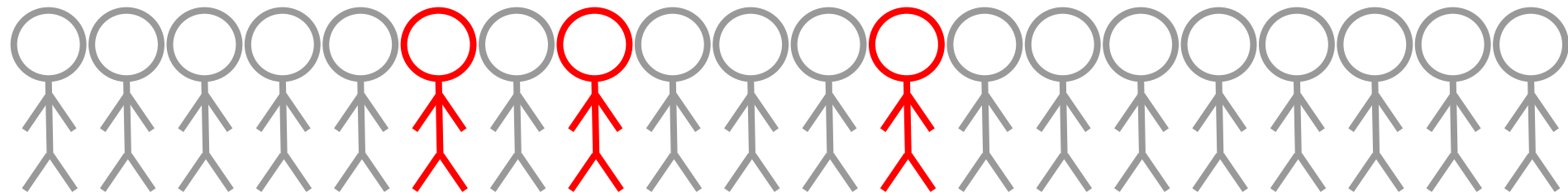


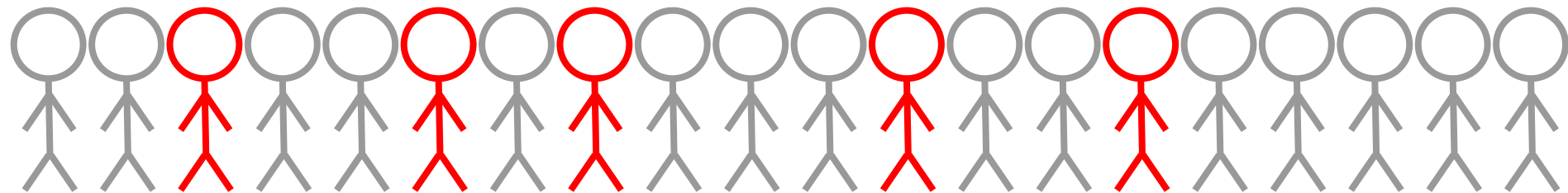
Improving Speed Dating Using Analytics

**Anil Kumar
July 12, 2021**

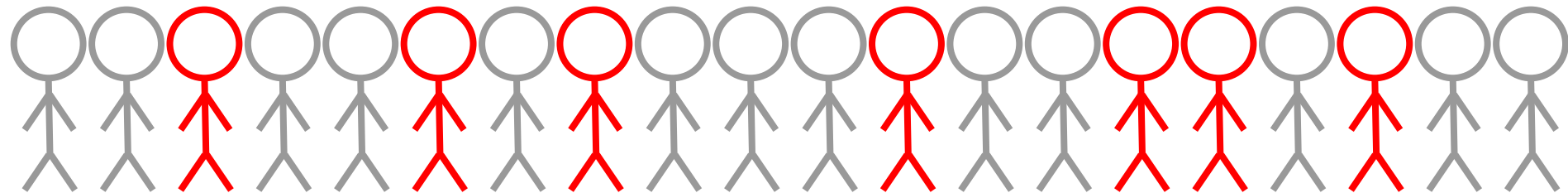
Current Situation



Goal



Even Better



Data and Diagnostic Model

Overview

- **8,378** speed dating encounters
 - **6,998 (83.5%)** non-matches
 - ~17 out of 20
 - **1,380 (16.5%)** matches
 - ~3 out of 20

Features

- **Attractiveness, Intelligence, Sincerity, Funny, Ambition, and Shared Interests**

- Person's rating of partner
- Partner's rating of person
- Person's self-rating
- How important is the trait to person?
- How important is the trait is to partner?

- **Socio-demographic**

- Gender
- Age
- Race
- Are the person and partner the same race
- Importance that partner is the same race
- Importance that partner is the same religion

- **Interests**

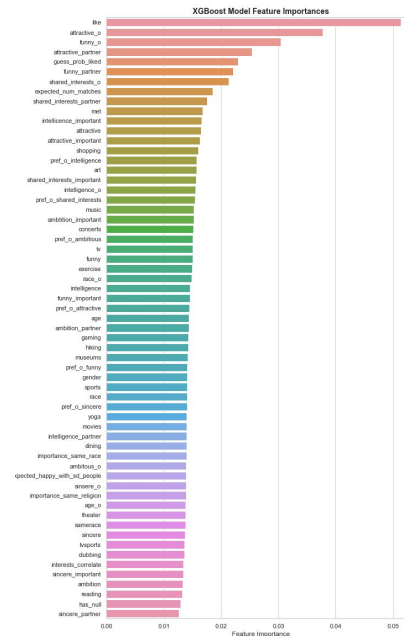
- Sports, TV Sports, Exercise, Dining, Museums, Art, Hiking, Gaming, Clubbing, TV, Reading, Theater, Movies, Concerts, Music, Shopping, Yoga

- **Other**

- How happy does the person expect to be with the people that they meet?
- How much does the person like their partner?
- How likely does the person think it is that their partner likes them?
- How many matches does the person expect to get?
- Have the person and partner have met before?

