



RECOMMENDATION SYSTEM WITH MINIMAL IDENTIFIABLE FEATURES

Exploring the possible feature / stretch goal



Initial_parameters.json



student_all_interests.csv

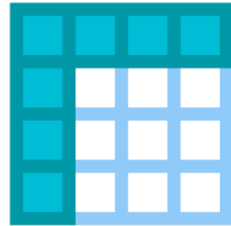


clubs_similarity.csv

Data and Computational workflow



pipeline.py



populate_dataset.py



club_similarity.py



recommend_clubs.py



AllUBCOClubs.json



clubs_similarity_index.csv



similar_students.csv



club_recommendations.csv



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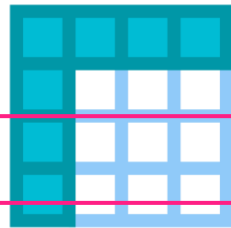


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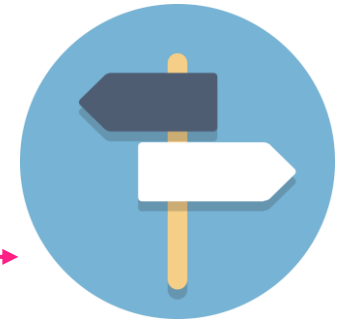
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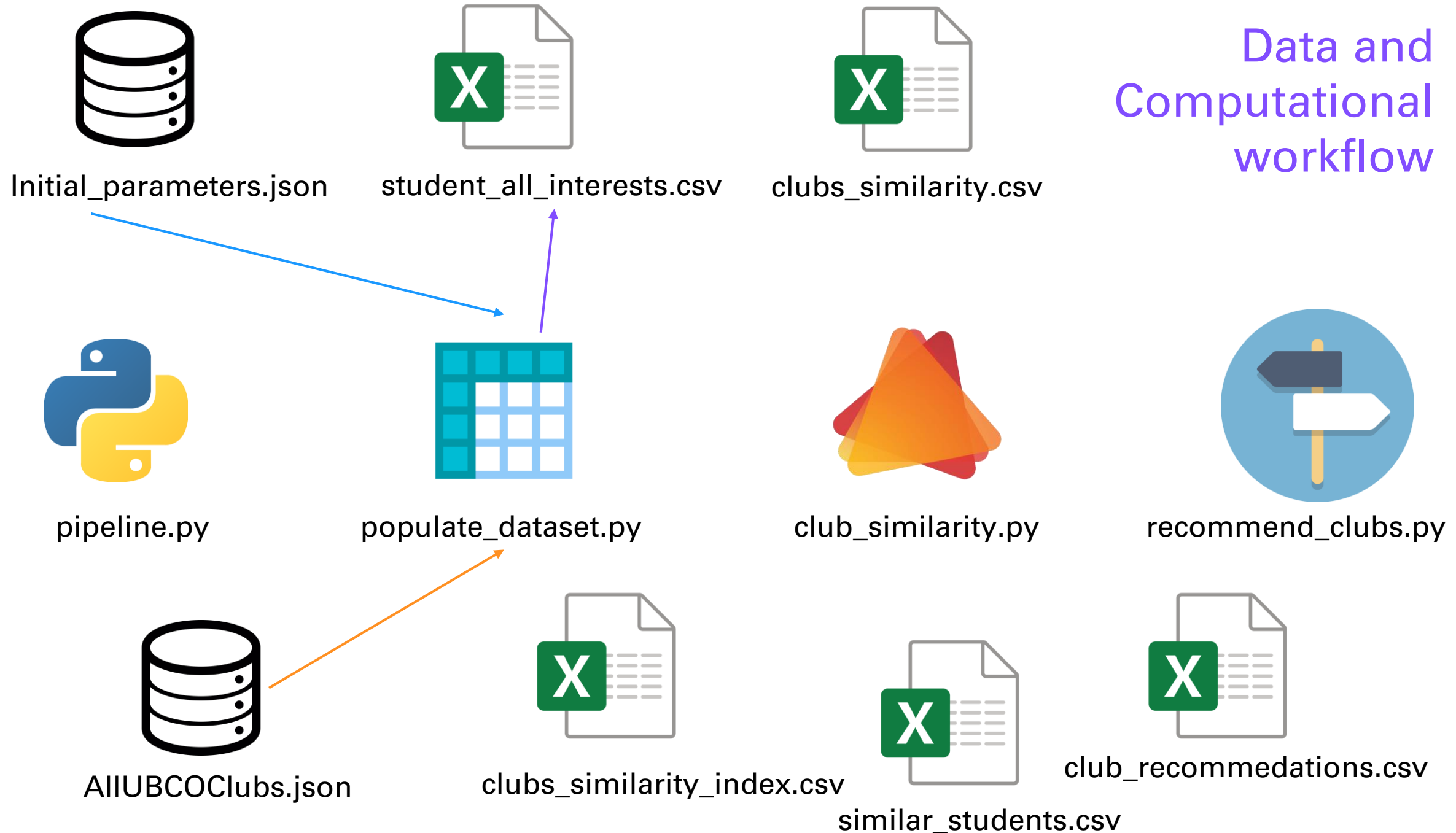


