

VisMatchmaker: Cooperation of the **User** and the **Computer** in Centralized Matching Adjustment

Po-Ming Law

Wenchao Wu

Yixian Zheng

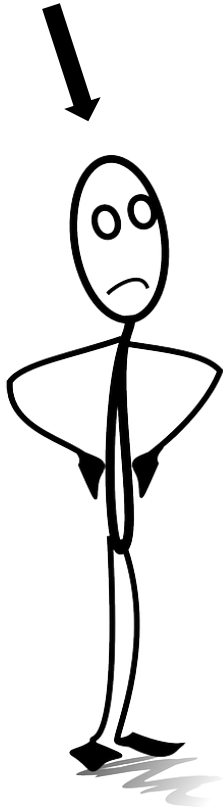
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AND TECHNOLOGY

Mentor-Mentee Matching

A puzzled advisor
at Georgia Tech



Mentors

Peter
Ken
Betty
Jane
John

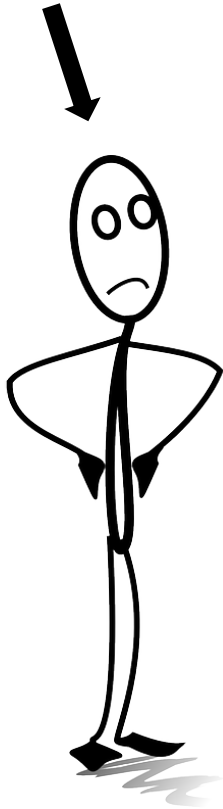
Mentees

David
Ben
Jack
Mary
May

Mentor-Mentee Matching

Mentees' Preference Lists

A puzzled advisor
at Georgia Tech

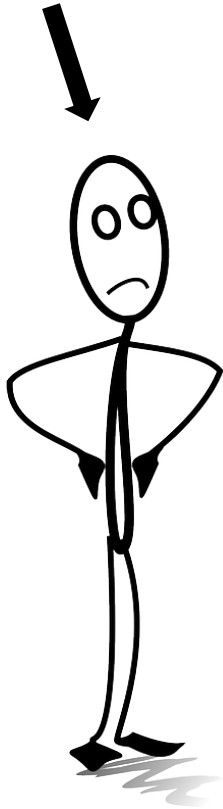


David	Ben	Jack	Mary	May
1. Jane	1. Ken	1. ...	1. ...	1. ...
2. Betty	2. Jane	2. ...	2. ...	2. ...
3. Peter	3. John	3. ...	3. ...	3. ...
4. John	4. Betty	4. ...	4. ...	4. ...

Mentor-Mentee Matching

Mentors' Preference Lists

A puzzled advisor
at Georgia Tech



Peter
1. David
2. May
3. Jack
4. Ben

Ken
1. May
2. Mary
3. Jack
4. Ben

Betty
1. ...
2. ...
3. ...
4. ...

Jane
1. ...
2. ...
3. ...
4. ...

John
1. ...
2. ...
3. ...
4. ...

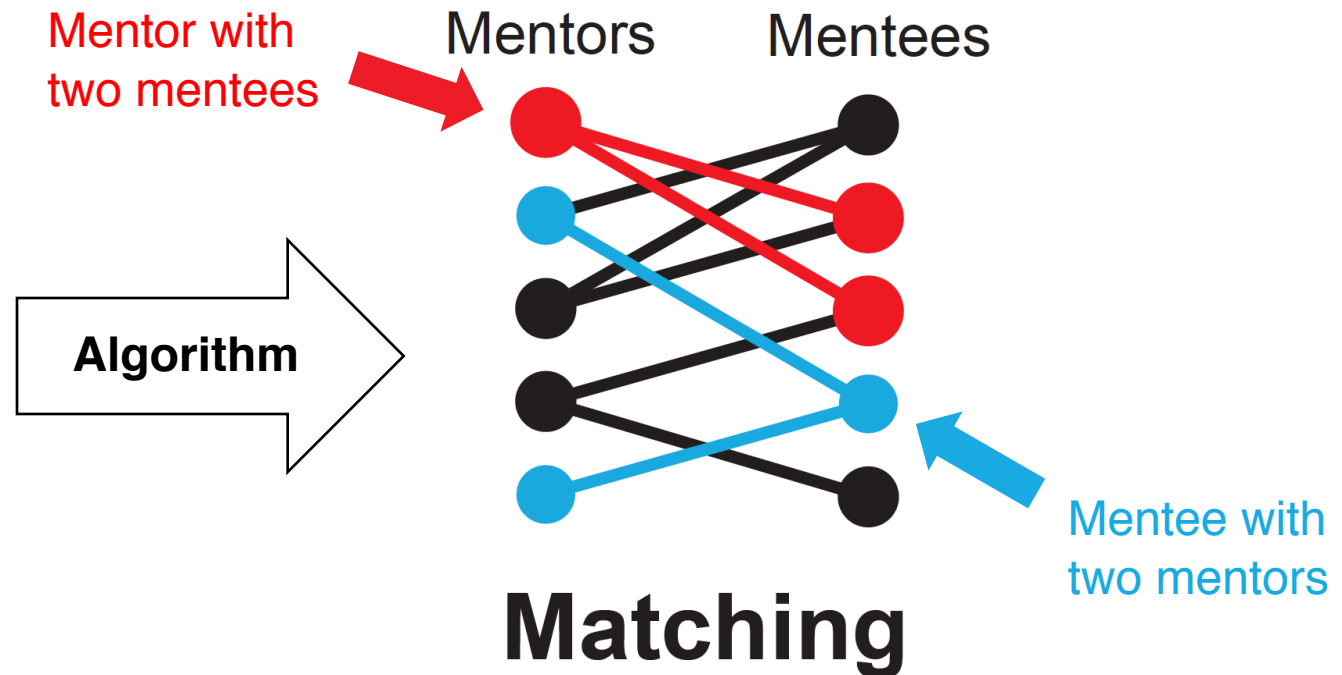
Mentor-Mentee Matching

Mentors' Preference Lists

Peter	Ken	Betty	Jane	John
1. David	1. May	1. ...	1. ...	1. ...
2. May	2. Mary	2. ...	2. ...	2. ...
3. Jack	3. Jack	3. ...	3. ...	3. ...
4. Ben	4. Ben	4. ...	4. ...	4. ...

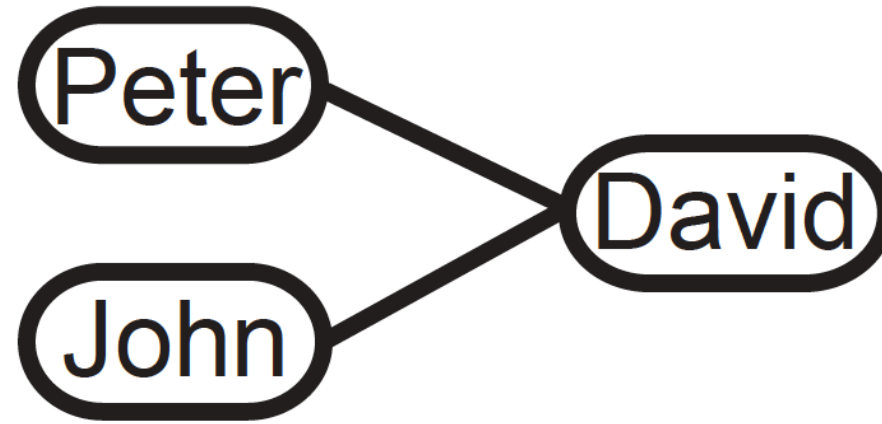
Mentees' Preference Lists

David	Ben	Jack	Mary	May
1. Jane	1. Ken	1. ...	1. ...	1. ...
2. Betty	2. Jane	2. ...	2. ...	2. ...
3. Peter	3. John	3. ...	3. ...	3. ...
4. John	4. Betty	4. ...	4. ...	4. ...



What if...

Mentors	Mentee
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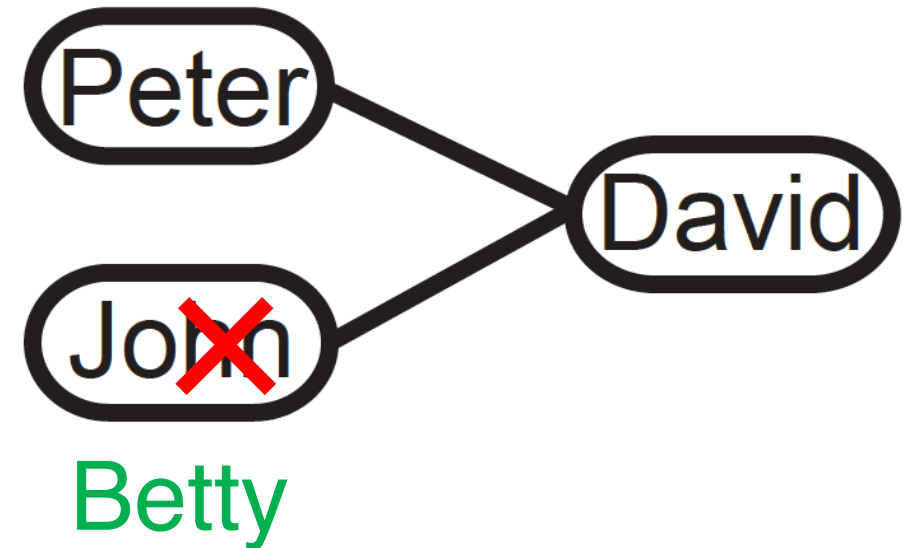
The Cascading Nightmare

David's Preference

1. Jane
2. Betty
3. Peter
4. John

Mentors

Mentee



Possible to improve?

