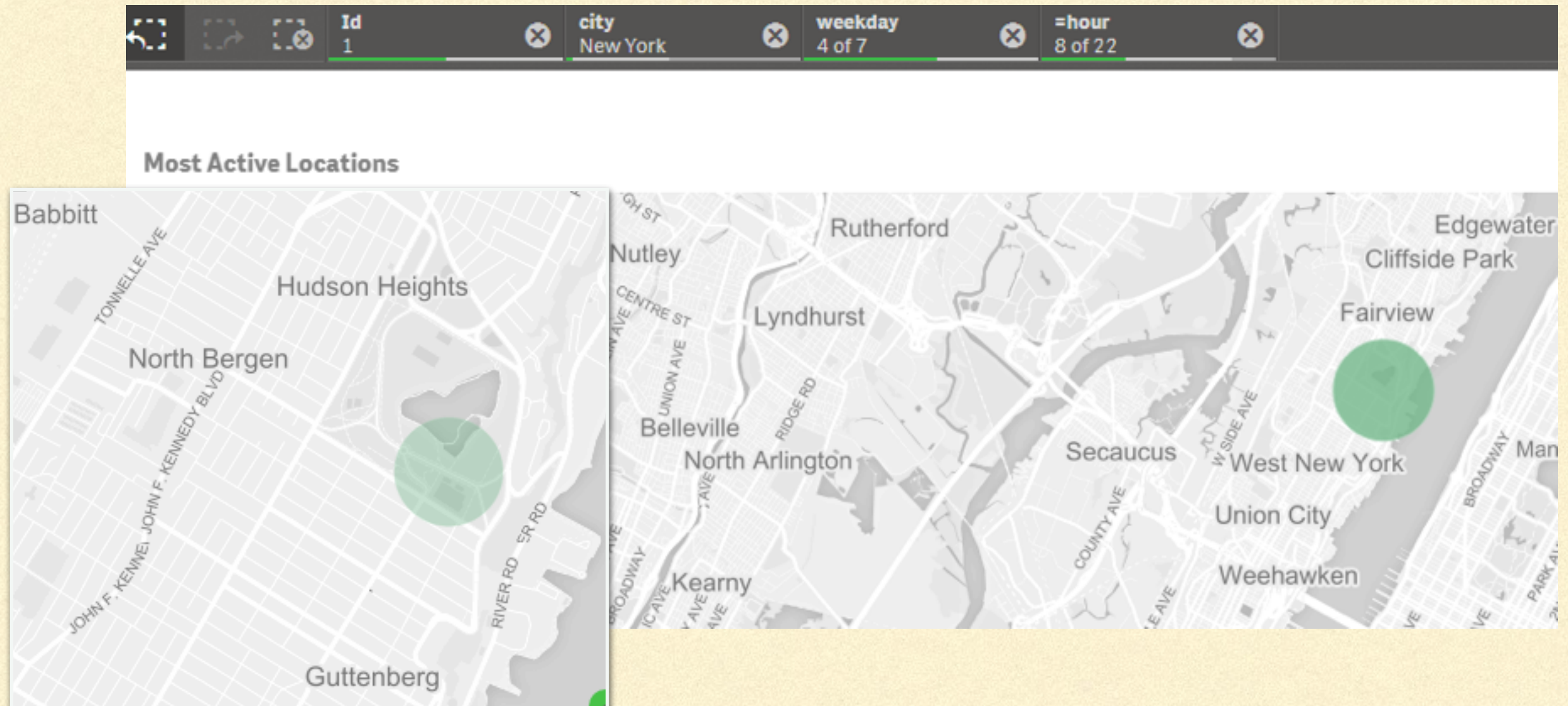
USER I - WHERE

- User I's working places: New York and Stockholm



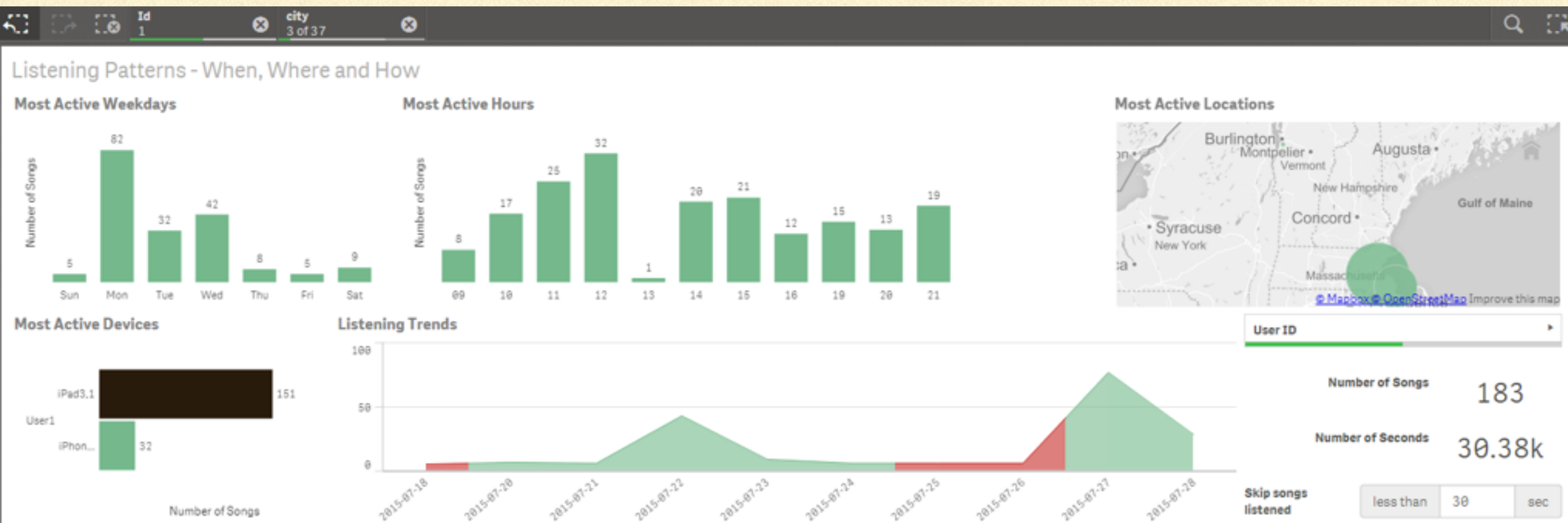
USER I - WHERE

- User I needs to take bus to work



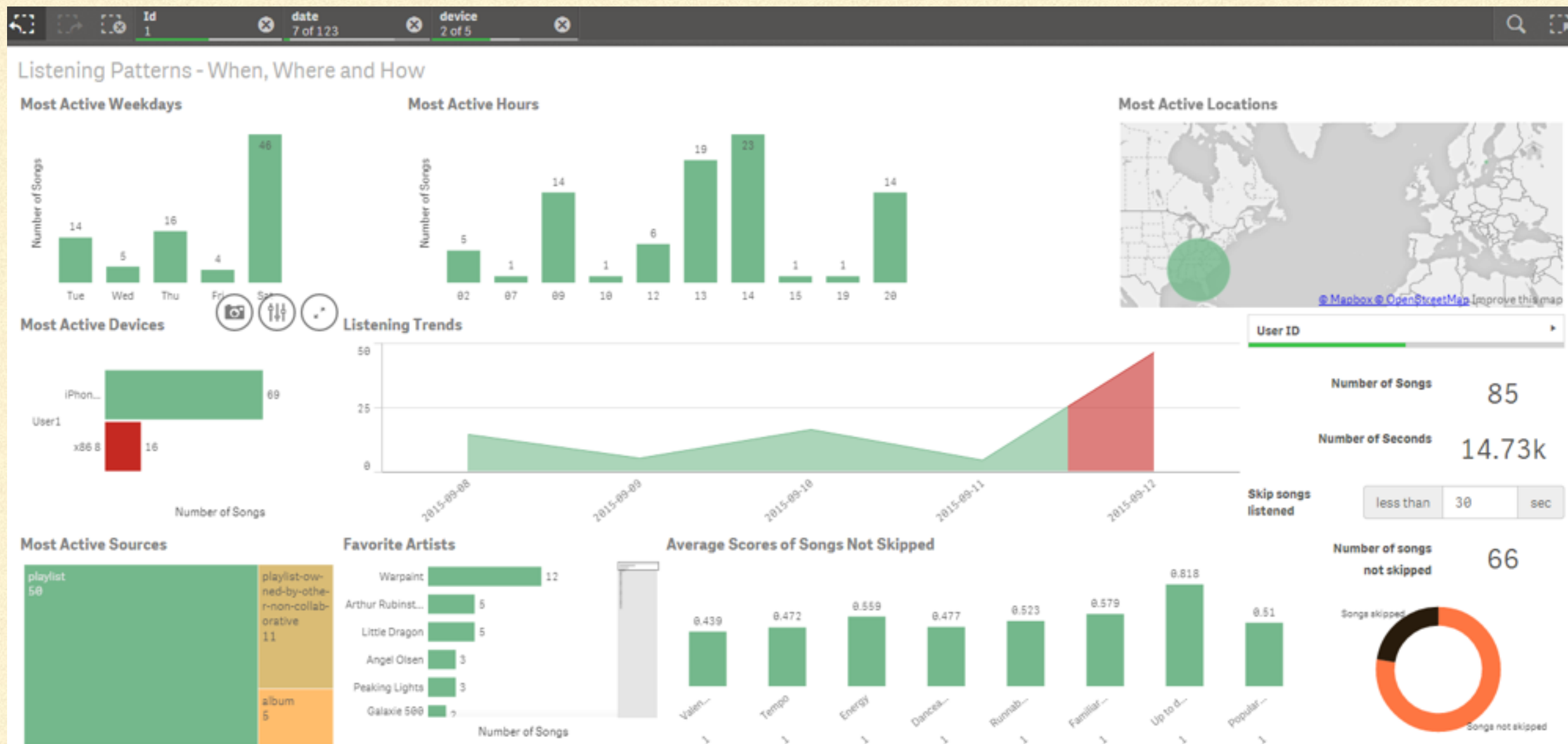
USER I - WHERE

