

FOURSQUARE AWARE

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Recommendation systems are a prominent application of **big data**, used everywhere from targeted advertising to online dating. We clustered users to make these systems more efficient and give users easy access to new venues based on their individual interests.

We first receive a user's **data** and **category preference**, and aggregate their check-in counts by venue category.

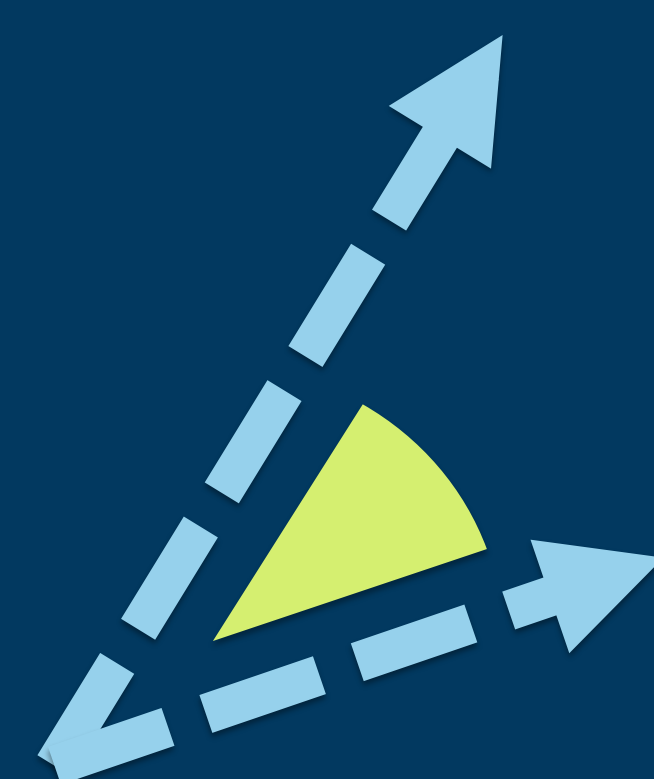


We then group the user into a **cluster of similar users**, reducing the user pool we must compare to.

Finally, we compare the **most similar user's** top venue choices against the input user, and recommend the top 3 unvisited venues with the highest score.