The Cascading Nightmare

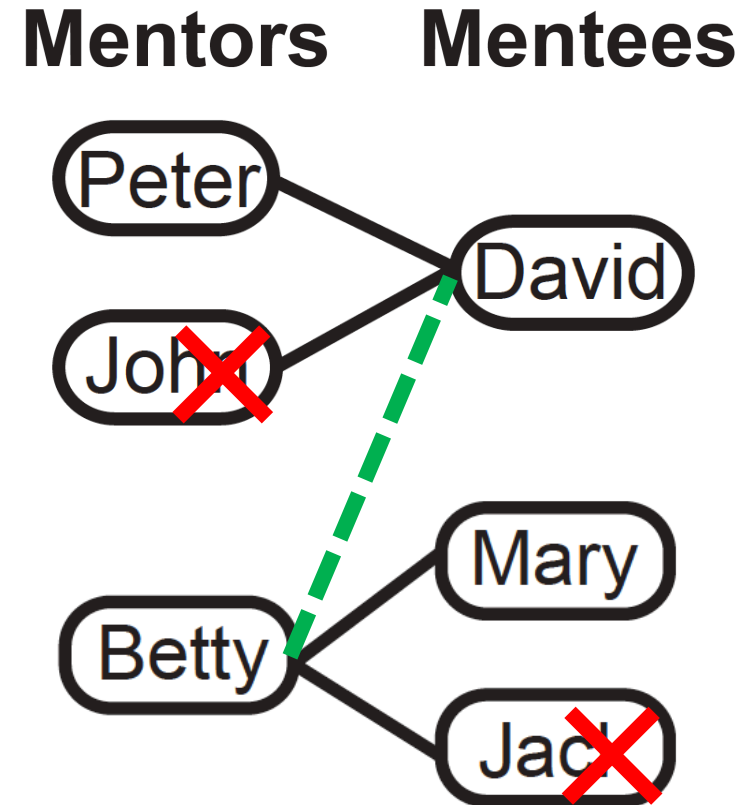
David's Preference

1. Jane

2. Betty

3. Peter

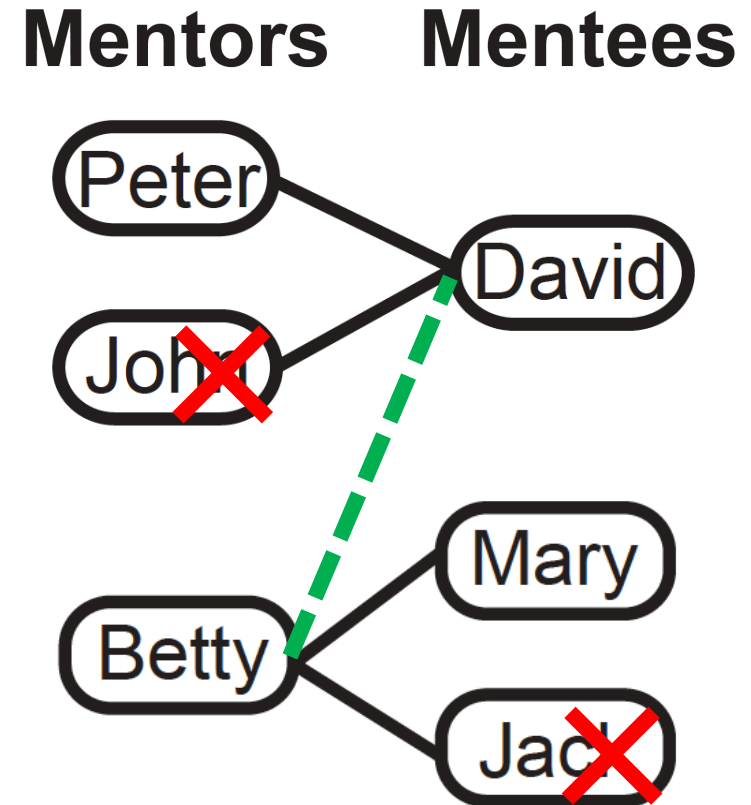
4. John



Betty can have at most two partners

The Cascading Nightmare

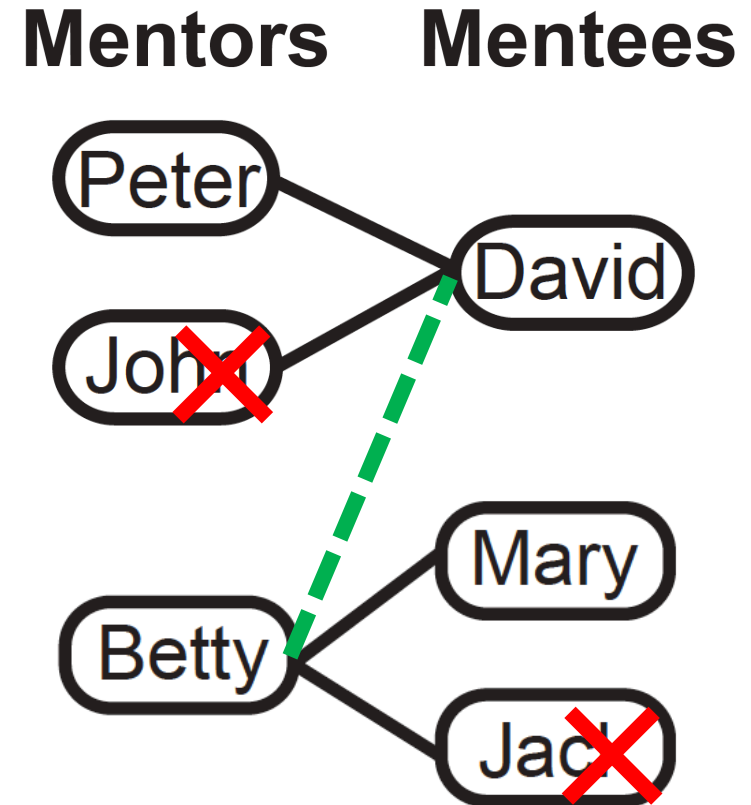
- Jack: one less mentor
- John: one less mentee
- Pair Jack with John?



Jack may not like John!

The Cascading Nightmare

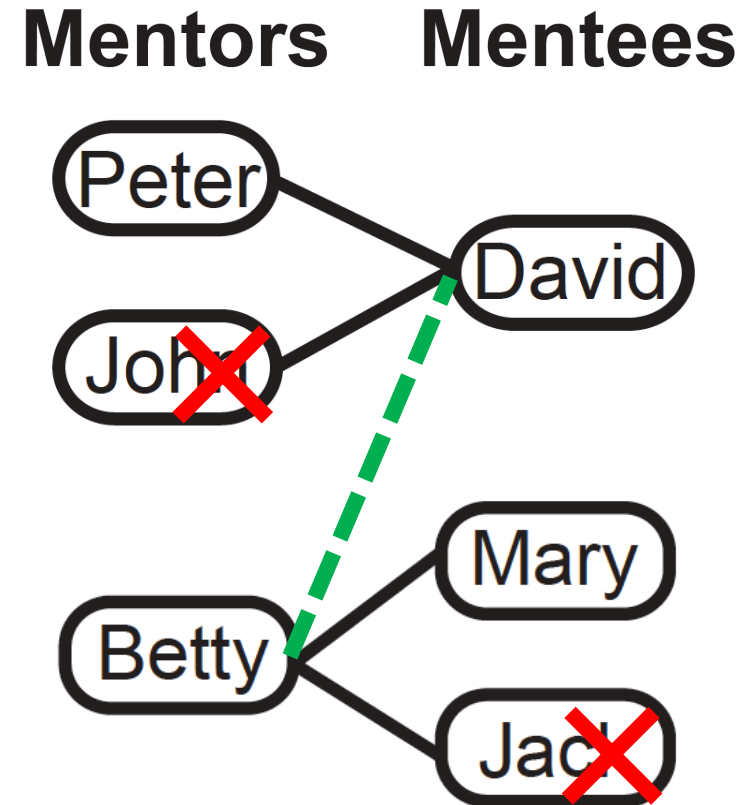
- Assign a different mentor to Jack and a different mentee to John?



The cascading nightmare continues!!

The Cascading Nightmare

- When you change a pairing, some people become unmatched
- Match them to the others, the pairing is changed again



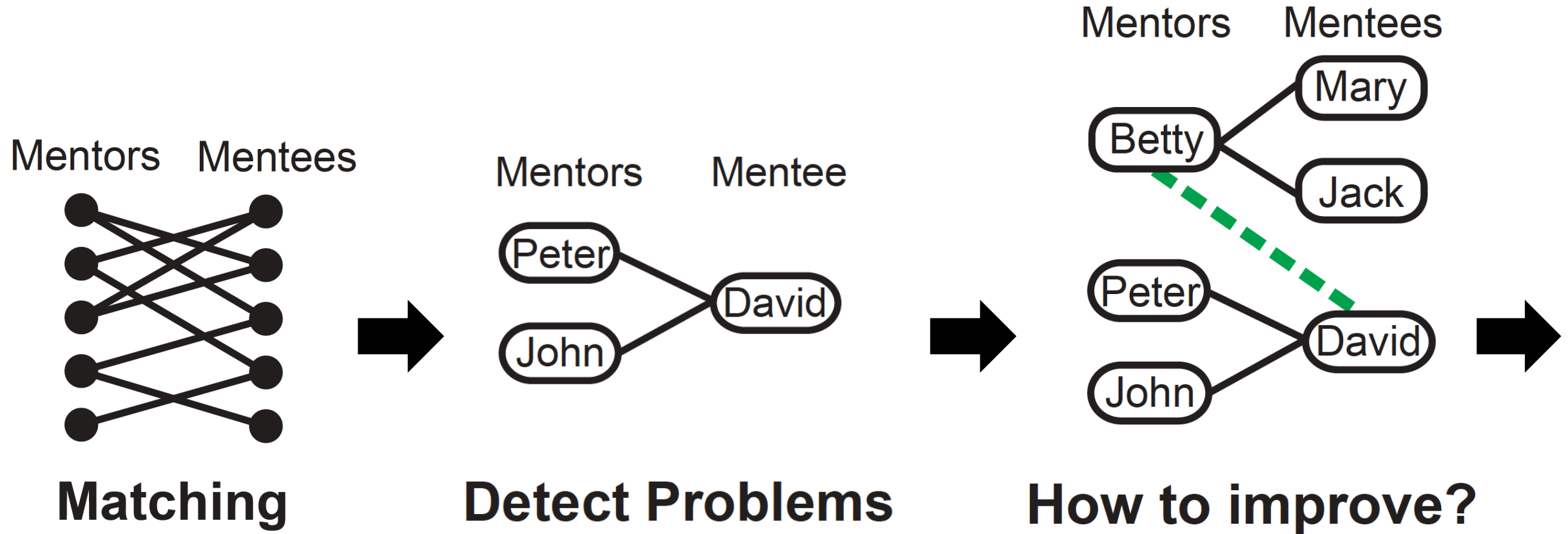
The cascading effect can go on and on!!!

Have an algorithm-generated matching
Want to adjust it and explore matching alternatives
HARD to adjust it manually!!