- User I might travel during 2015-07-18 and 2015-07-28
 - Working device x86 8 was not used
 - Itinerary: Duxbury -> Shelburne -> Boston



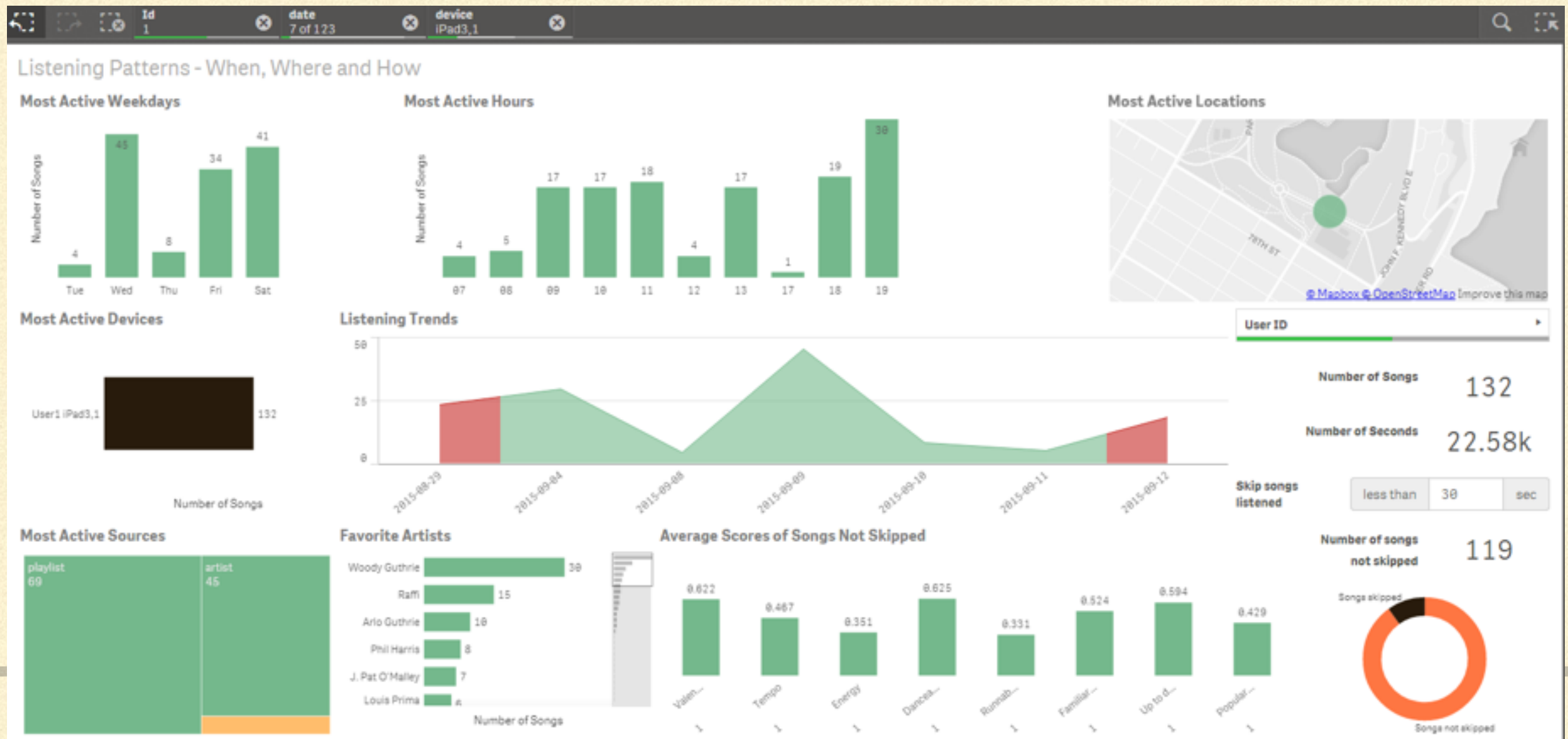
USER1 - WHERE

- User1 might bring his/her iPhone and x86 8 to Stockholm during 2015-09-8 and 2015-09-12



