



Interactive View Recommendation

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Background and Motivation

Visual data analysis tools use **utility functions** (UFs) to rank and recommend interesting views.

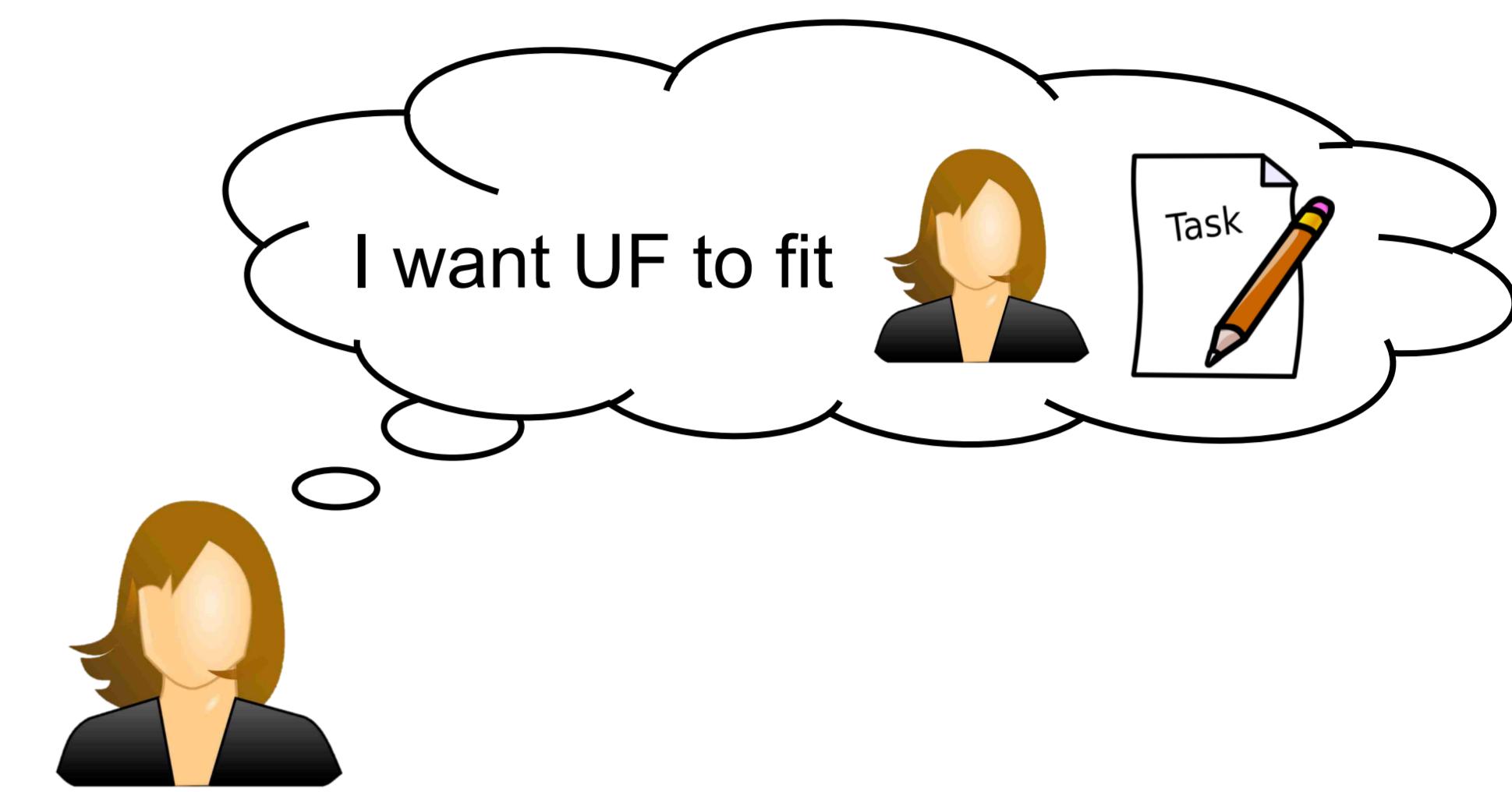


Utility Functions

A UF combines different **utility measures** (UMs) to derive a score representing **view interestingness**.