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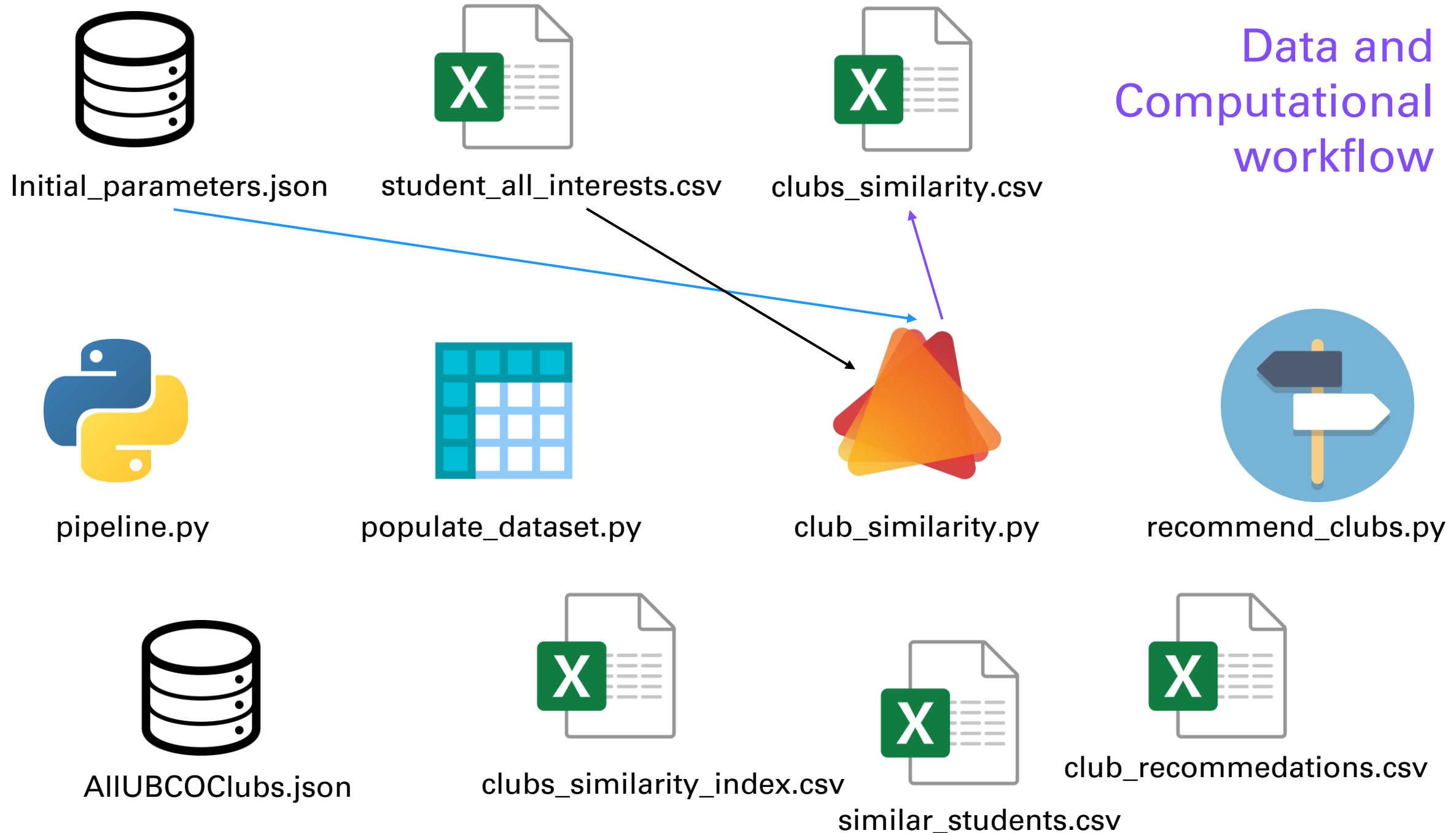