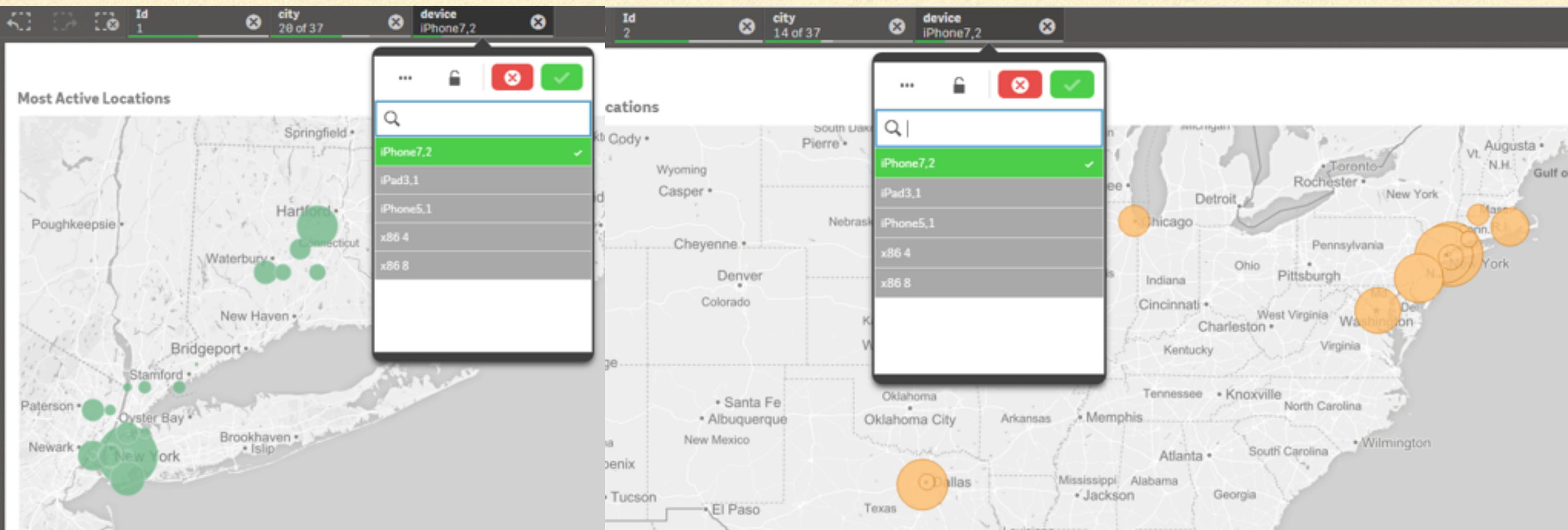
USER I - WHERE

- User I's Spotify account might be shared
 - During 2015-09-8 and 2015-09-12, iPad3 might be used by another person
 - The styles of musics listened via iPhone and via iPad are very different



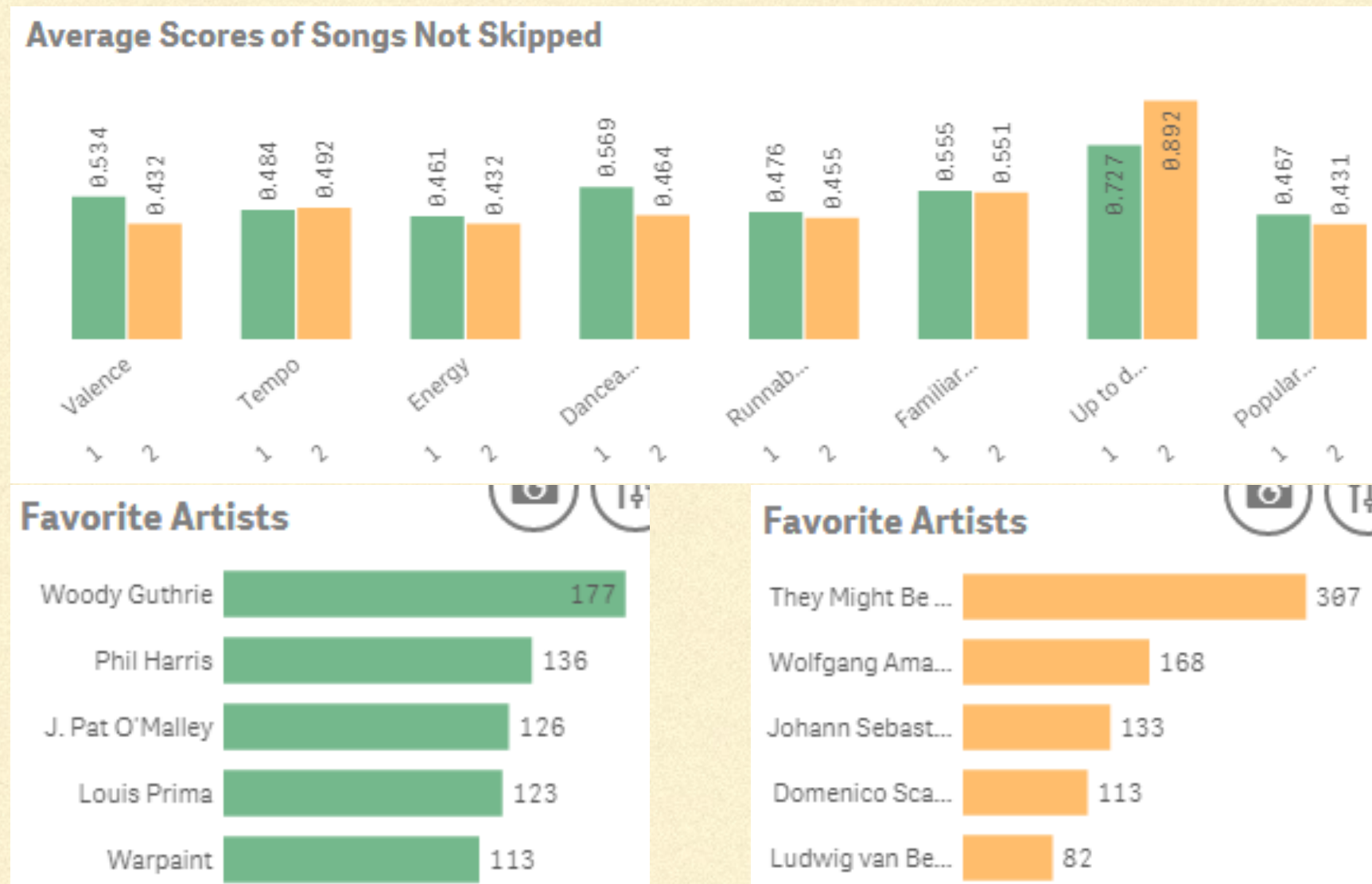
WHERE

- When having trips to places closing to New York, User1 only brought iPhone
- When traveling, User2 only brought iPhone