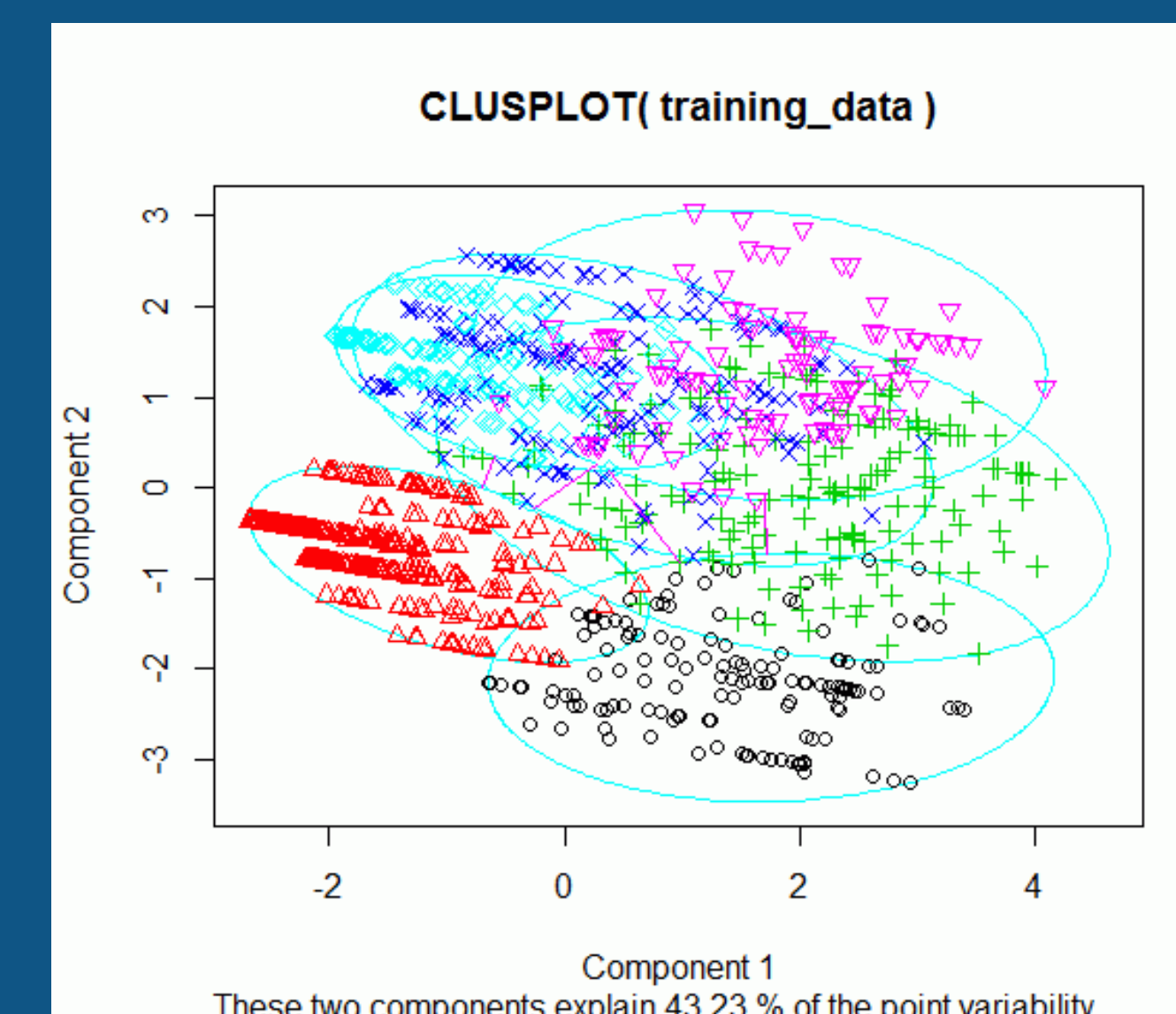


Our dataset was a collection of **227,428** Foursquare check-ins from **1,083** users in New York City over **10** months.



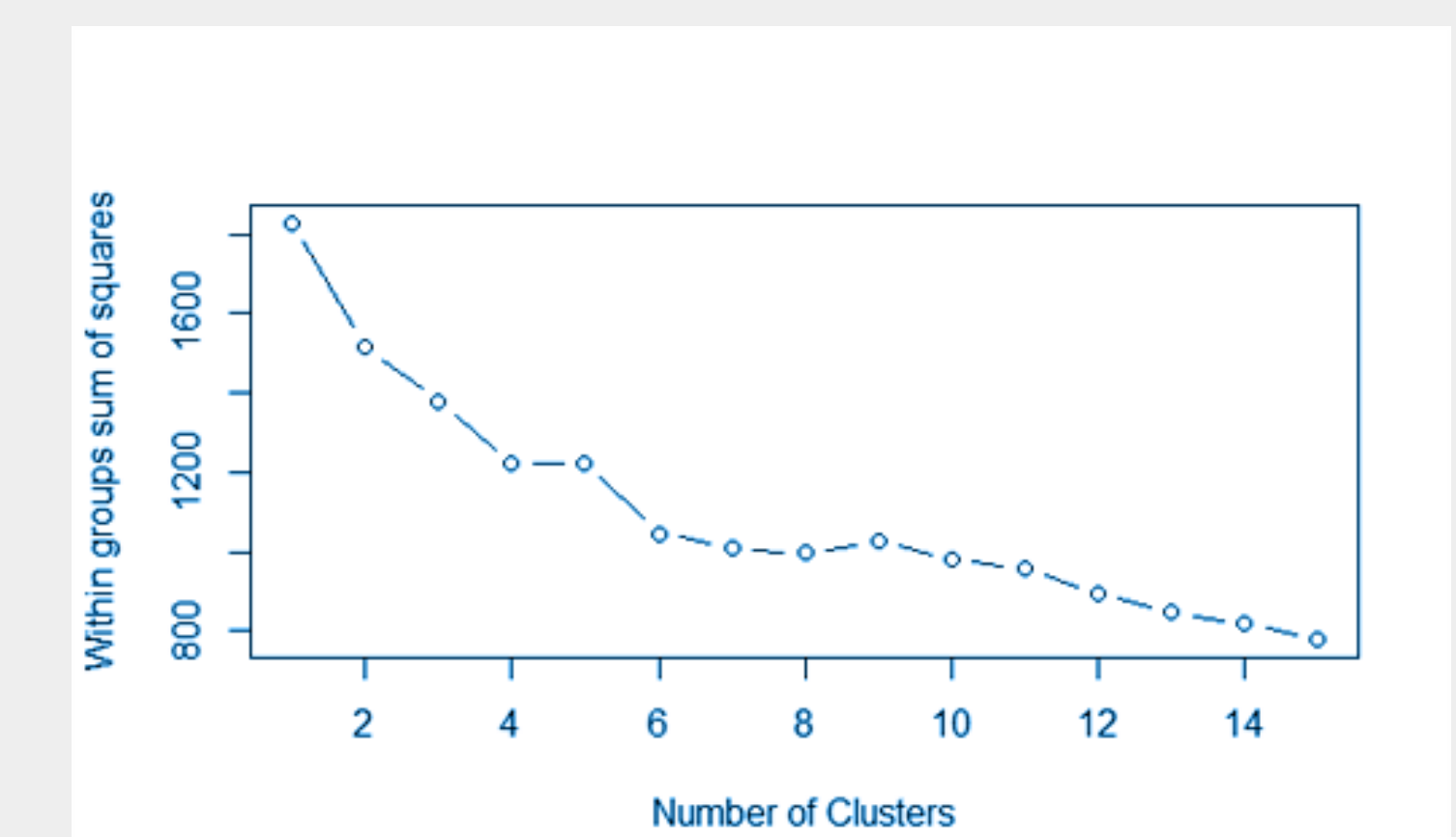
We used **cosine similarity** to find similar users after considering a number of different metrics and optimizing system performance.