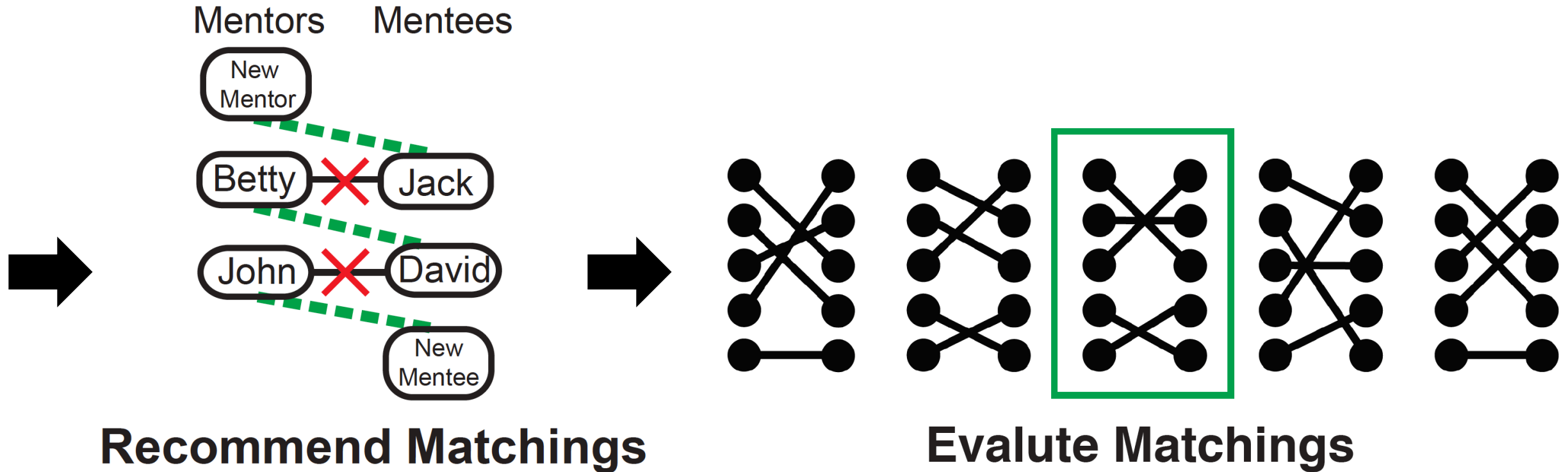
Cooperation of user and computer

Users	→	What to adjust
System	→	How to adjust

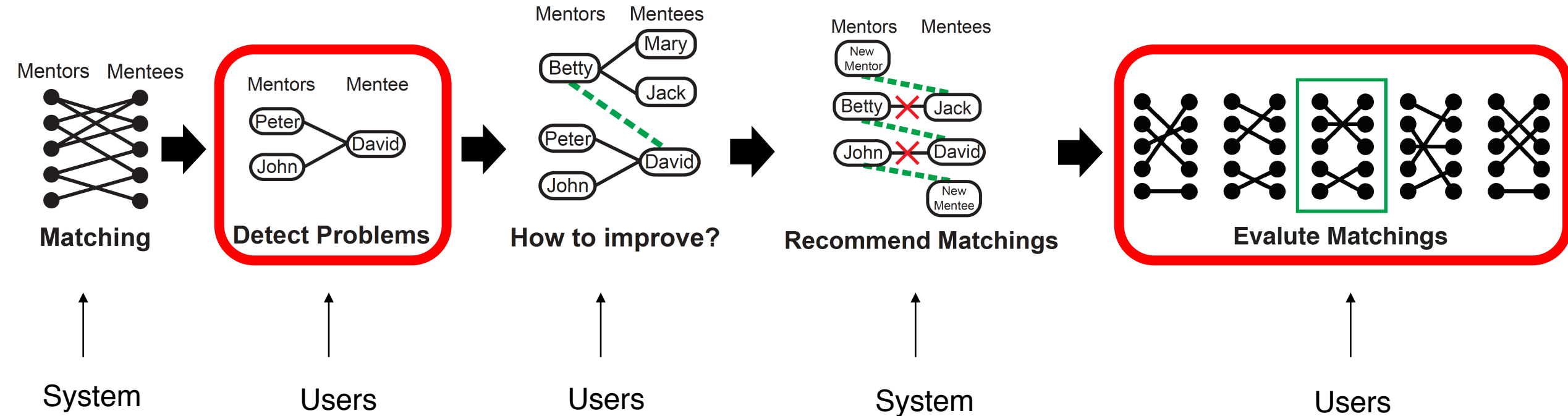
Cooperation of User and Computer



Cooperation of User and Computer



Cooperation of User and Computer

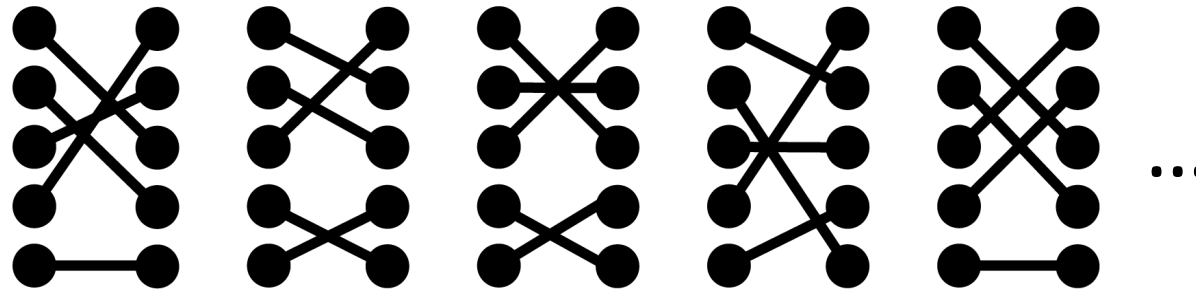


Problem Detection

Questions about allocation result	Questions about allocation mechanism
Who are those who get poor matches? Improve their matches	There is a popular person. Why pair A not B with him? Try to pair the popular person with the others

Matching Evaluation

Many matchings... How to compare them?



- Agent-level comparison
- Comparison between two matchings
- Comparison among multiple matchings

Matching Evaluation

Agent-level comparison

- Want to improve David's match
- Prefer matchings which give David a better match

Need to compare the quality of match of each person in multiple matchings

Matching Evaluation

Comparison between two matchings

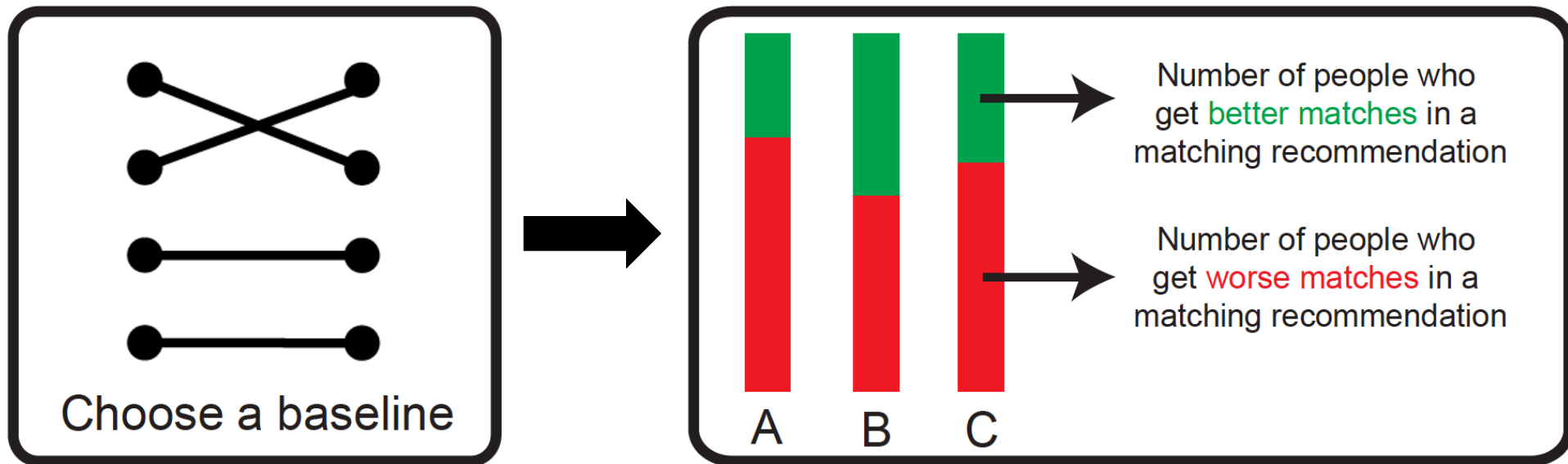
- Number of people who get better matches in matching 1 (in 2)
- Who are the people who get better matches in matching 1 (in 2)?
- By how much are their matches better?

May prefer matching 1 if...

- Most people get better matches in matching 1

Matching Evaluation

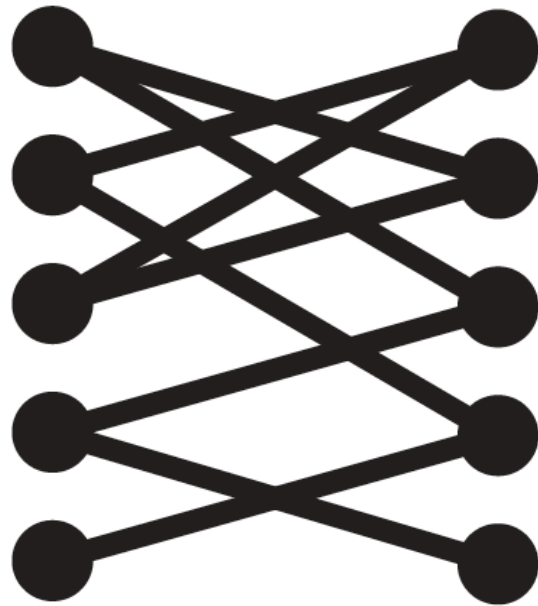
Comparison among multiple matchings



1. Choose original matching as a baseline
2. Compute the differences between the other matchings and the baseline
3. Compare differences across many matchings