WHAT

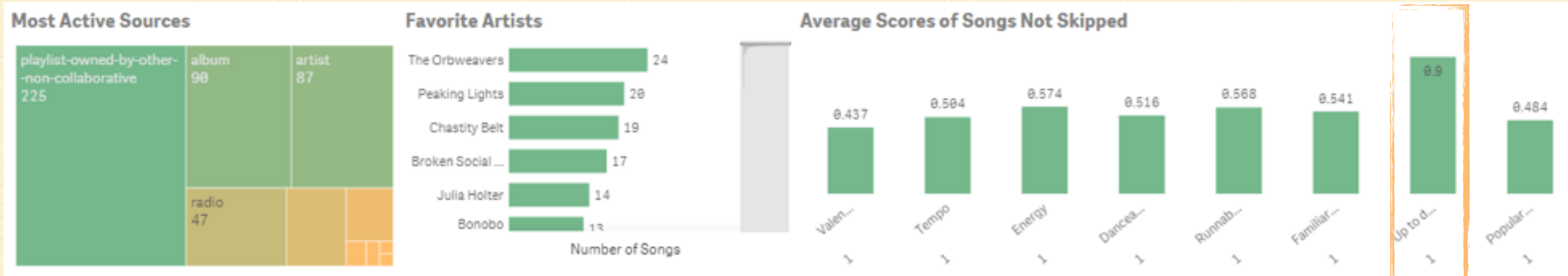
- User 1 likes songs suitable for dancing, while User 2 likes some new songs



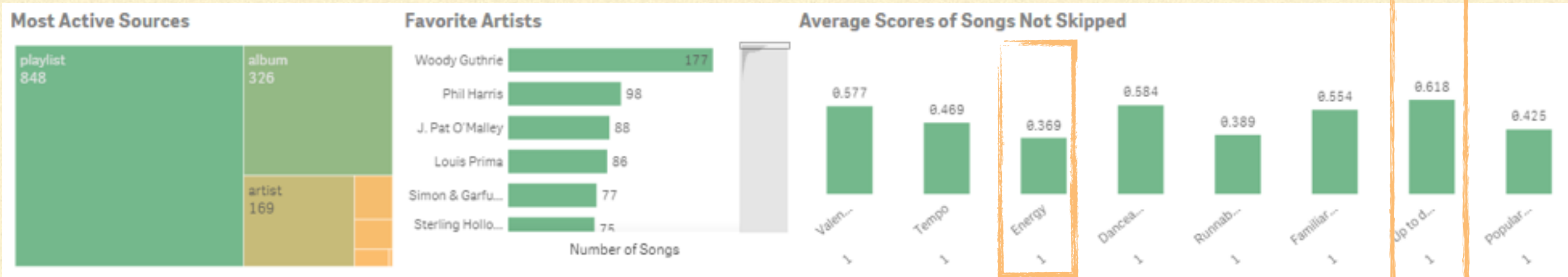
* Up to date score is the normalized year (1940-2015).