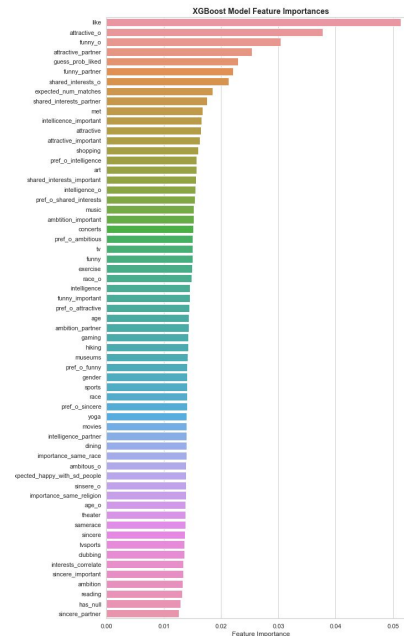
Most Predictive Features

- Created a machine learning model that analyzed the feature impacts on predicting a match
- Features with the most impact
 - Attractiveness
 - Funny
 - Liking each other
 - Shared Interests



Most Predictive Features

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 - Attractiveness
 - Funny
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 - **Shared Interests**



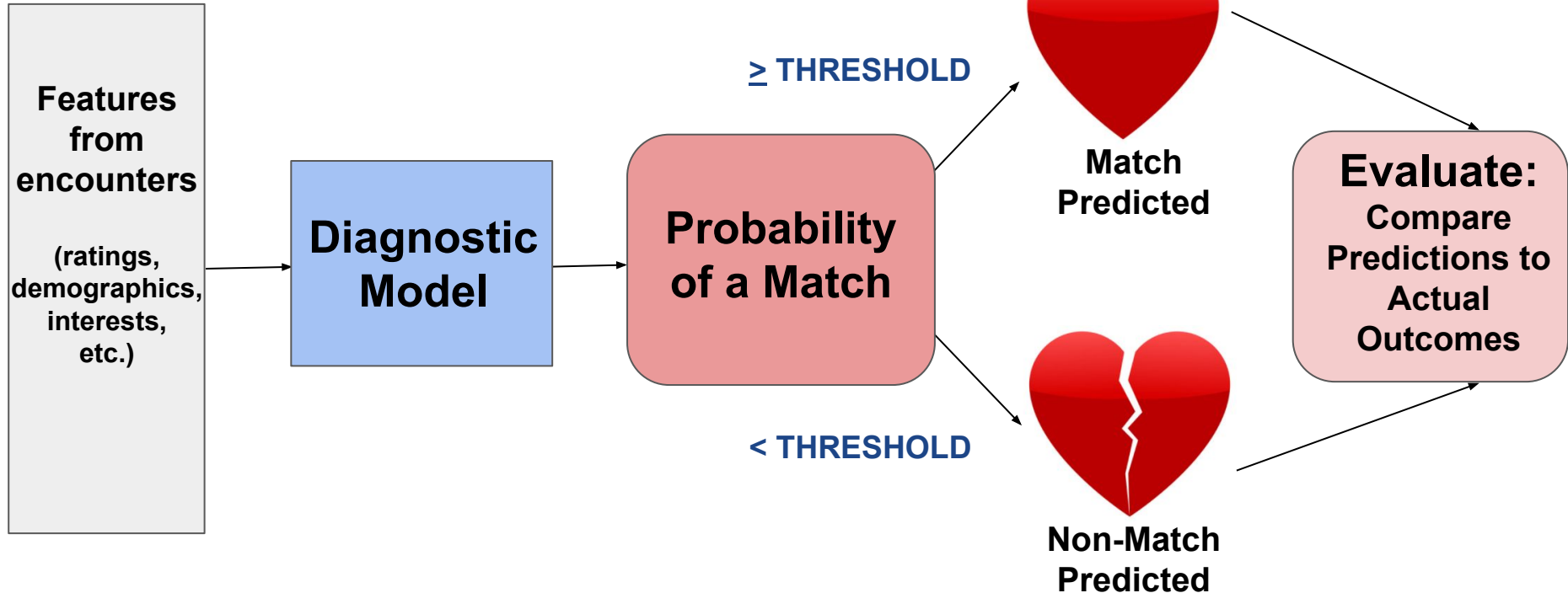
Match Rates Based on Shared Interest Ratings