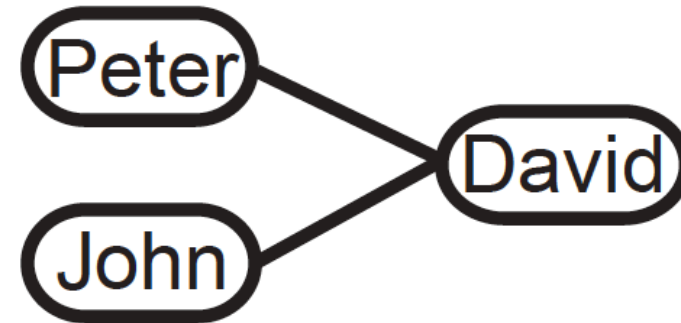
Number Lines

Mentors Mentees

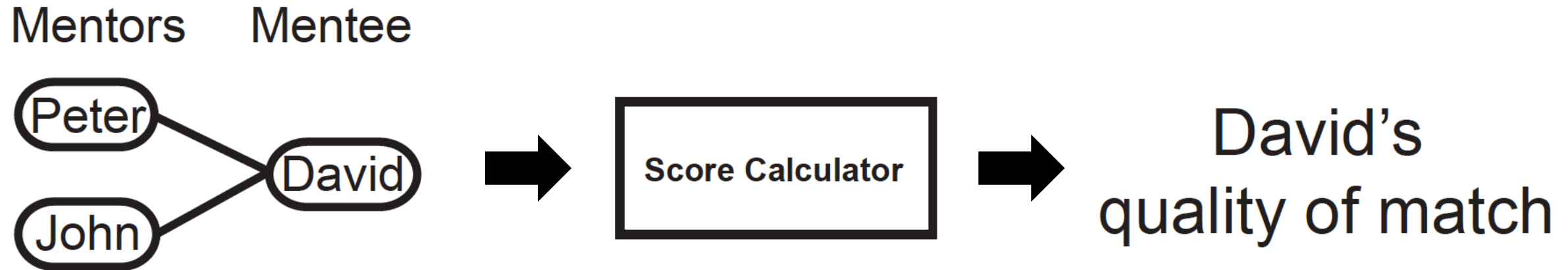


Algorithmic Result

Mentors Mentee

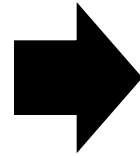


Number Lines

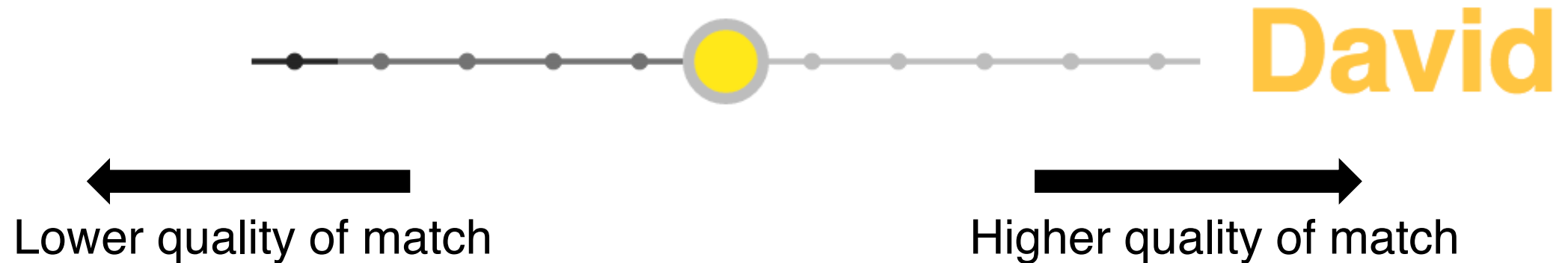


Number Lines

David's
quality of match

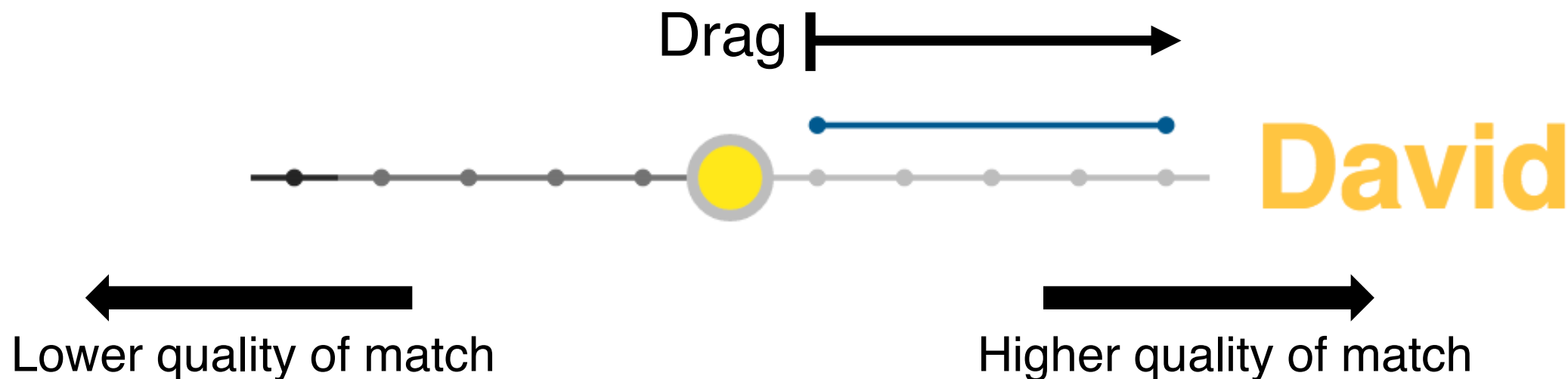


Position on
David's number line



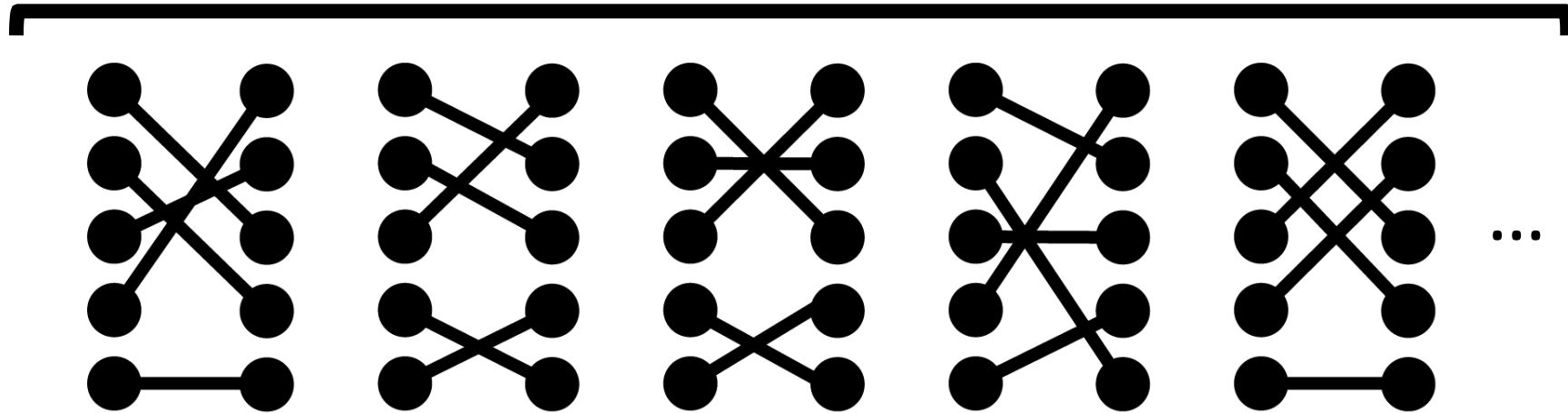
Number Lines

“Find matchings in which David’s quality of match falls within this range”