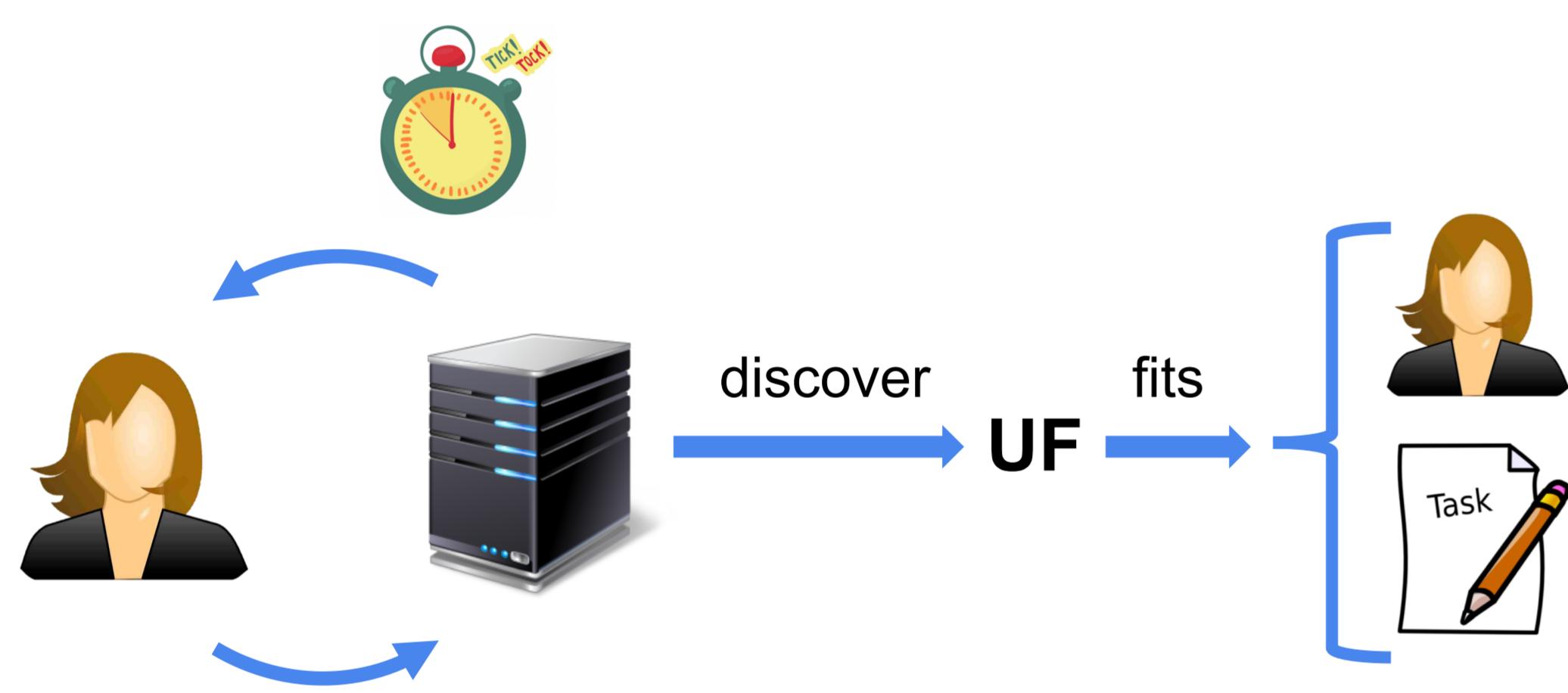
The UF in traditional view recommendation is **defined a priori**, so it **cannot adapt** to the analysis context.