	Match Rate
Both Rated 0-5	5.3%
One Rated 0-5, Other Rated 6-8	14.2%
One Rated 0-5, Other Rated 9-10	18.7%
Both Rated 6-8	32.1%
One Rated 6-8, One Rated 9-10	44.9%
Both Rated 9-10	60.0%

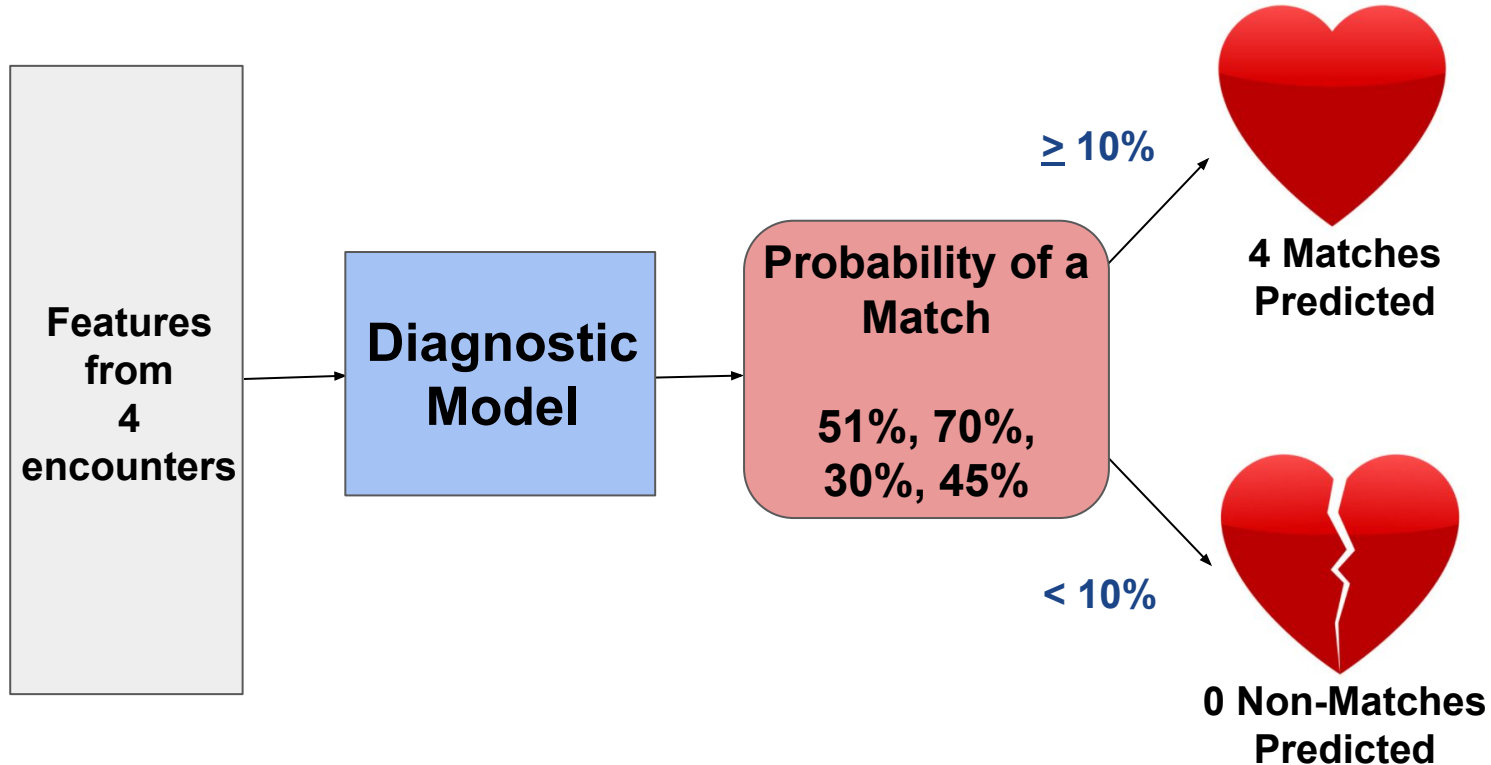
Match Rates Based on Shared Interest Ratings