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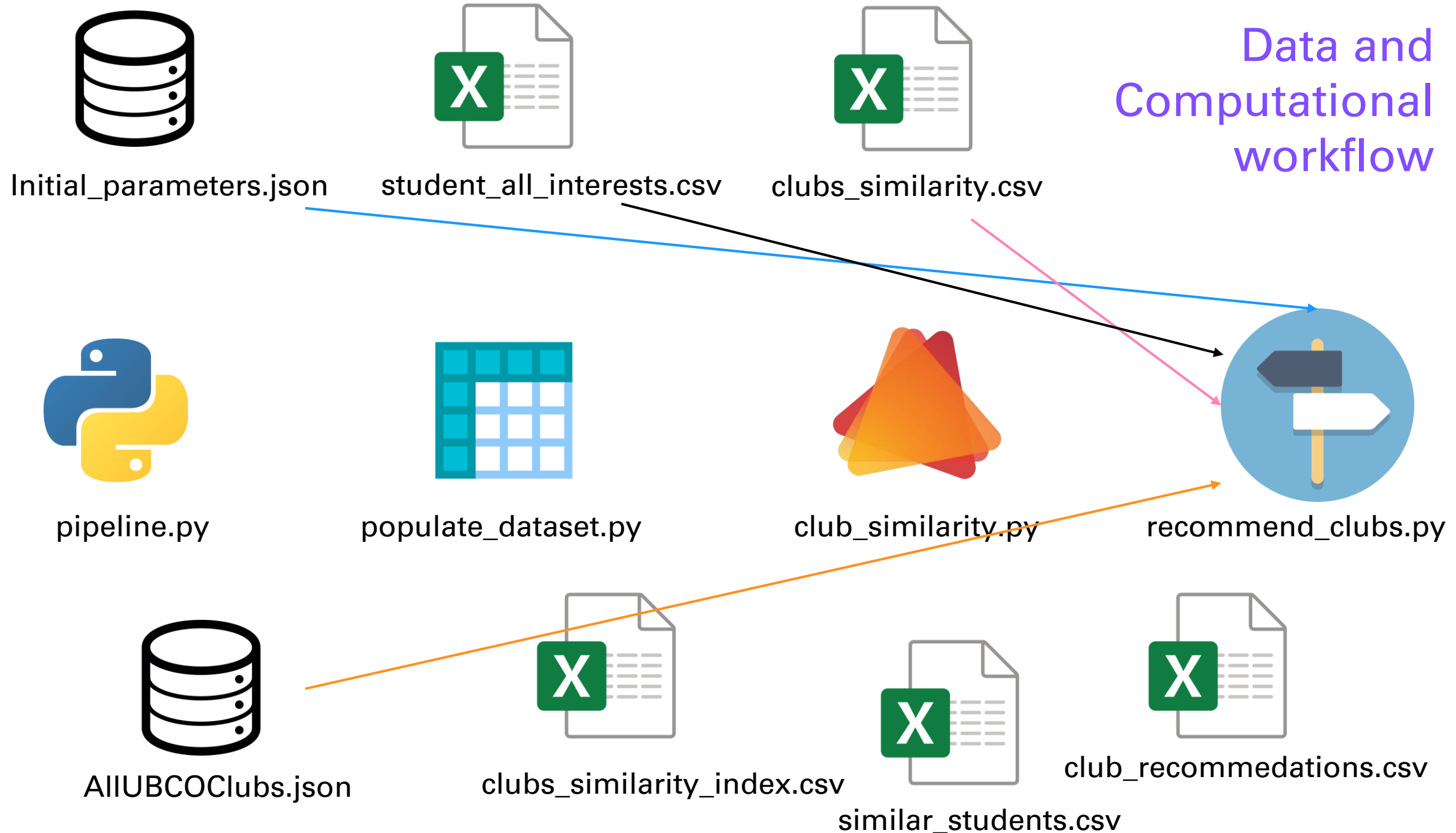
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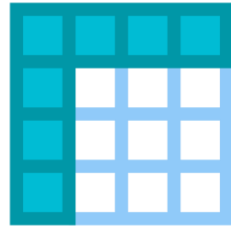