USER 1 - WHAT

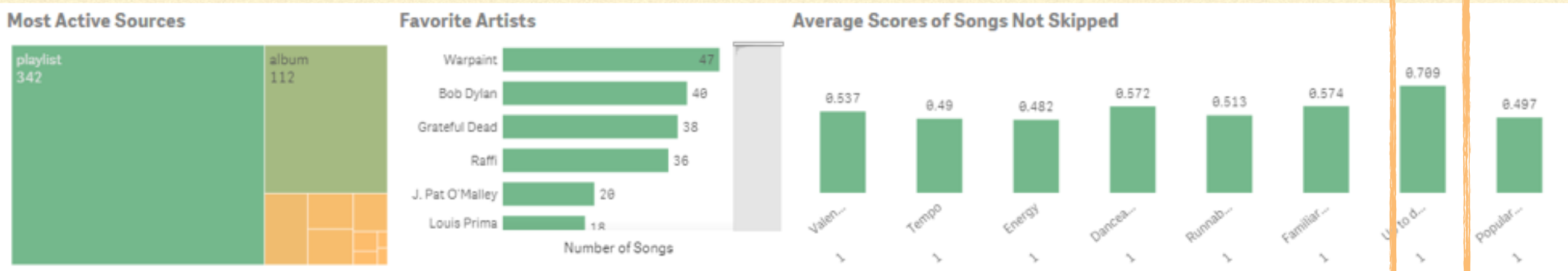
WORK



HOME



OUTDOOR RELAX

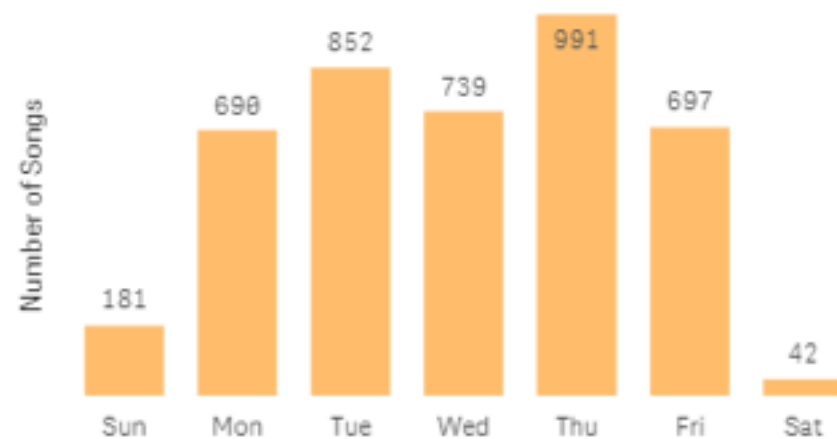


USER 2

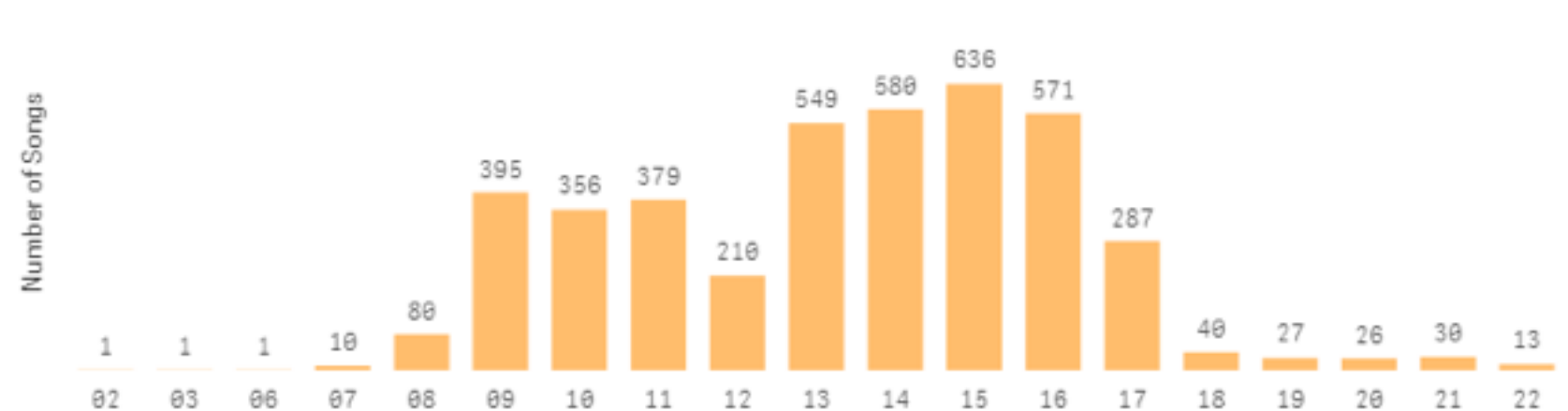
USER2 - HOW

- x86 8 is User2's working device, which is the same as User 1

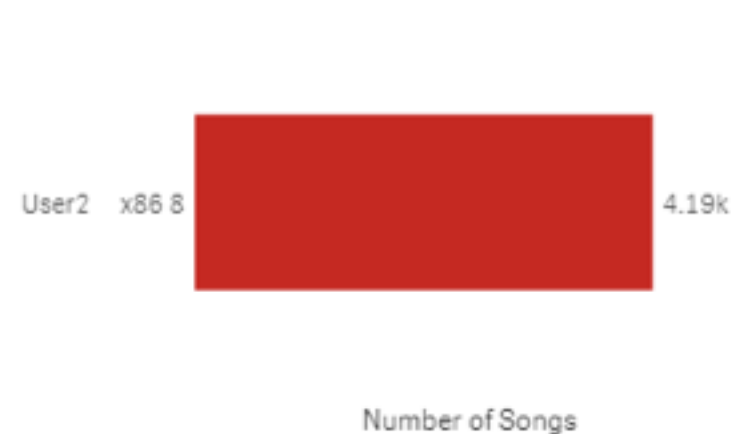
Most Active Weekdays