	Match Rate	% of Encounters	72.2% of encounters
Both Rated 0-5	5.3%	35.5%	
One Rated 0-5, Other Rated 6-8	14.2%	36.7%	
One Rated 0-5, Other Rated 9-10	18.7%	4.3%	
Both Rated 6-8	32.1%	16.8%	
One Rated 6-8, One Rated 9-10	44.9%	5.6%	
Both Rated 9-10	60.0%	1.2%	

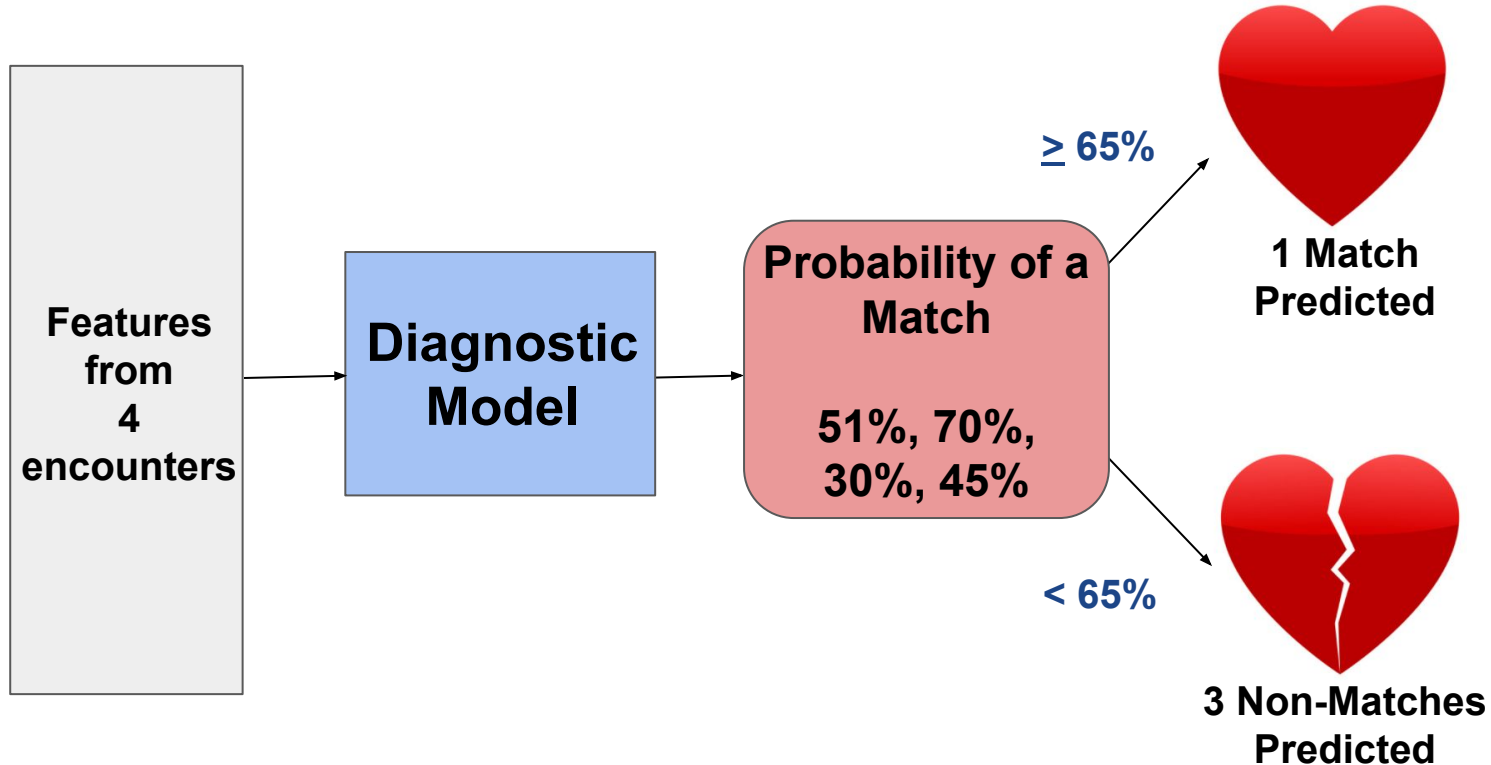
Diagnostic Model - Evaluation



Hypothetical Threshold 10%



Hypothetical Threshold 65%



Results from Diagnostic Model

100 encounters (expect 16-17 matches)	Threshold 10%	Threshold 50%	Threshold 65%
% Encounters predicted as matches	33.7% 33-34 matches predicted	11.2% 11-12 matches predicted	7.3% 7-8 matches predicted