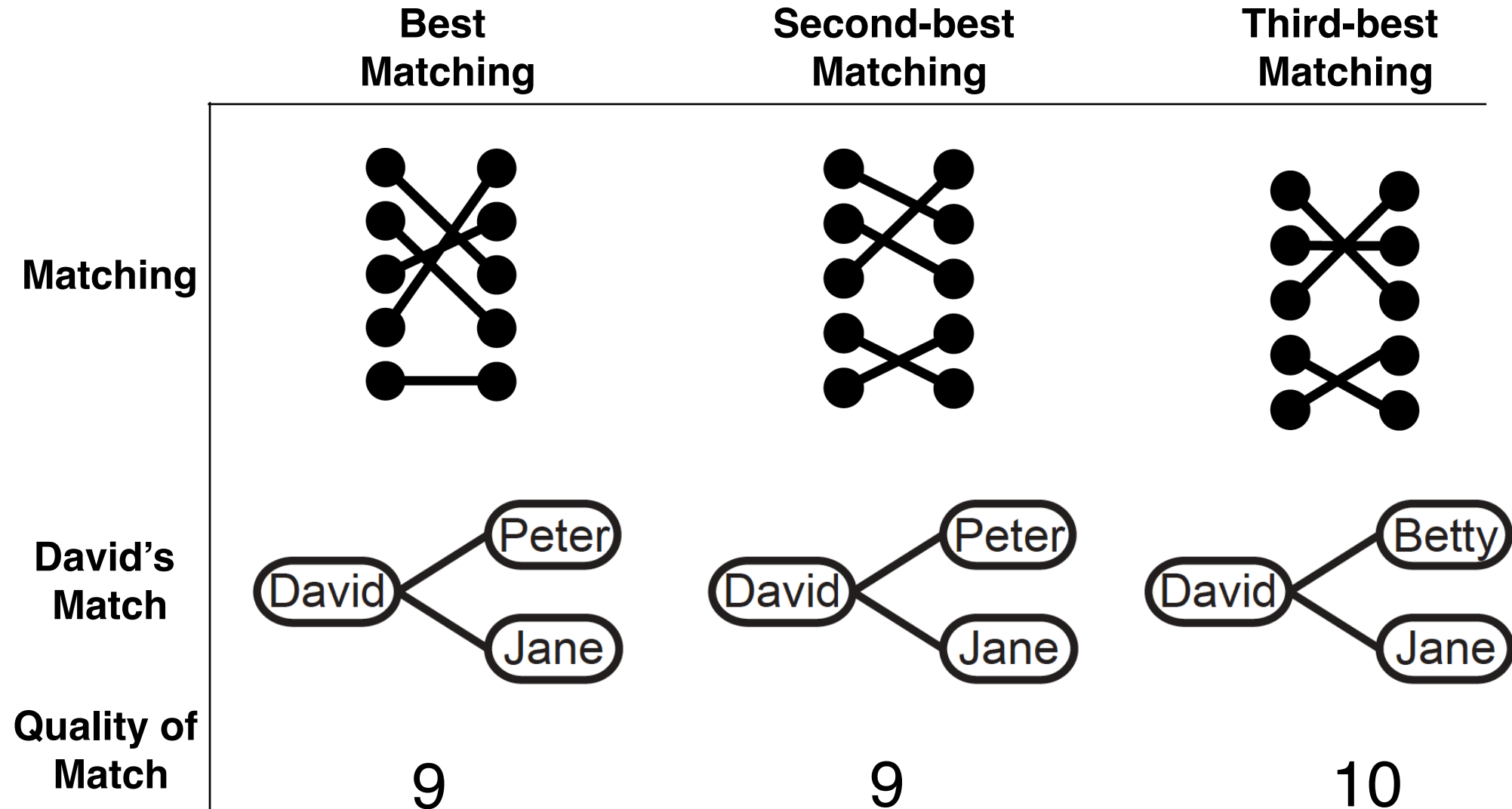
Number Lines

System finds 100 matchings



Ranked by a performance metric

Number Lines



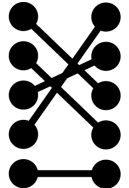
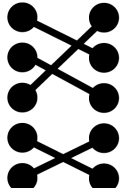
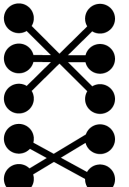
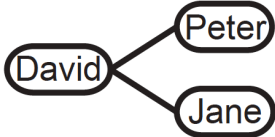
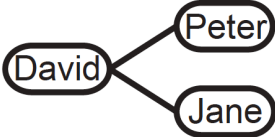
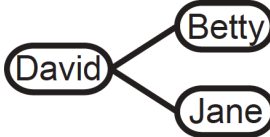
Number Lines

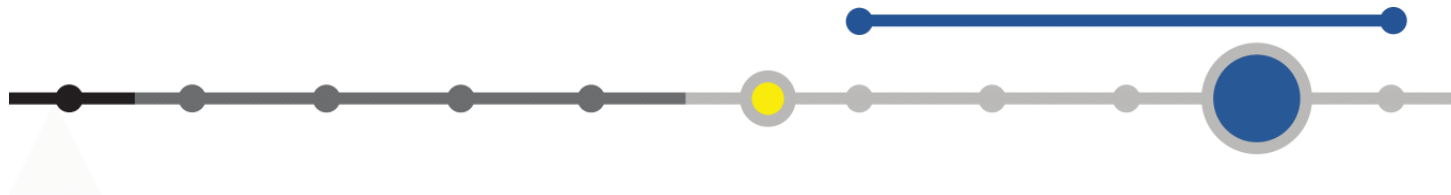
	Best Matching	Second-best Matching	Third-best Matching
Matching			
David's Match			
Quality of Match	9	9	10



David


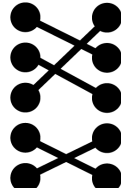
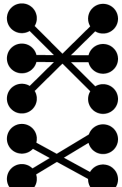
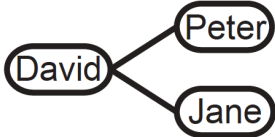
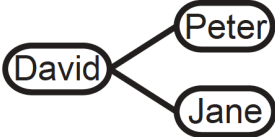
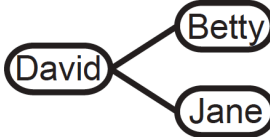
Number Lines

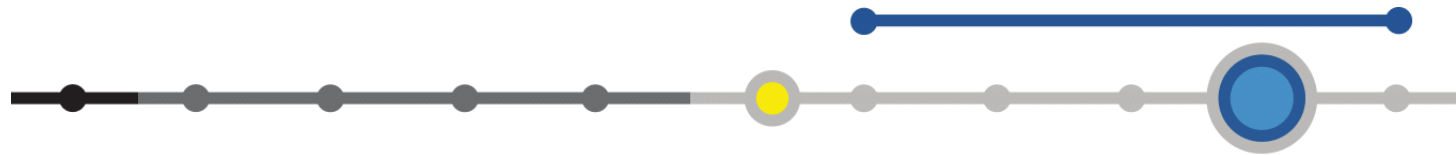
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Matching			
David's Match			
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David

Number Lines

	Best Matching	Second-best Matching	Third-best Matching
Matching			
David's Match			
Quality of Match	9	9	10




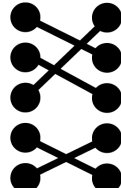
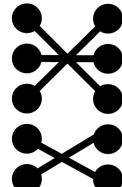
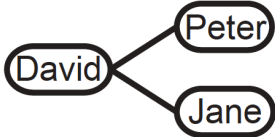
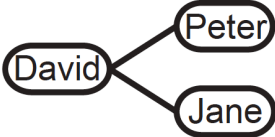
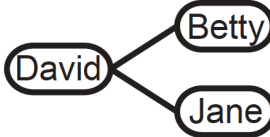
David

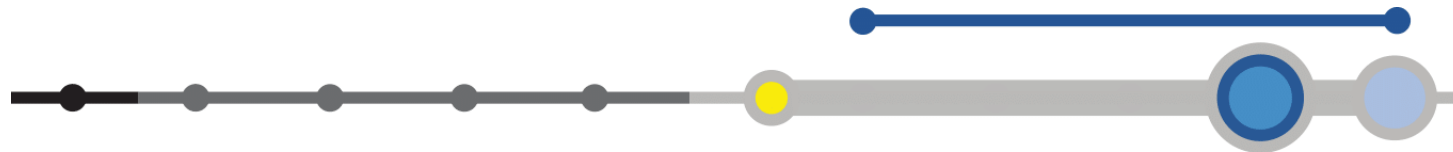
Number Lines

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Matching			
David's Match			
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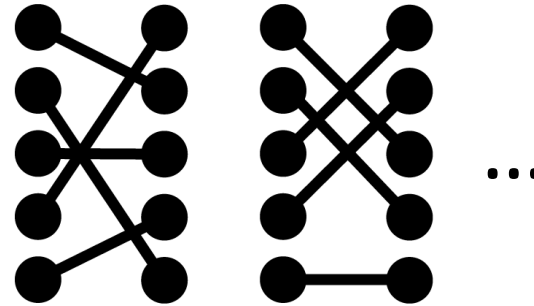
Number Lines

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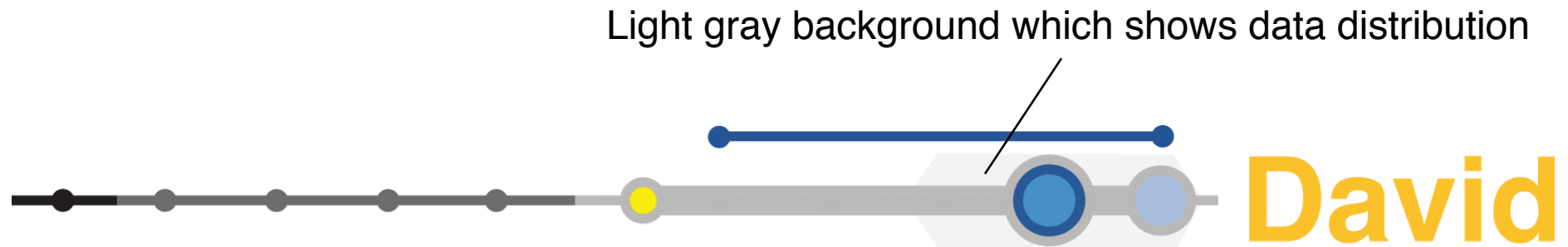