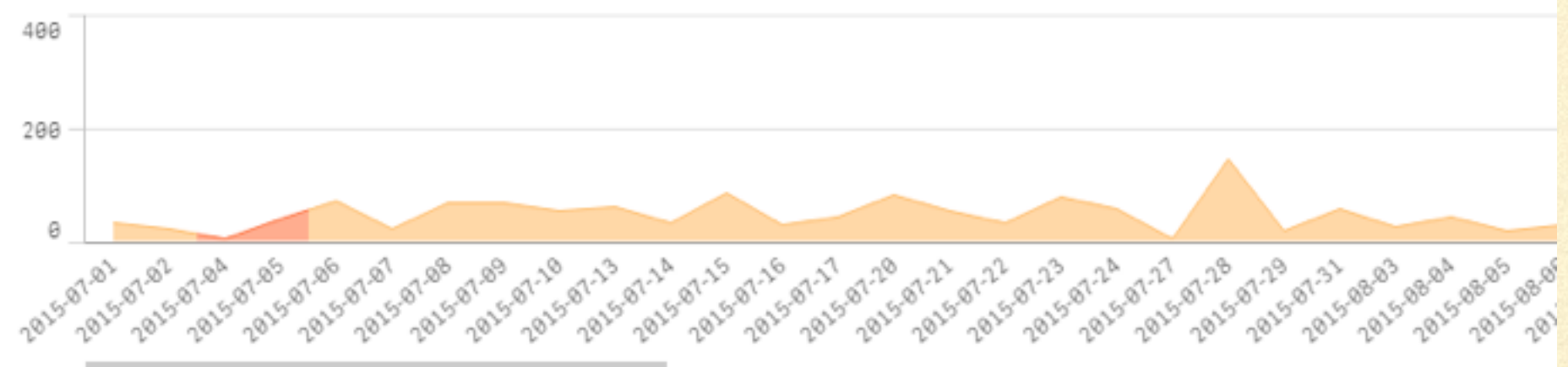
Most Active Hours



Most Active Devices

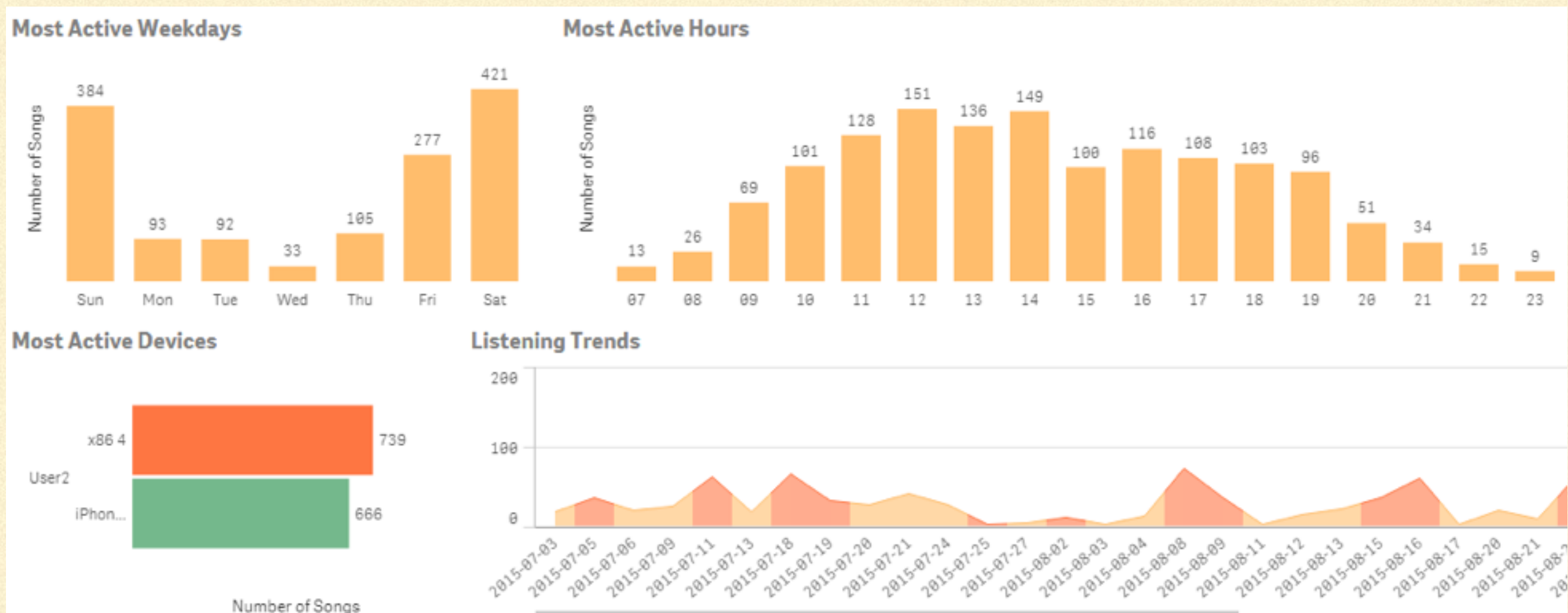


Listening Trends



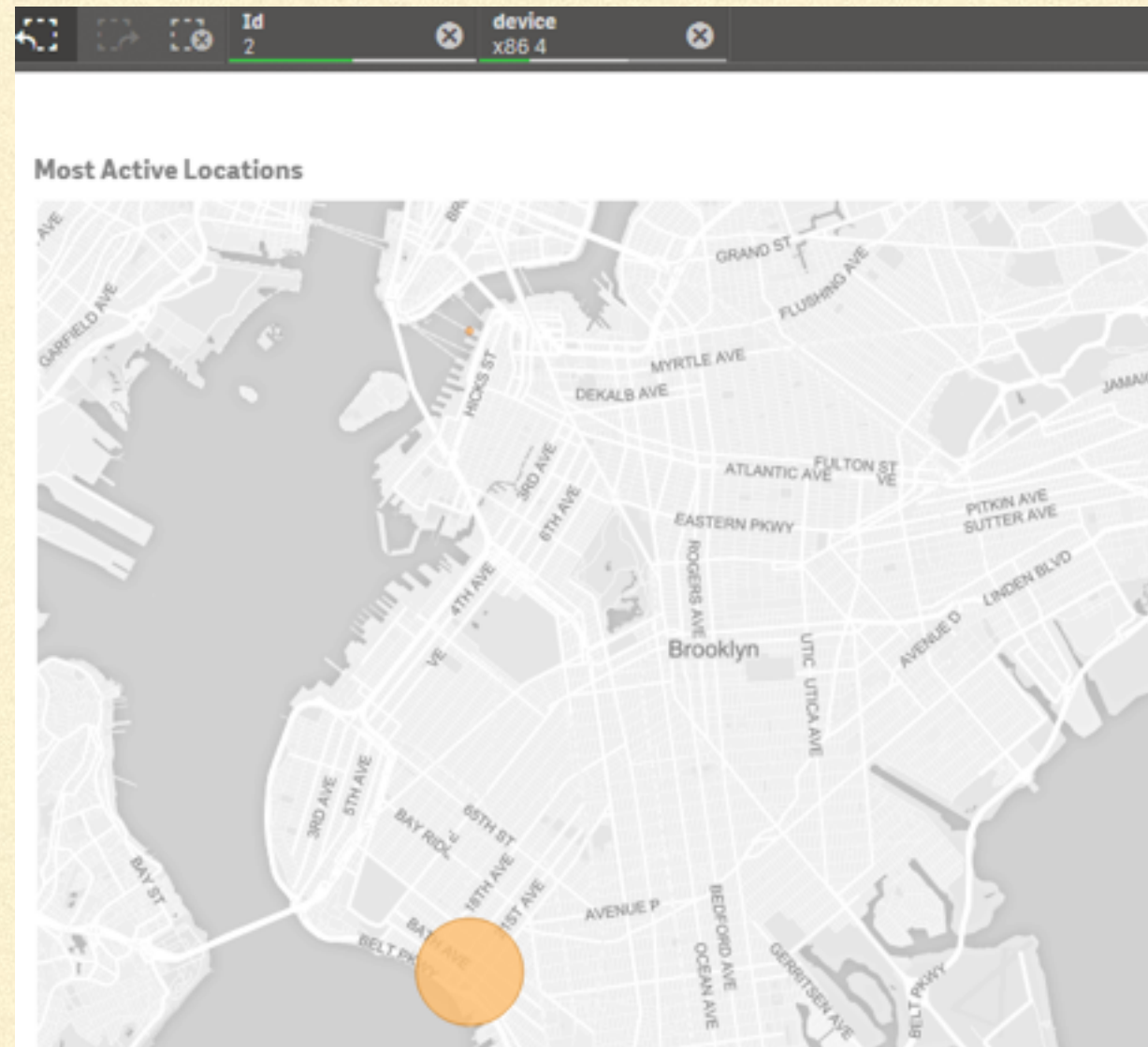
USER2 - HOW

- User2 has a x86 4 device (OS system) as personal laptop. He / she does not use Spotify on iPad