ViewSeeker: An Interactive View Recommendation Tool

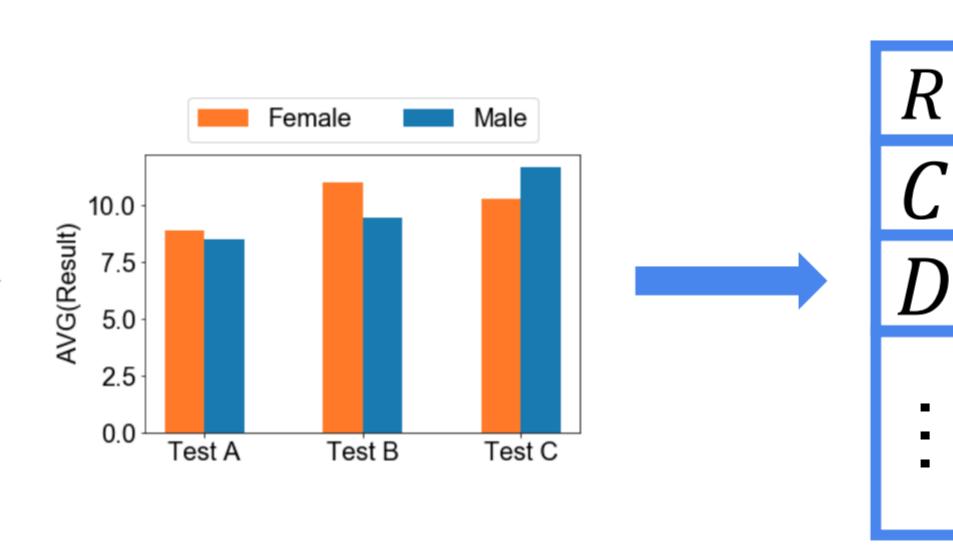
Interactive View Recommendation (IVR)

- IVR interacts with the user to **discover** the UF **most suitable** to the analysis context.
- IVR needs to have **interactive response time**.



ViewSeeker Phase 1: View Generation

It creates the views from the data and uses the **UM vector** of each view as its representation.