David

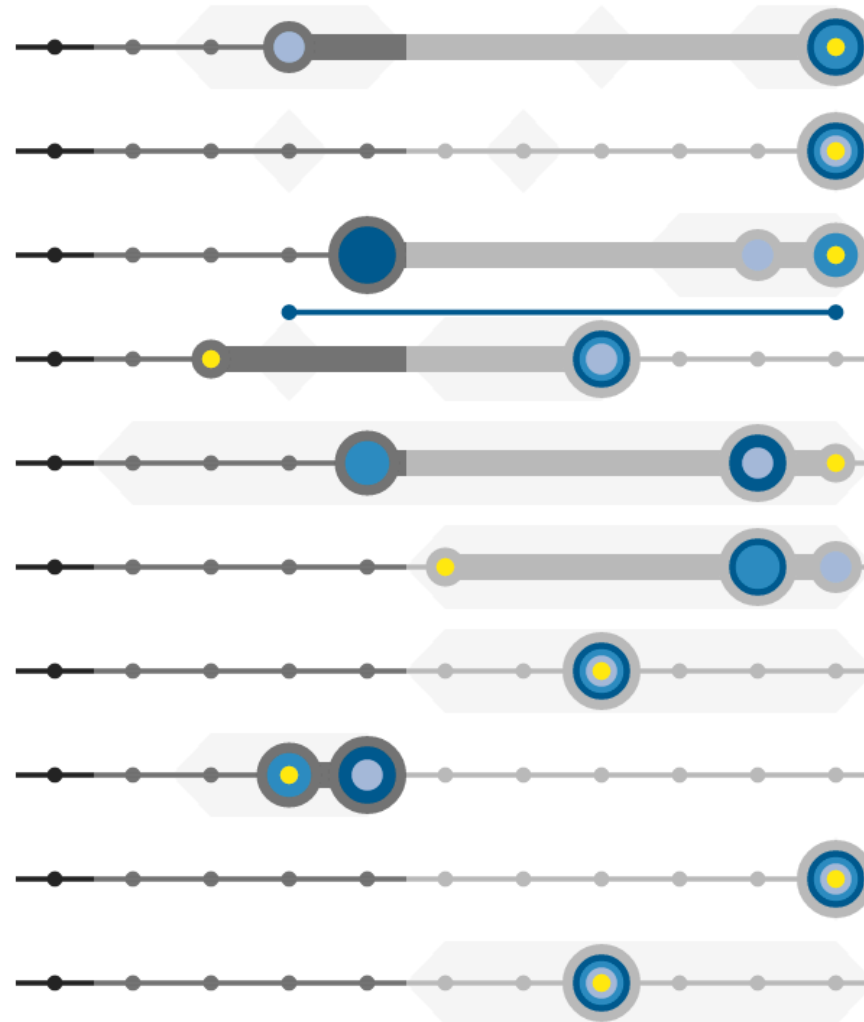
Number Lines



Matchings with lower rankings



Visual Preference List



Peter

Ken

Betty

Jane

John
David

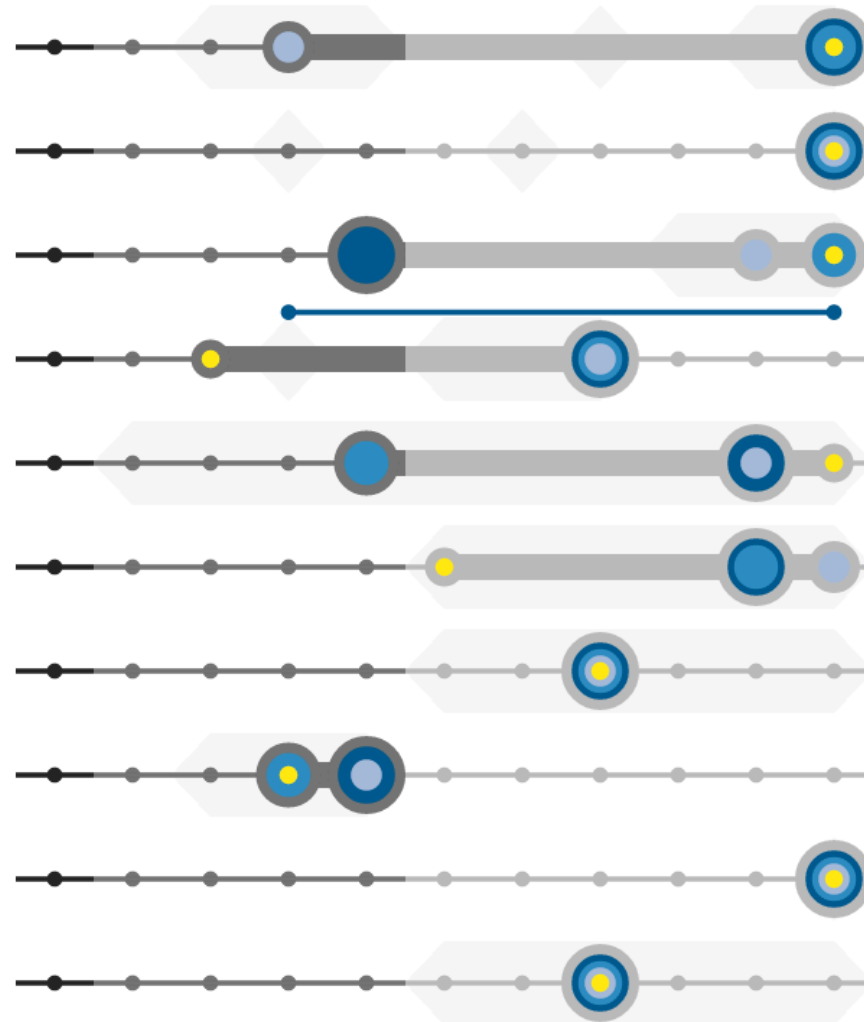
Ben

Jack

Mary

May

Visual Preference List



Peter	
Ken	
Betty	
Jane	
John	
David	
Ben	
Jack	
Mary	
May	

Visual Preference List

Label size: popularity

Can have at most 2 partners

Jack

First choice

Fourth choice

- Not matched
- Matched
- Not chosen by that person

Visual Preference List



“John and Jack work well together.
Match them together”



Jack



Jack

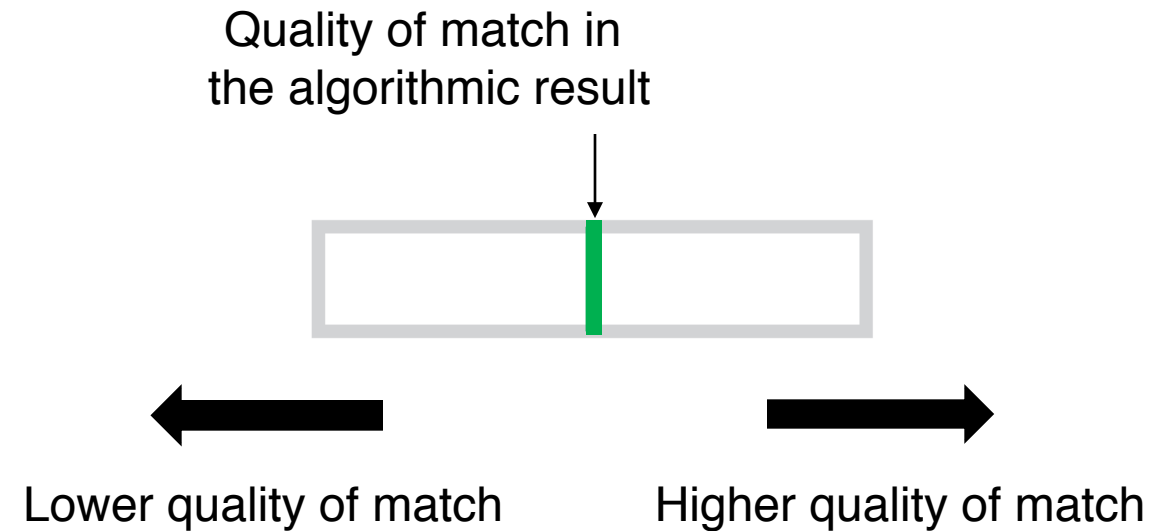
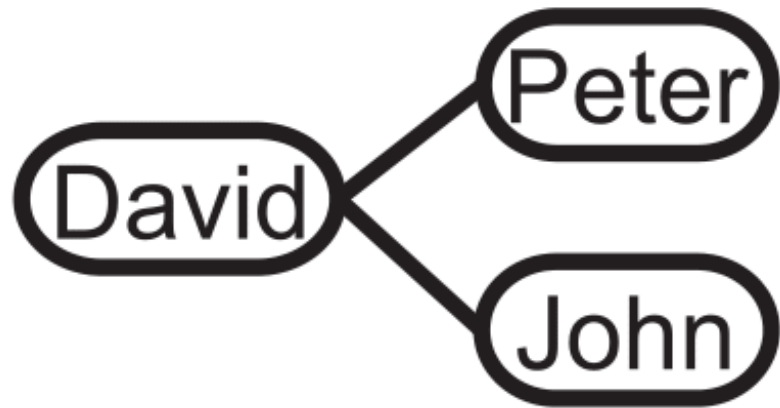


May



Stacked Graph

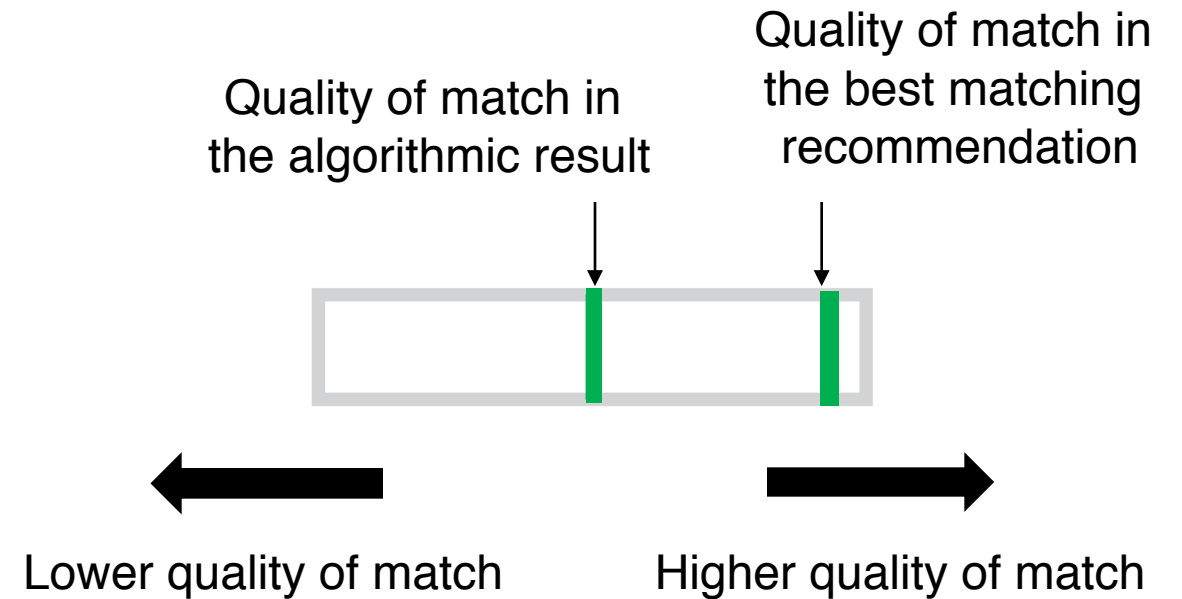
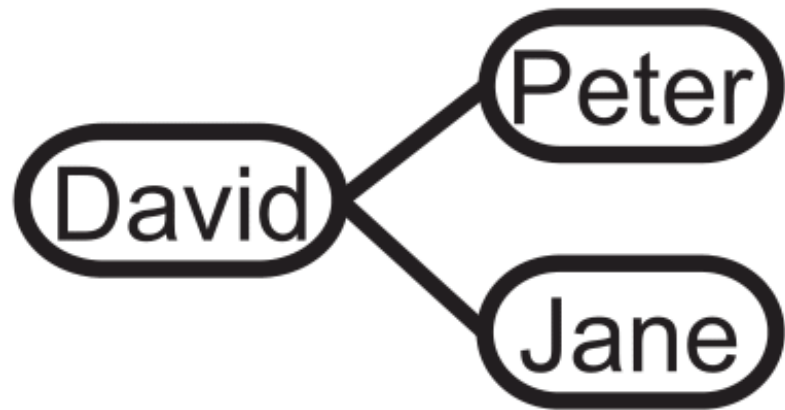
Consider the algorithmic result and the best matching recommendation