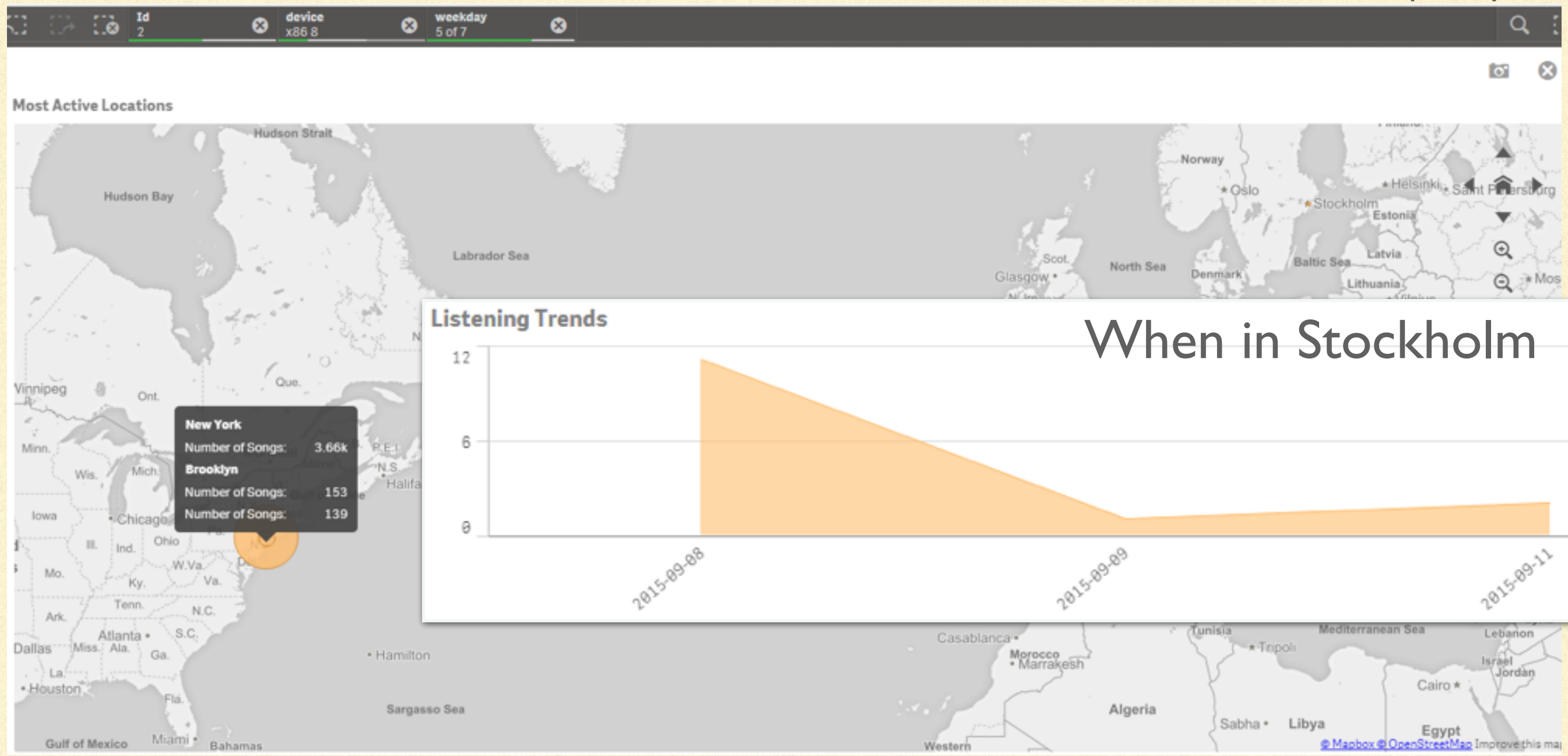
USER2 - WHERE

- User2 lives in Brooklyn



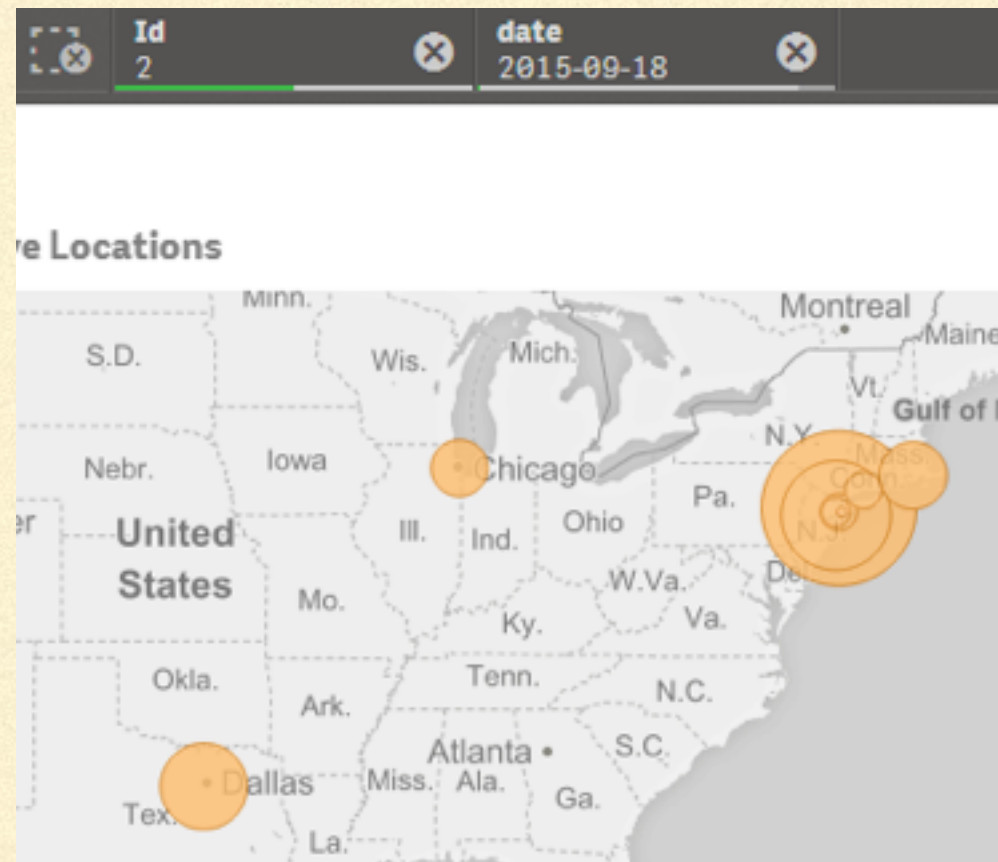
USER2 - WHERE

- User2 also works in New York and Stockholm, but he / she works from home more frequently



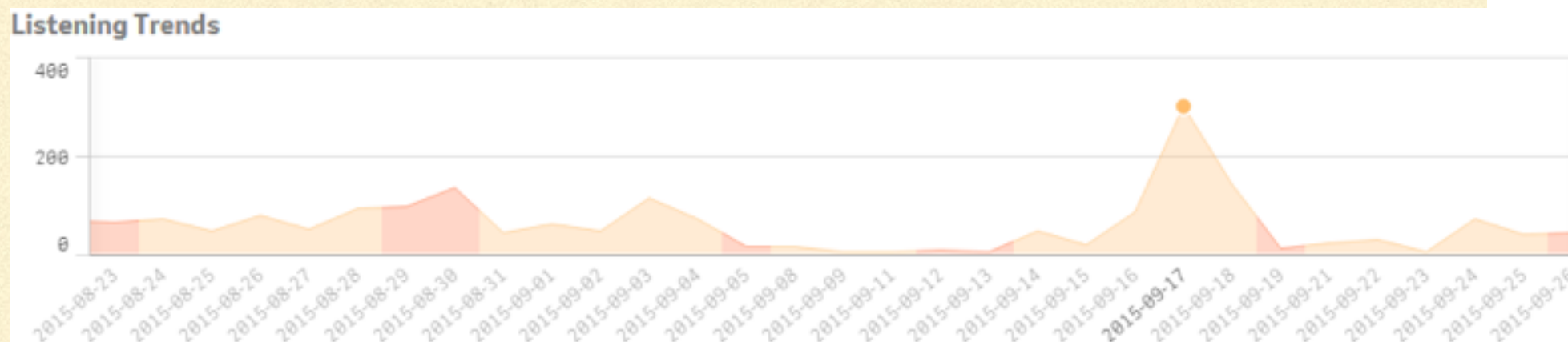
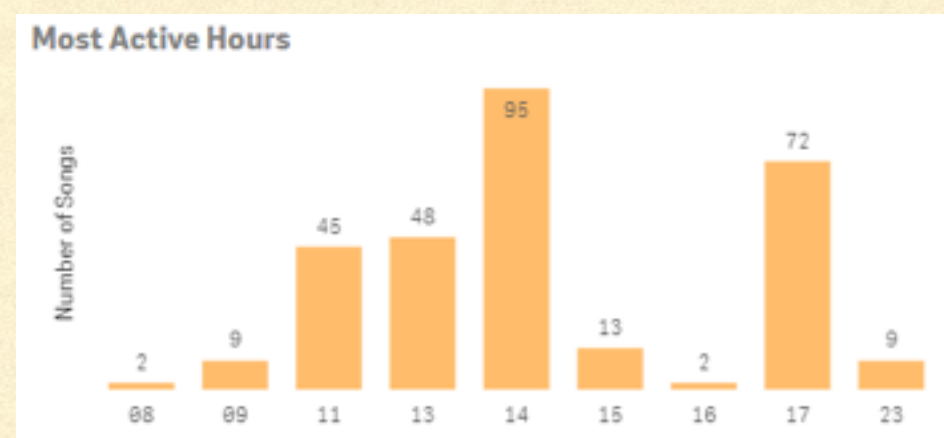
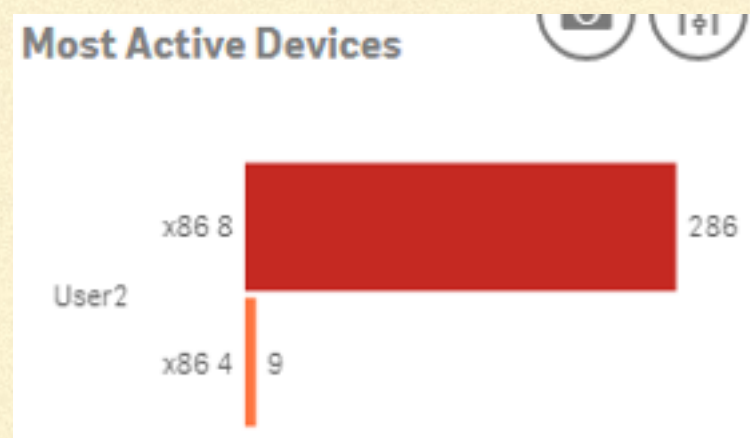
USER2 - WHERE

- User2 used Spotify at many locations on 2015-09-18
 - The user might use VPN
 - Other reasons?