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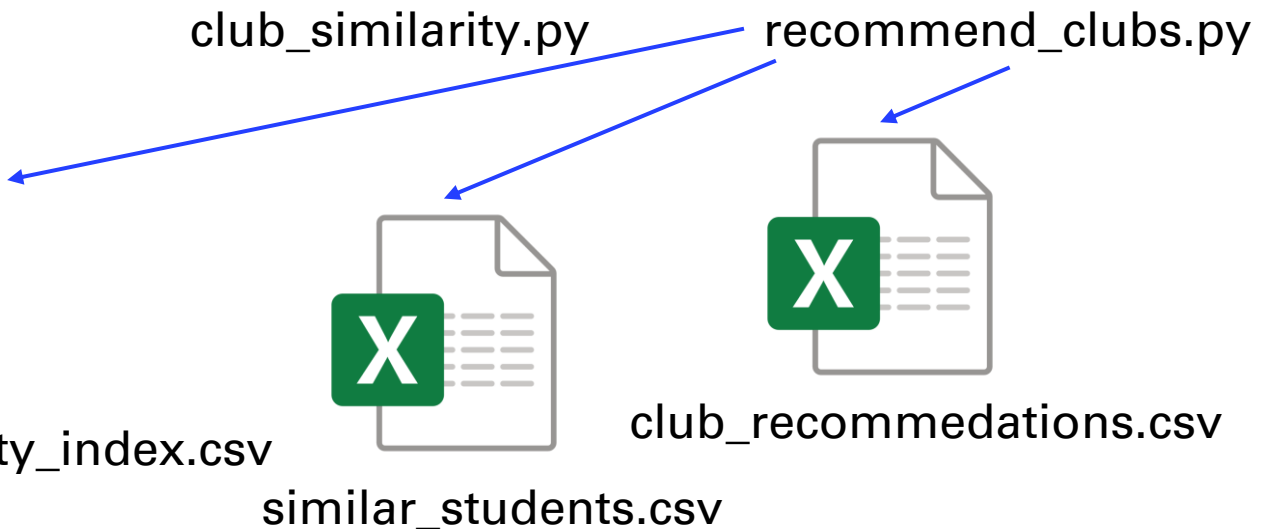
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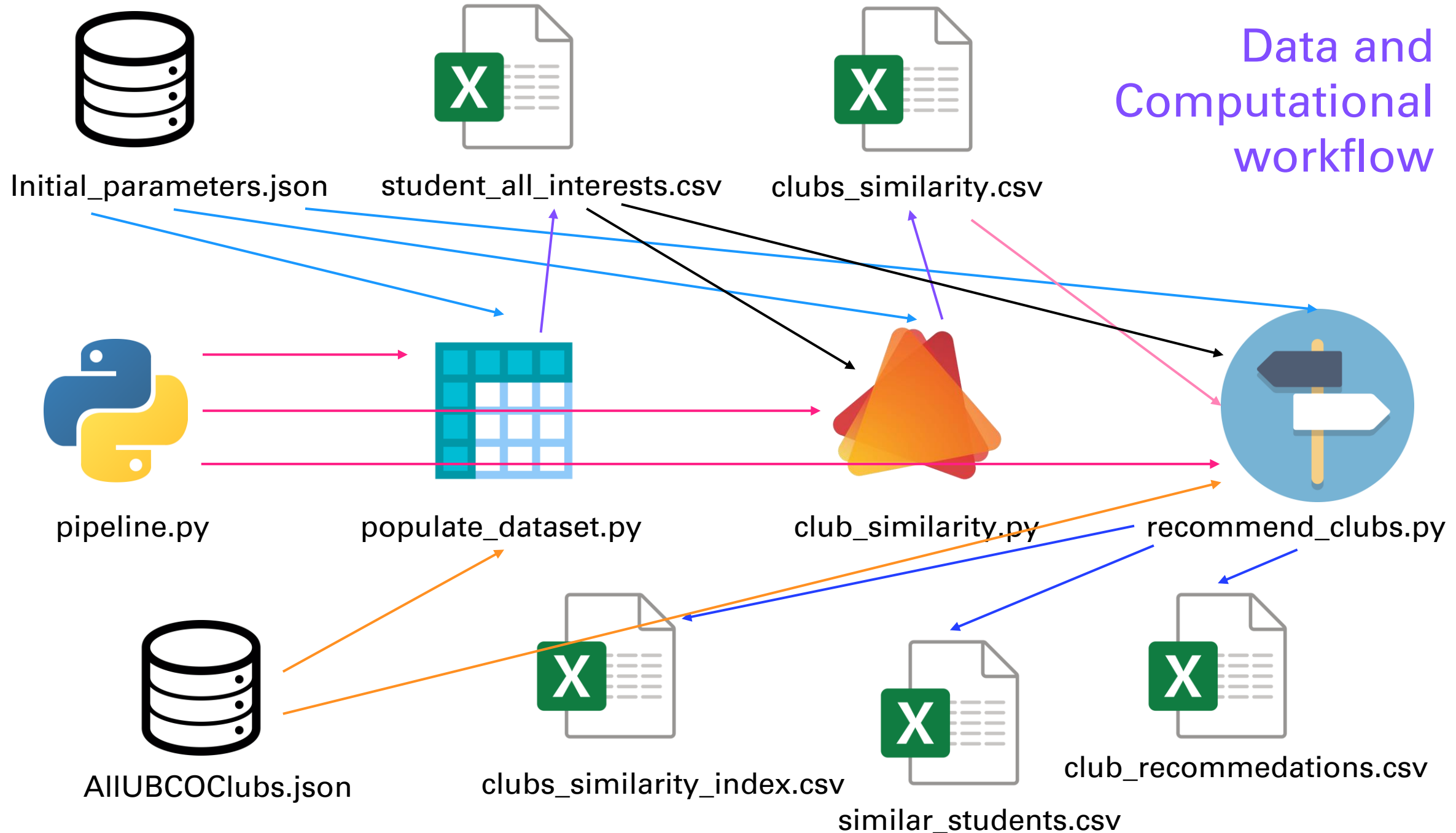
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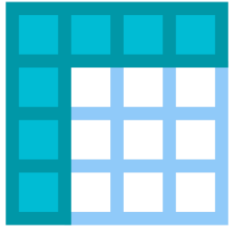