In the algorithmic result

Stacked Graph

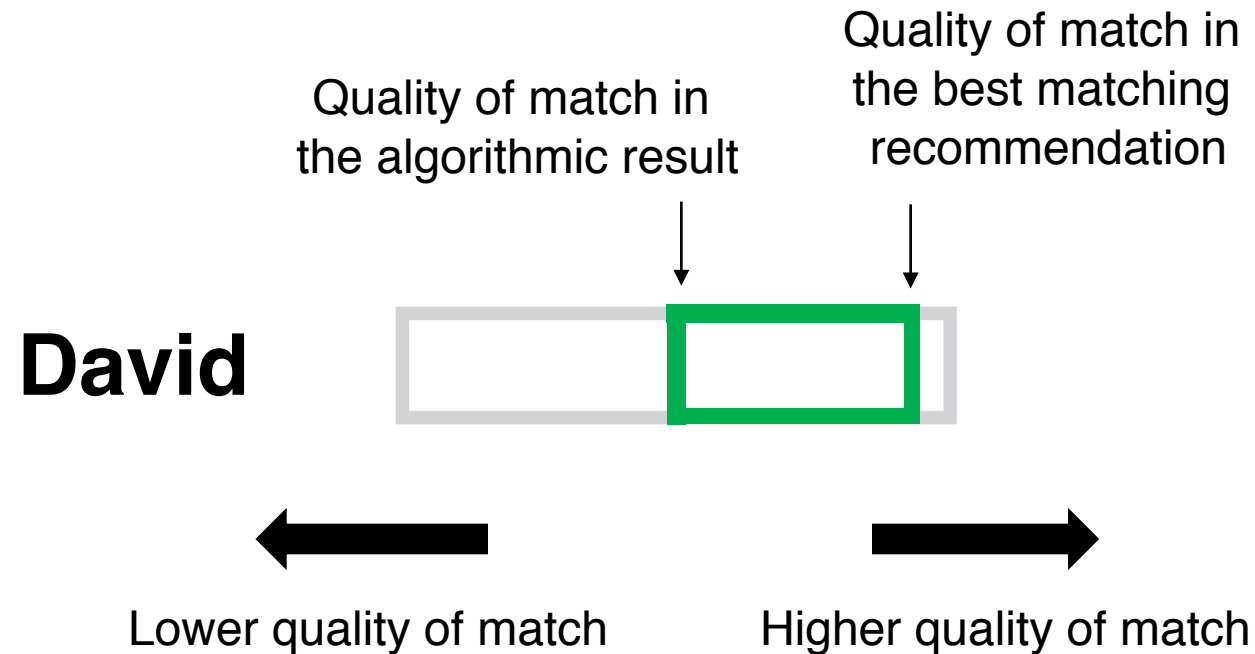
Consider the algorithmic result and the best matching recommendation