Preprocessing: Using Foursquare's API, we **grouped** check-ins to individual venues by category for easier comparisons, and **scaled** these values down to indicate relative frequency.

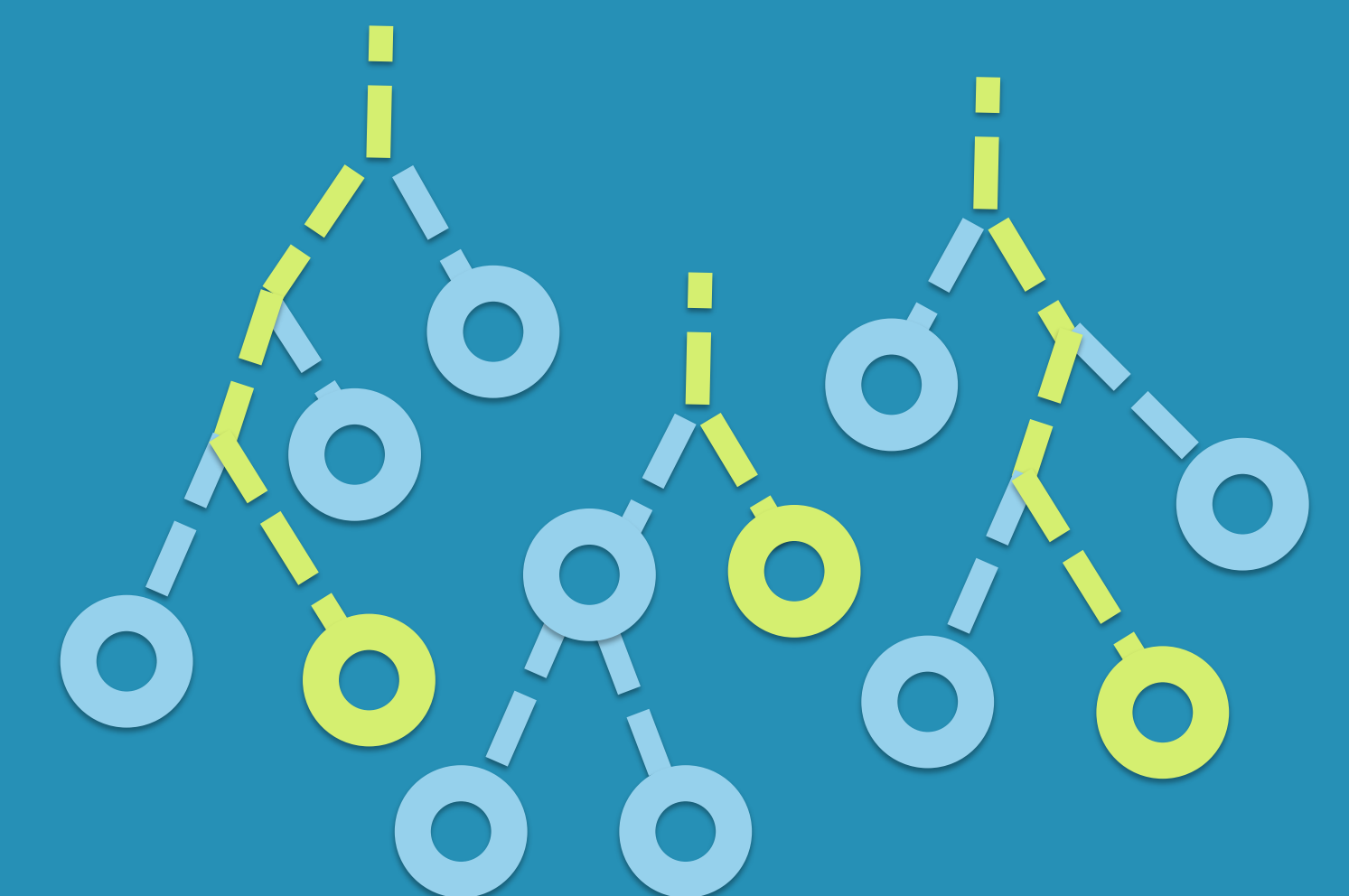


PCA after running k-means w/ 6 centers

Finding the optimal number of clusters (vs. sum of squares error)



We fit querying users into clusters using a **decision tree** generated during our clustering.



Our system successfully returns venue recommendations based on similar users' preferences. In the future, this system can be extended by factoring in **timestamp data** to recommend trending locations, or finding new friends at suggested venues.