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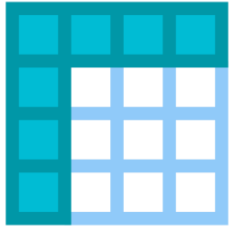




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Objective: Creates a random dataset for n students

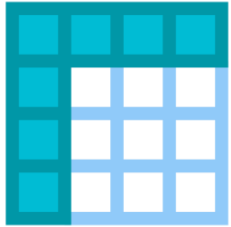
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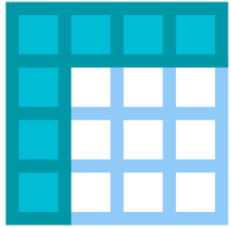
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- Generates random event interests [still under development]



club_similarity.py

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- Computes the intersection of a pair of club lists from two different students
- Generates club_similarity.csv for better interpretability of the recommendation system



recommend_clubs.py

Objective: Generates a .csv of club recommendations

- Finds similarity indices between a pair of different students (how many clubs are similar?)
 - Stored as clubs_similarity_index.csv
- Sorts these values for each student to find most similar students
 - Stored as similar_students.csv

- Computes importance of a similar user
 - Importance = `similarity_index(this_user, other_similar_user)`



`recommend_clubs.py`



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- Normalise the weights for each club
 - Store them from highest weight to lowest weight (i.e. most recommended to least recommended)
 - Generates club_recommendations.csv

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- Generates a **unique order of all clubs** and not just a subset of recommended clubs. This can be used as a native club-view order for each student (each student sees all clubs but in a unique order, personalized to them)



Areas of improvement

- For better personalized lists, **more data** is needed
 - Women in engineering
 - Indonesian students of Okanagan
 - African Caribbean Students Club
 - Asian Student Association
 - Bible Discussion Club

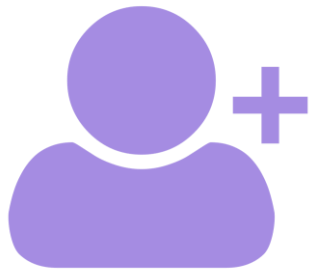


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- Personal data vs Personalized results **trade-off**
 - Example: biased results (Women in engineering)
 - Using “categories” data in AllUBCOClubs.json

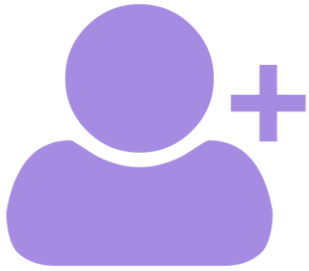
What about new users?

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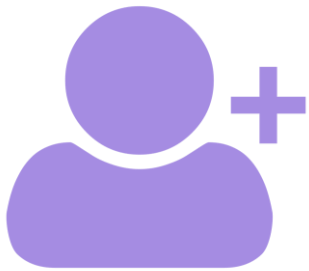
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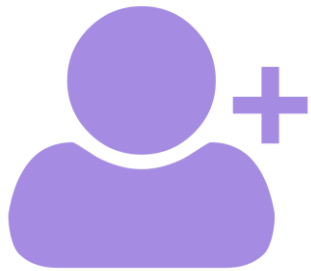


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 3. (Best case but more high-effort)
 - Obtain the **dataset of current students** and what clubs they are a part of
 - Generate an average portfolio by faculty
 - Display this order