Future Directions

Create Pre-Event Survey

- **Use diagnostic model results to create a pre-event survey that collects data on the most predictive features**
 - Shared Interests
 - Preferences for a partner
 - If repeat customer, use data from previous events
- **Appropriate length**
 - Reduce data quality issues (i.e. missing or invalid entries)

Create Prediction Model

- **Create model to predict matches based on pre-event data**
- **Arrange clients into events that give them the best chances at finding matches**
 - Enhanced experiences
 - Referrals
 - Reviews
 - Reputation
 - Repeat customers

Longitudinal Data Collection

- **Clients that matched**
 - Did matches last in long-term romantic relationships or friendships?
- **Clients that didn't match**
 - What were the issues?
 - What can we do to help them find matches?

Customer Lifetime Value

CLTV = *Customer Value x Average Customer Lifespan*

CLTV = \$30/event x 1 event = **\$30**

Over the course of 1,000 customers: **\$30,000**

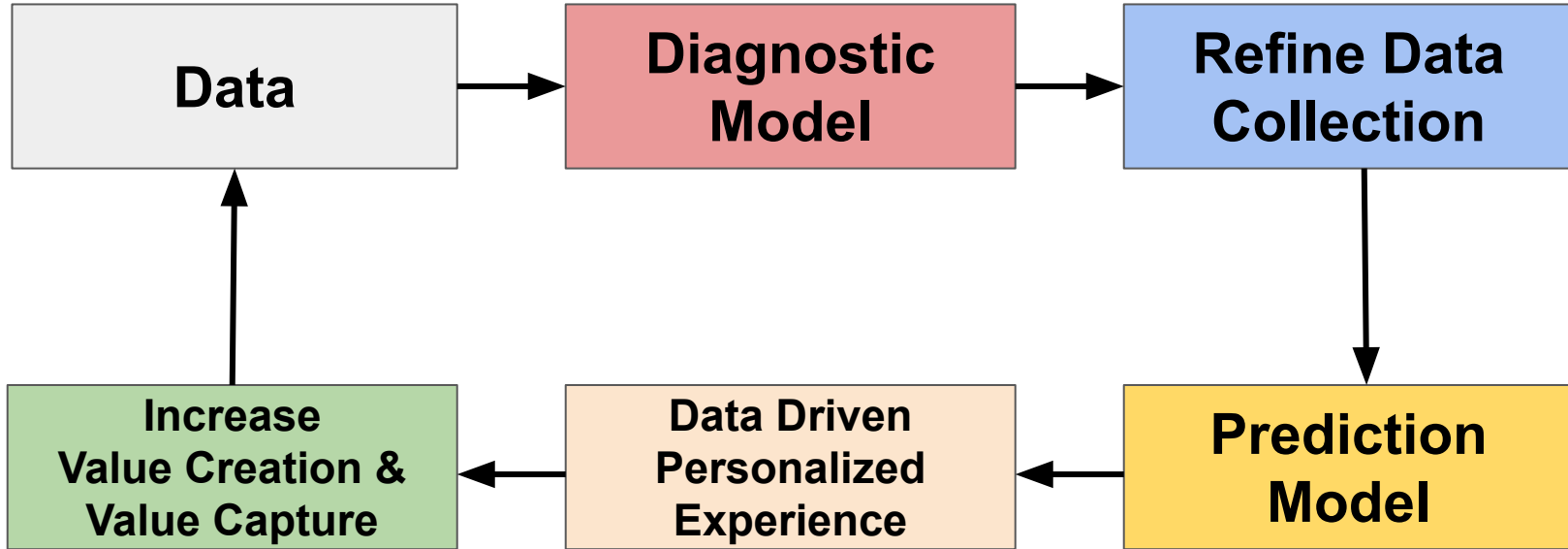
If 500 out of 1,000 customers elect to attend 2 events:

Average Customer Lifespan = 1.5 events

CLTV = \$30/event x 1.5 events = **\$45**

Over the course of 1,000 customers: **\$45,000**

Summary



Questions?