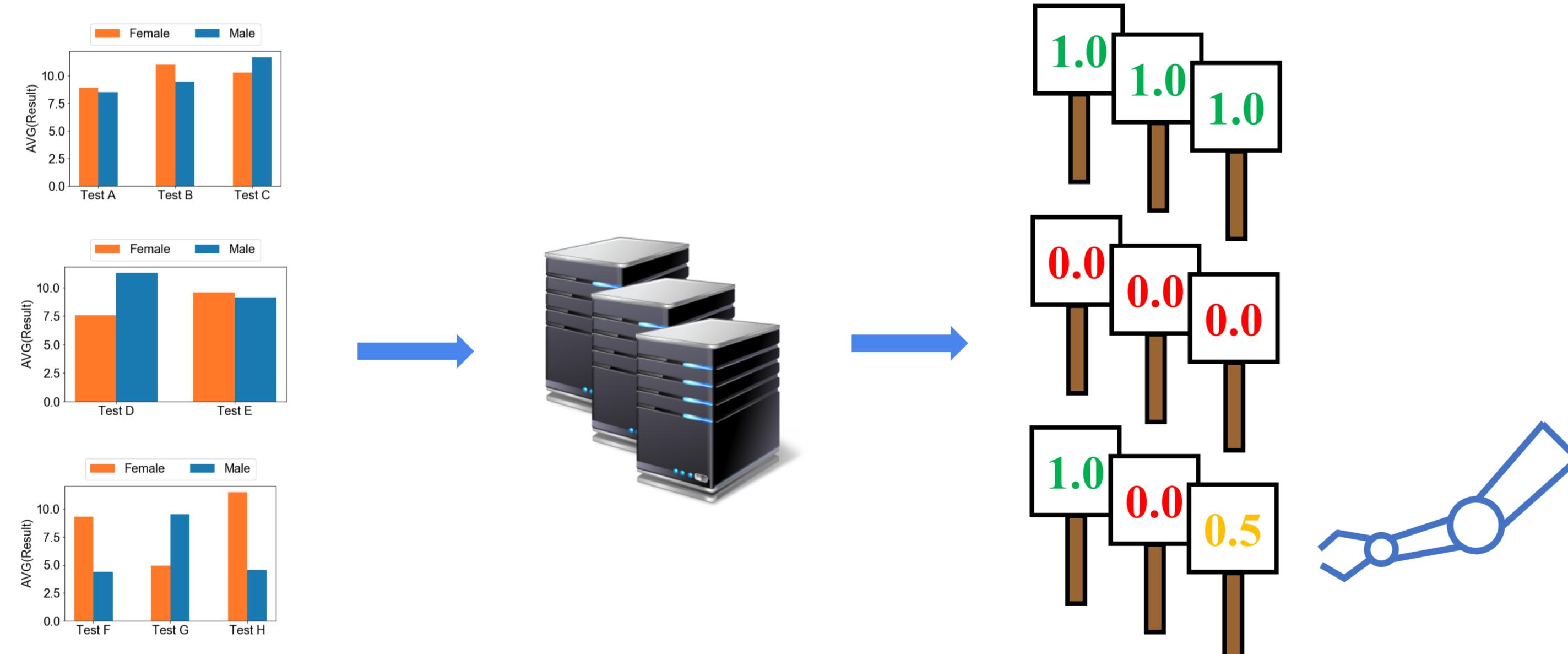


ViewSeeker Phase 2: User Interaction



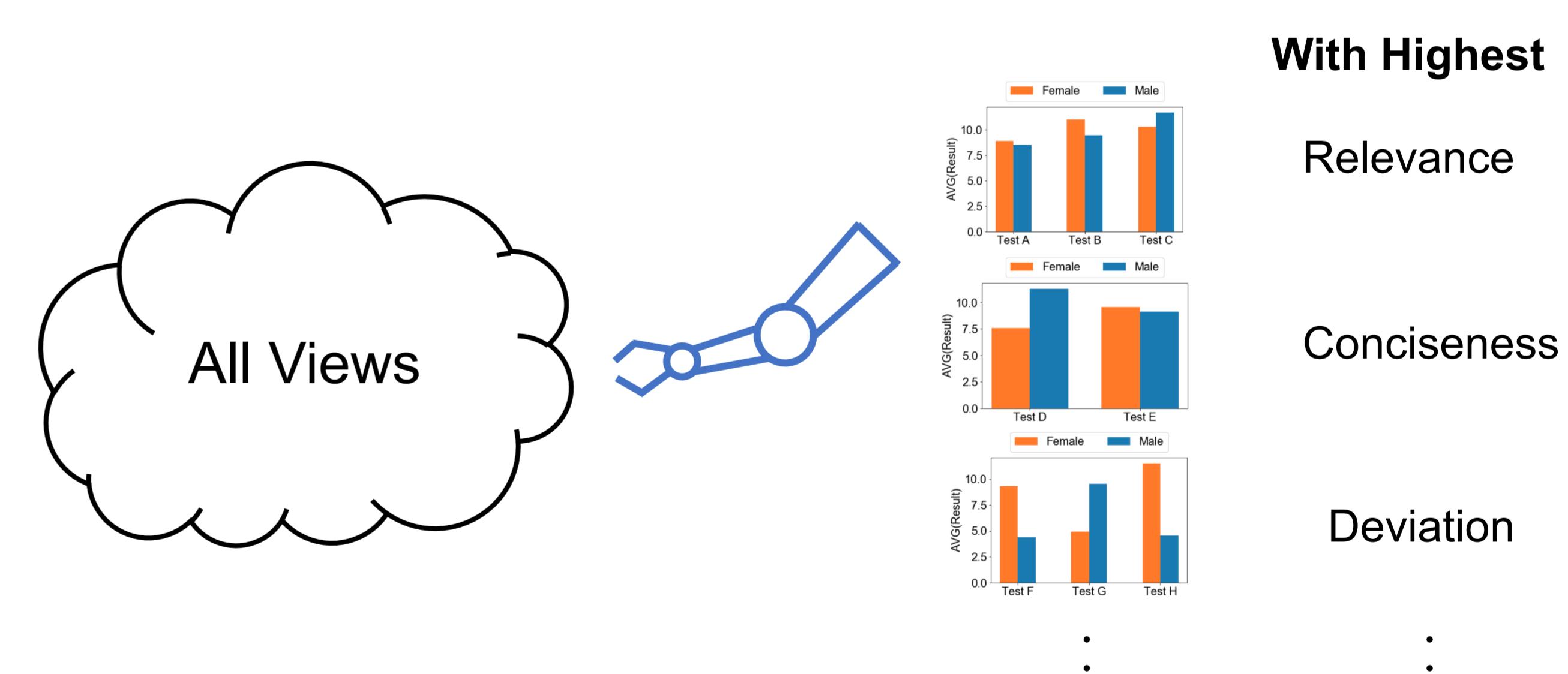
Example Selection Strategy: Query-By-Committee (QBC)