In the best matching recommendation

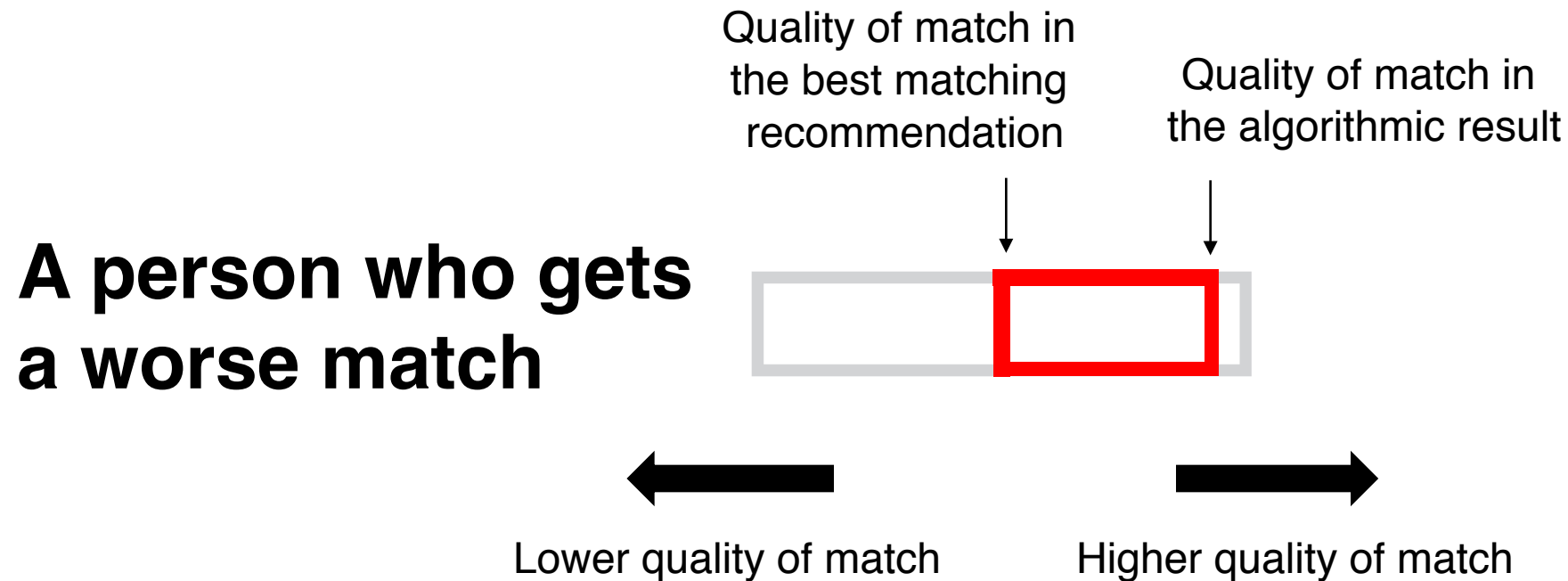
Stacked Graph

Consider the algorithmic result and the best matching recommendation

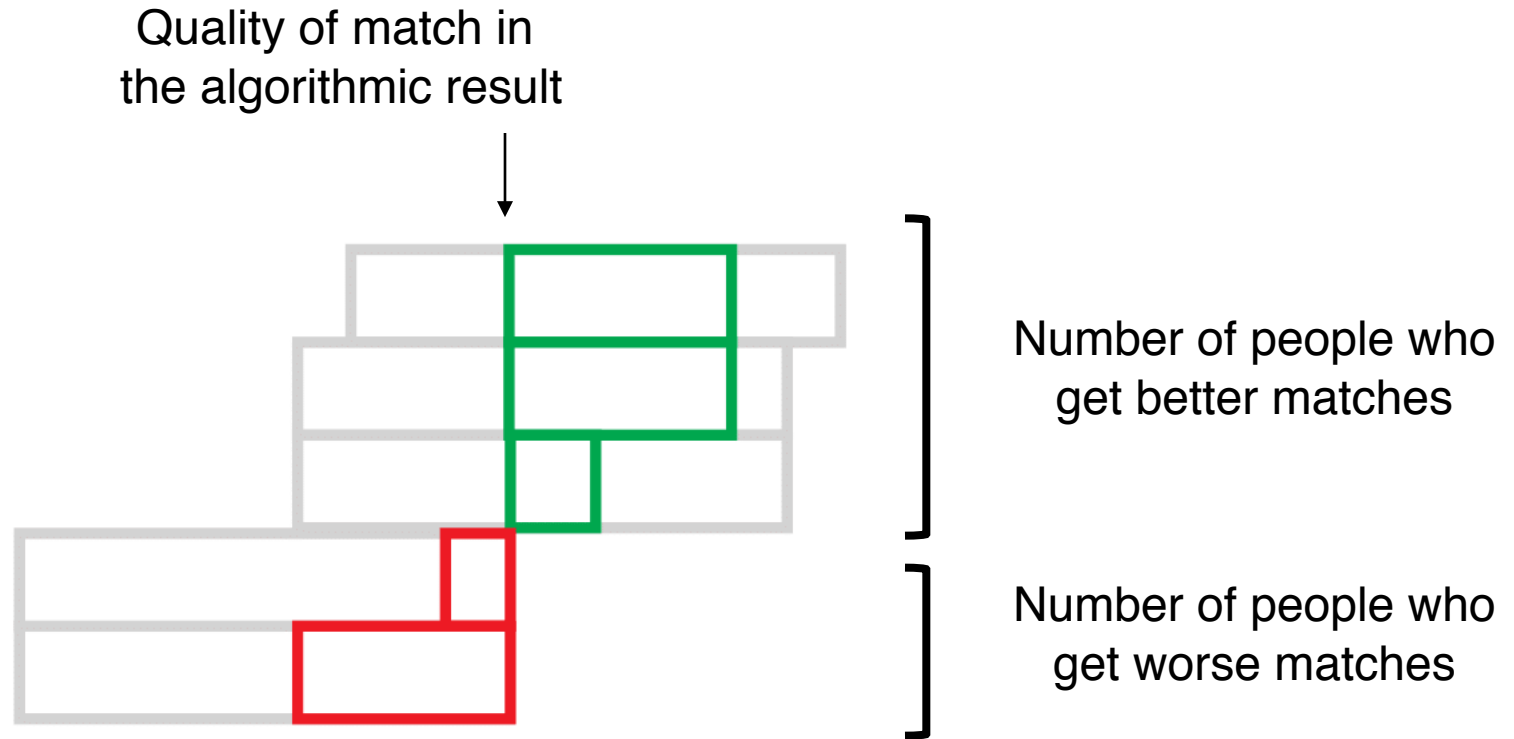


Stacked Graph

Consider the algorithmic result and the best matching recommendation



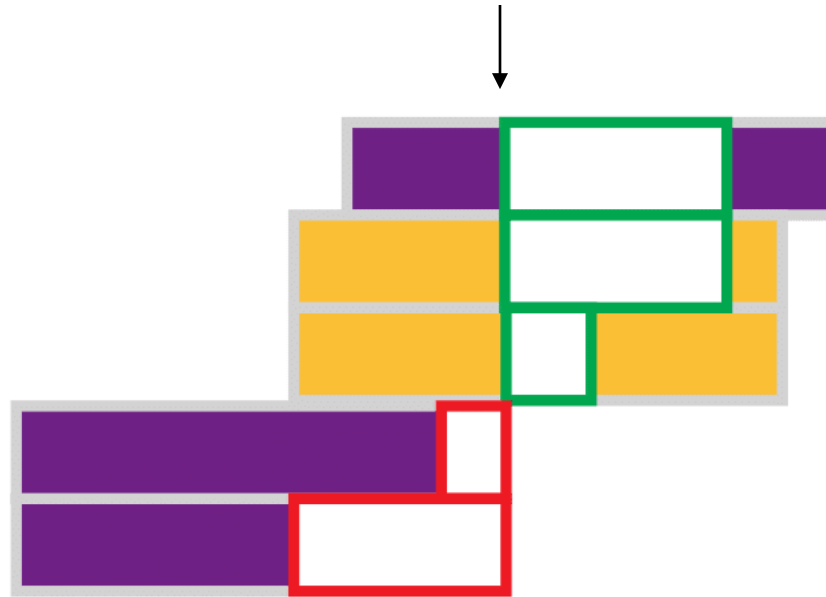
Stacked Graph



Stacked Graph

● **Mentor**
● **Mentee**

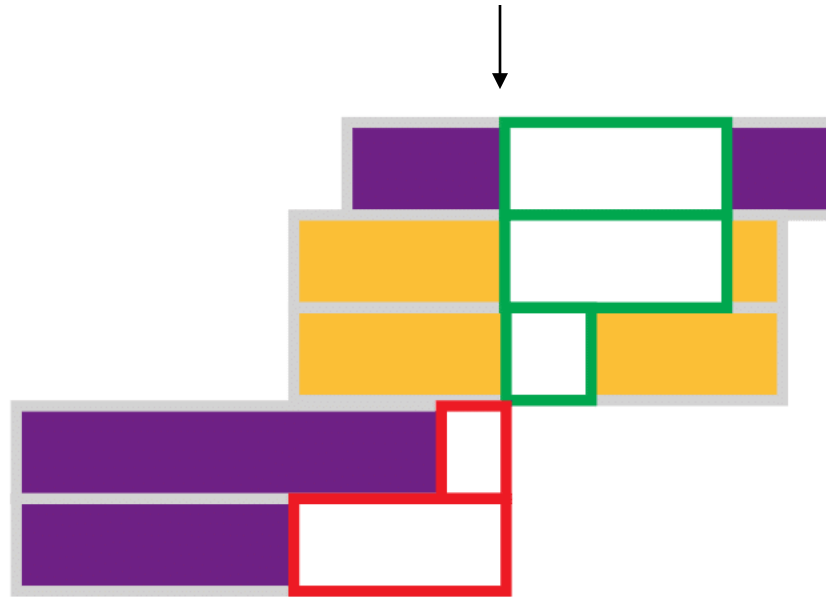
Quality of match in
the algorithmic result



Stacked Graph

● Mentor
● Mentee

Quality of match in
the algorithmic result



Comparing the algorithmic result and
the best matching recommendation

Stacked Graph

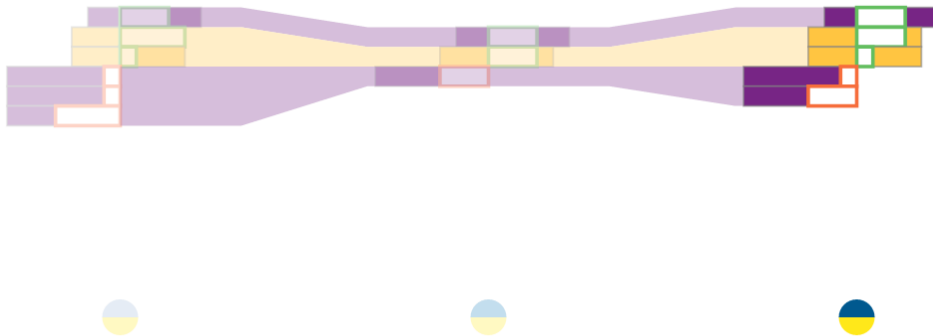
● Mentor

● Mentee

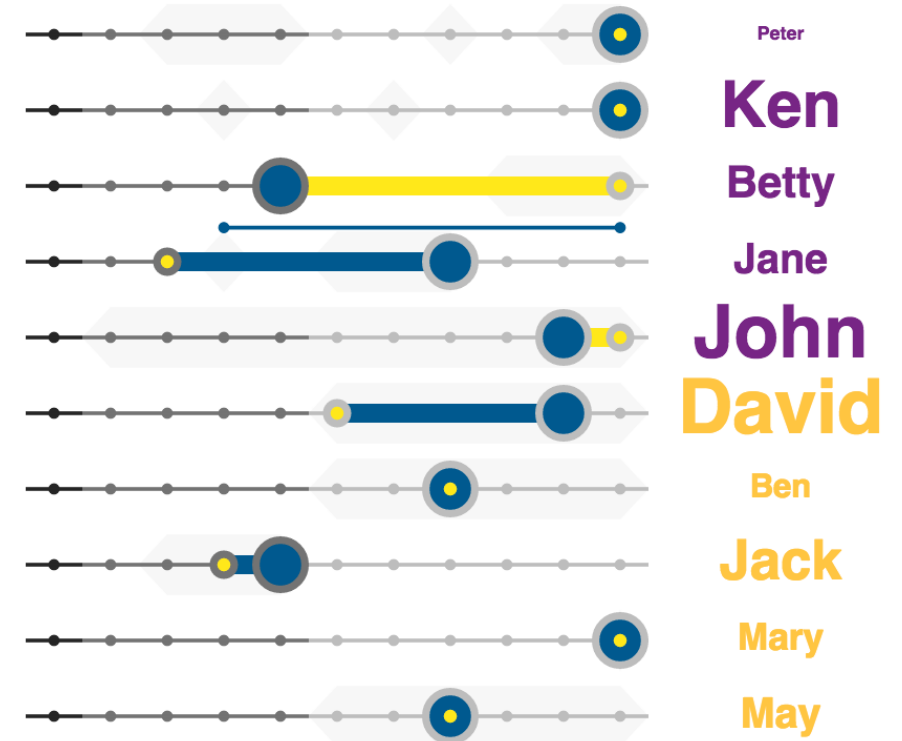


Stacked Graph

● Mentor
● Mentee

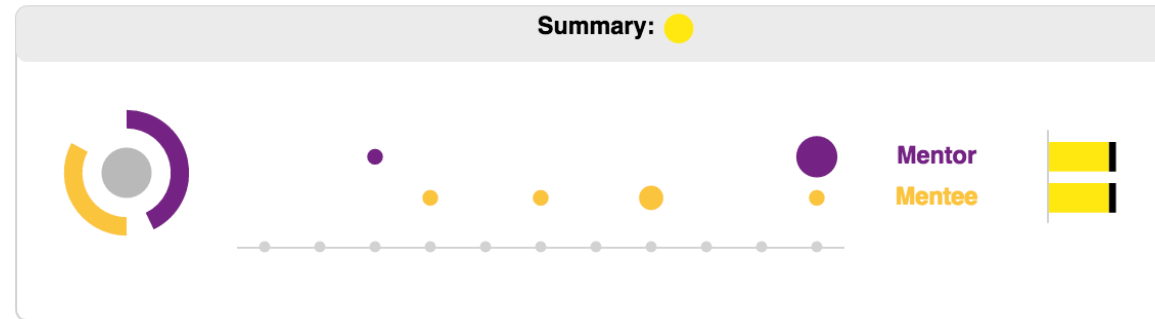


Hover
(detail on demand)

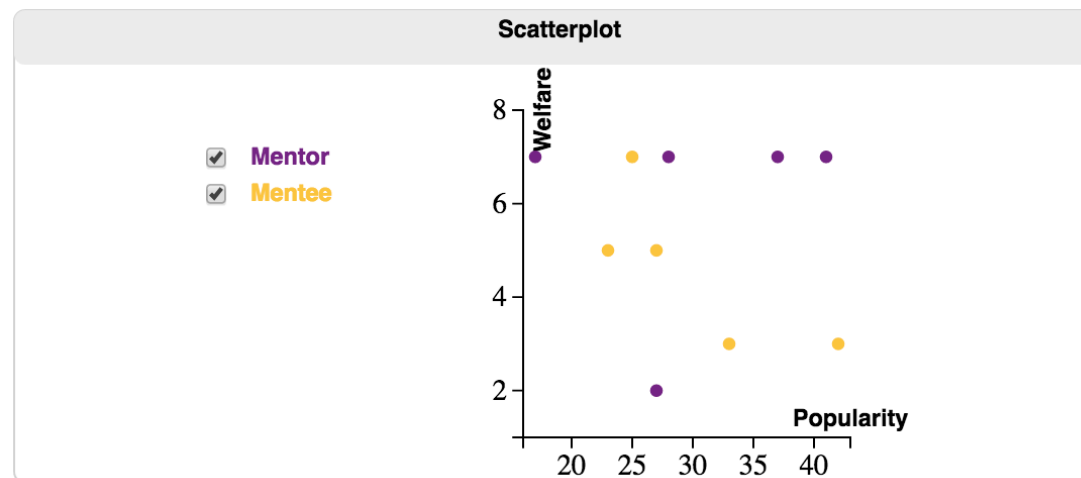


Other Views

Glyphs for summary statistics



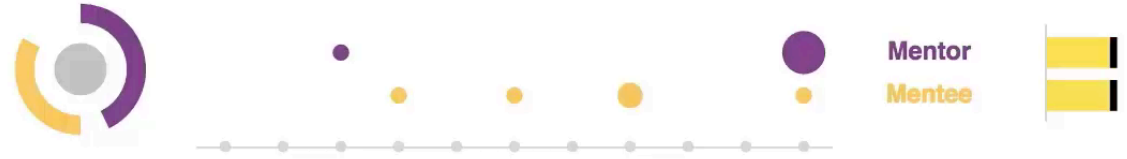
A scatterplot for filtering the number lines



Stacked Graph View



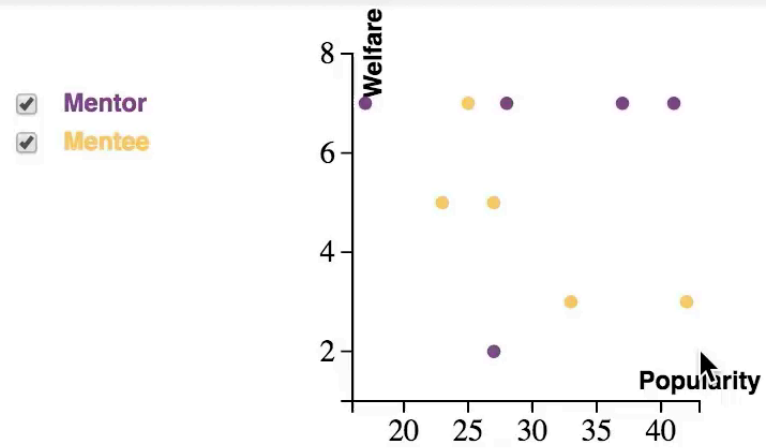
Summary: ●



Number Line View



Scatterplot



Expert Interview

Better than spreadsheet for assignment

More confident in assignment

Learning curve may be steep

Conclusion and Future Work

- Developed a way for users and the system to co-create a matching
- Designed visualizations for matching comparison
- Steep learning curve
 - Improve the usability of the system
- Slow and non-optimal matching recommendation algorithm
 - More efficient algorithm to prune the huge matching search space

Thank You!

Po-Ming Law

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Yixian Zheng

Huamin Qu