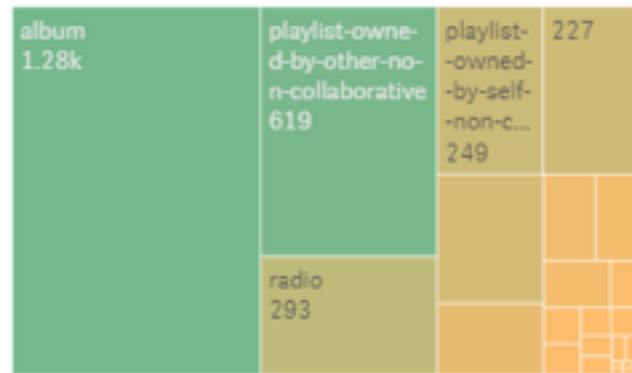
USER2 - WHAT

- The skipping rate is very high on 2015-09-17
 - User2 might test Spotify or check a lot of Spotify player embedded on websites



USER2 - WHAT

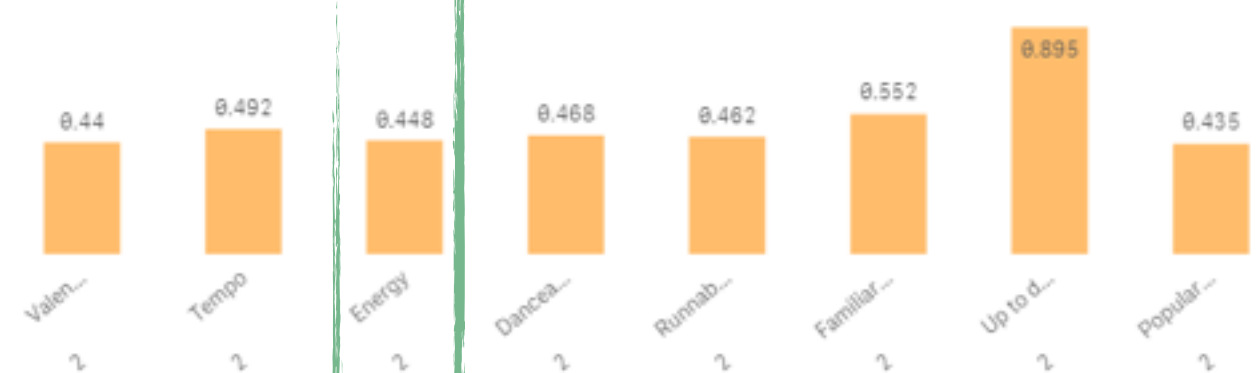
Most Active Sources



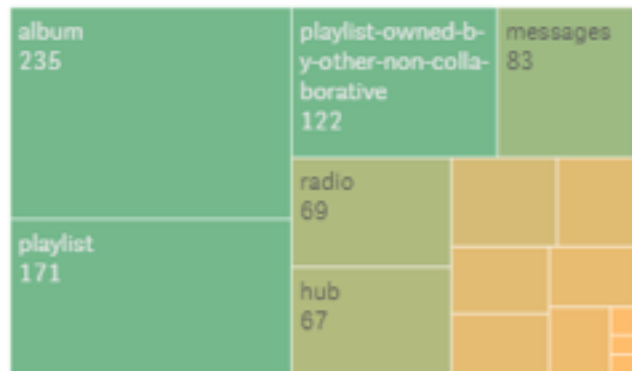
Favorite Artists



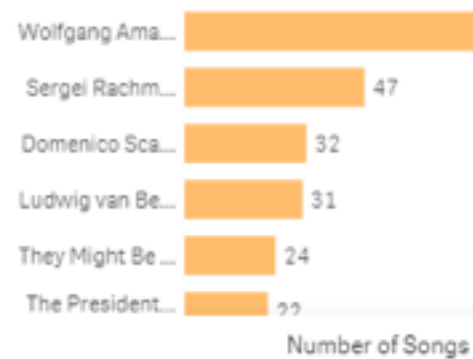
Average Scores of Songs Not Skipped



Most Active Sources



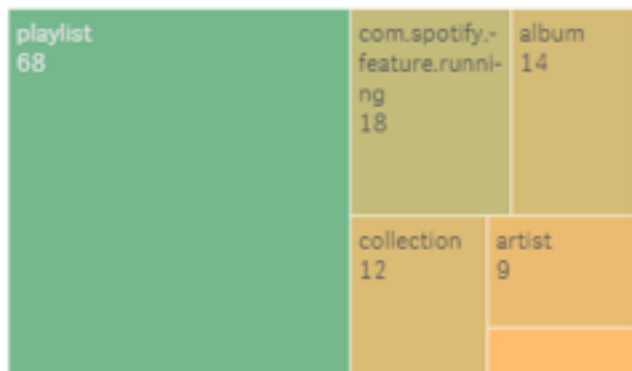
Favorite Artists



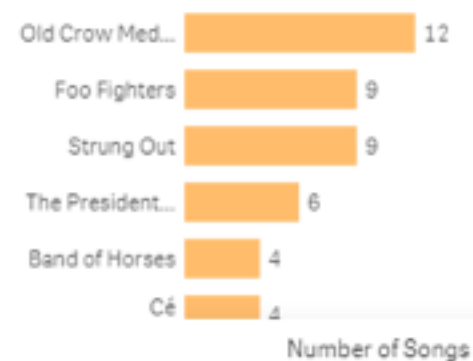
Average Scores of Songs Not Skipped



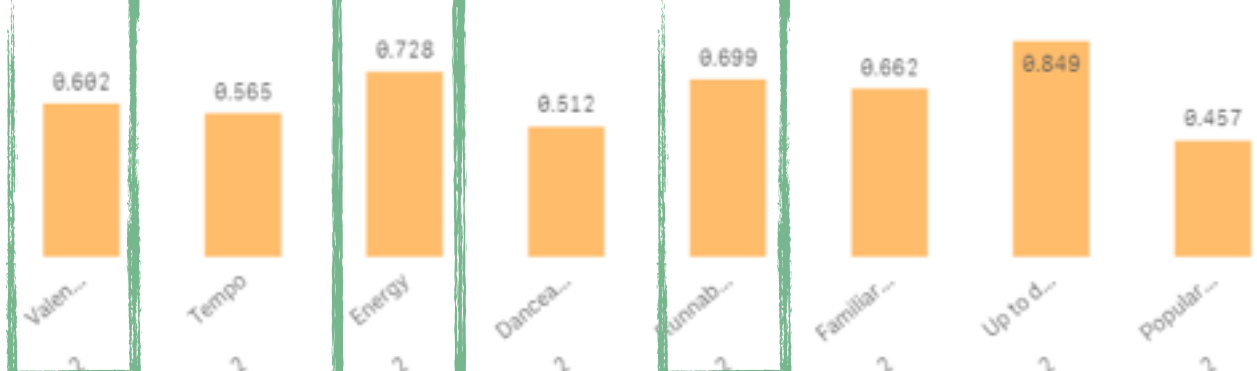
Most Active Sources



Favorite Artists



Average Scores of Songs Not Skipped



WORK

HOME

OUTDOOR RELAX

RECOMMENDATION & RISK

- Recommend different types of music based on users' activities
 - Type of music is decided by artists, scores of songs etc.
 - Activities includes work, home and outdoor
 - Activities can be drawn from time, devices and locations
 - Many people may use same user account
 - Location information may be incorrect
-

FURTHER STEPS

- Obtain genre of each tracks and analyze favorite genre of each user
 - Text mining of album name and track name
 - Understand the meaning of sources field and analyze it
 - Get more accurate longitude and latitude data
 - Change city to county or other detailed address information
-

THE END