- User labeling effort is very expansive in IVR.
- ViewSeeker** uses QBC to reduce user effort and achieve fast **model parameter convergence**.
- QBC selects the example on which the learner committee has the **largest disagreement**.



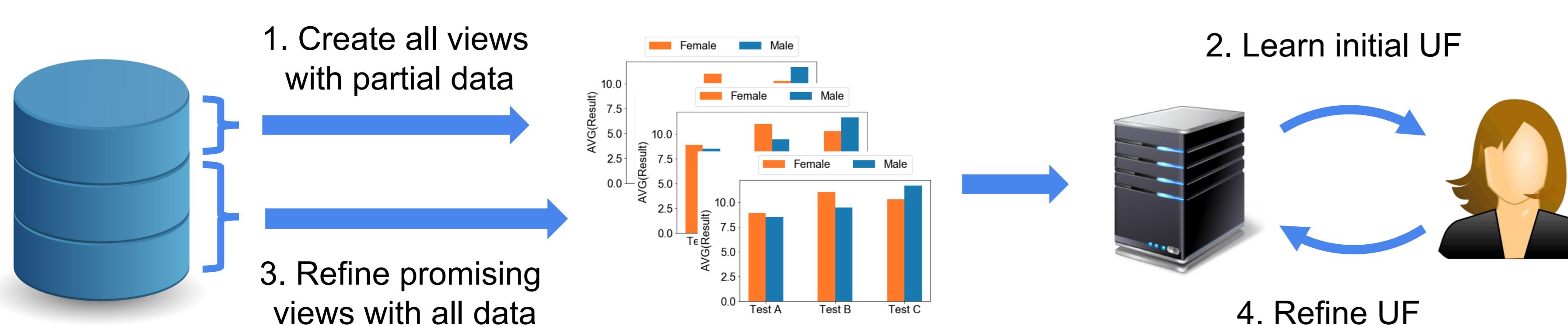
Optimization: Initial Example Selection

- In Phase 2, **ViewSeeker** first selects views with the highest score for each UM.
- These views are more likely to provide useful information about the UF in the **high view interestingness range**, in which the user has more interest.



Optimization: 2-Stage View Generation

- To achieve the interactive time limit t_l in Phase 1, **ViewSeeker** uses α percent of the data to generate all the views, with $\alpha = \frac{t_l \cdot s}{V}$, where s is the view generation speed and V is the total view count.
- To achieve t_l in Phase 2, in each interaction iteration, **ViewSeeker** selects n views with the highest interestingness based on the UF learned from user feedback for view refinement using all data, with $n = t_l \cdot s$.



Experimental Results

ViewSeeker outperforms the best baseline with defined-a-priori UF by **3X** in recommendation precision, and achieves 100% precision with only **9 - 16** labeled views.

Legend: KL, EMD, L1, L2, MaxDiff, Usability, Accuracy, P-Value, ViewSeeker

