

# DiscoverTED

## A TED Talk Recommender

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# TED

**IDEAS** WORTH **SPREADING**

# Overview

Recommend talks to learn both deeper and wider

## Difficulties

- Sparse Data → model selection
- Cold Start for New Users → required user input
- Evaluation

# “Informative” Talks for users interested in “machine learning big data”



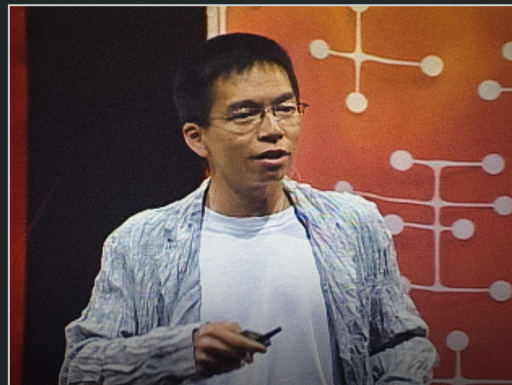
The jobs we'll lose  
to machines -- and  
the ones we won't

Anthony Goldbloom  
(TED2016)

**TALKS FOR DEEPER**

How to fool a GPS

Todd Humphreys  
(TEDxAustin)



My journey in  
design

John Maeda  
(Serious Play 2008)

**TALKS FOR WIDER**

The beauty of data  
visualization

David McCandless  
(TEDGlobal 2010)



# Data

## Talk Data

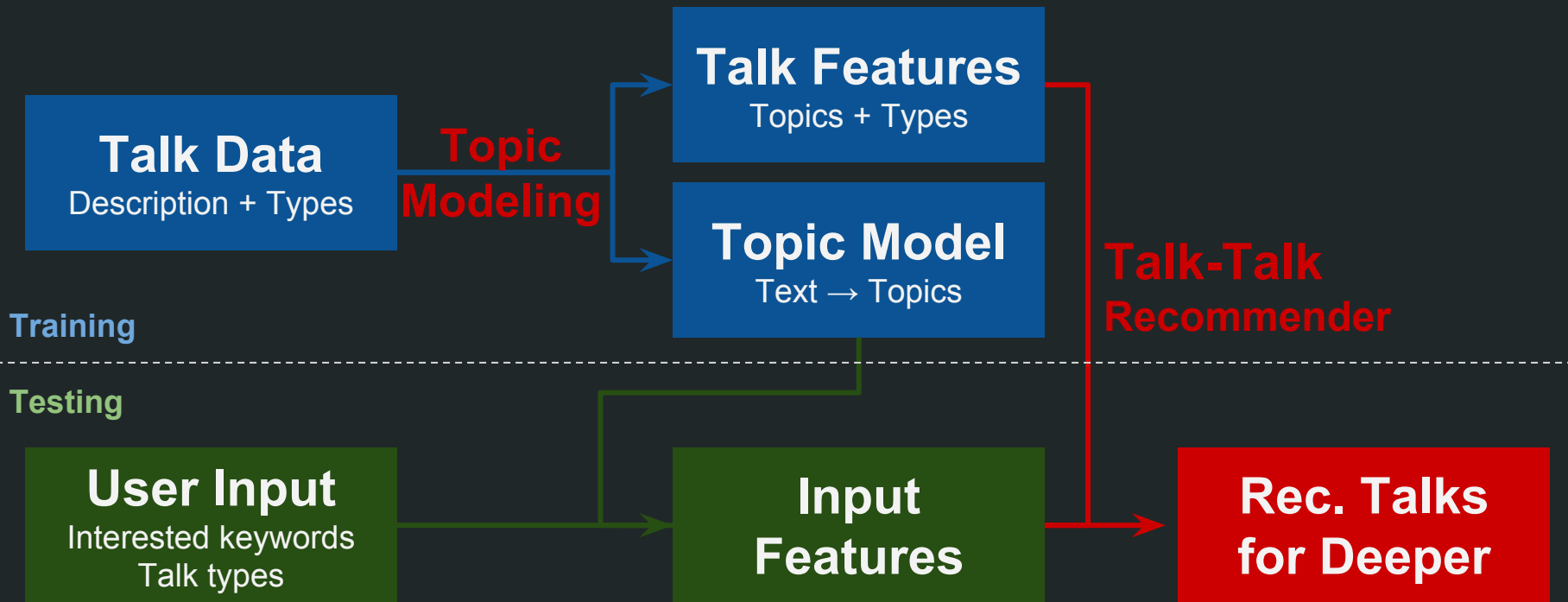
- Fields: titles, tags, description, talk types, ...
- Total 2,318 talks (1,201 talks are favorited)
- On average, there are 84.3 users per favorited talk
- Source: Scraped from **TED.com**

## User-Talk Data

- Fields: users, favorite talks
- Total 12,401 users. 6,449 active users with 4+ favorite talks (52% of total)
- On average, there are 9.3 favorite talks per user
- Source: **IDIAP**

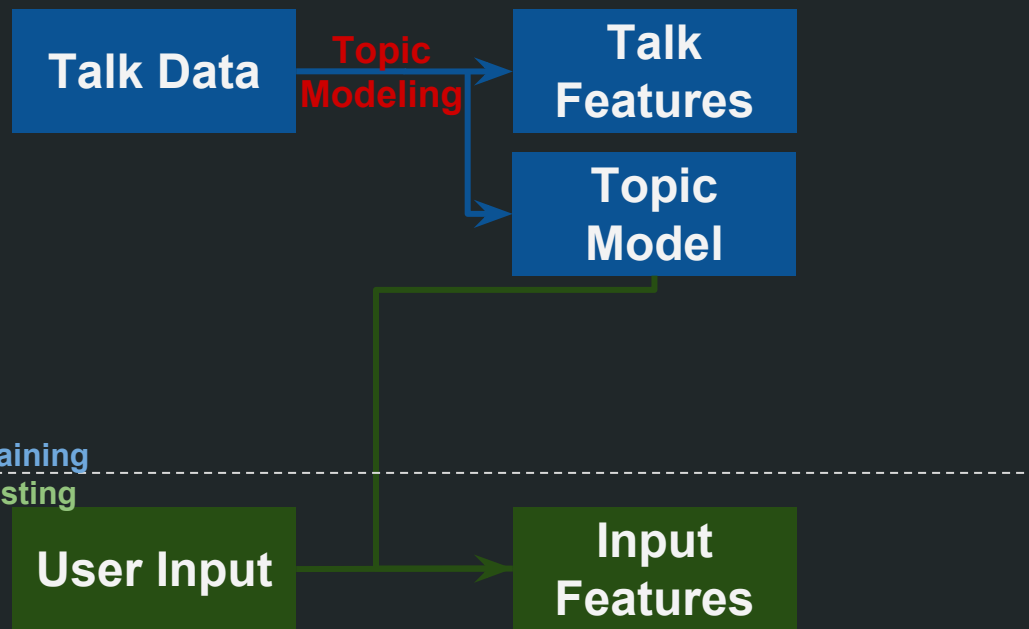
# Learn Deeper: Talk-Talk Recommender

Recommend talks closest to a user's interested keywords and talk types



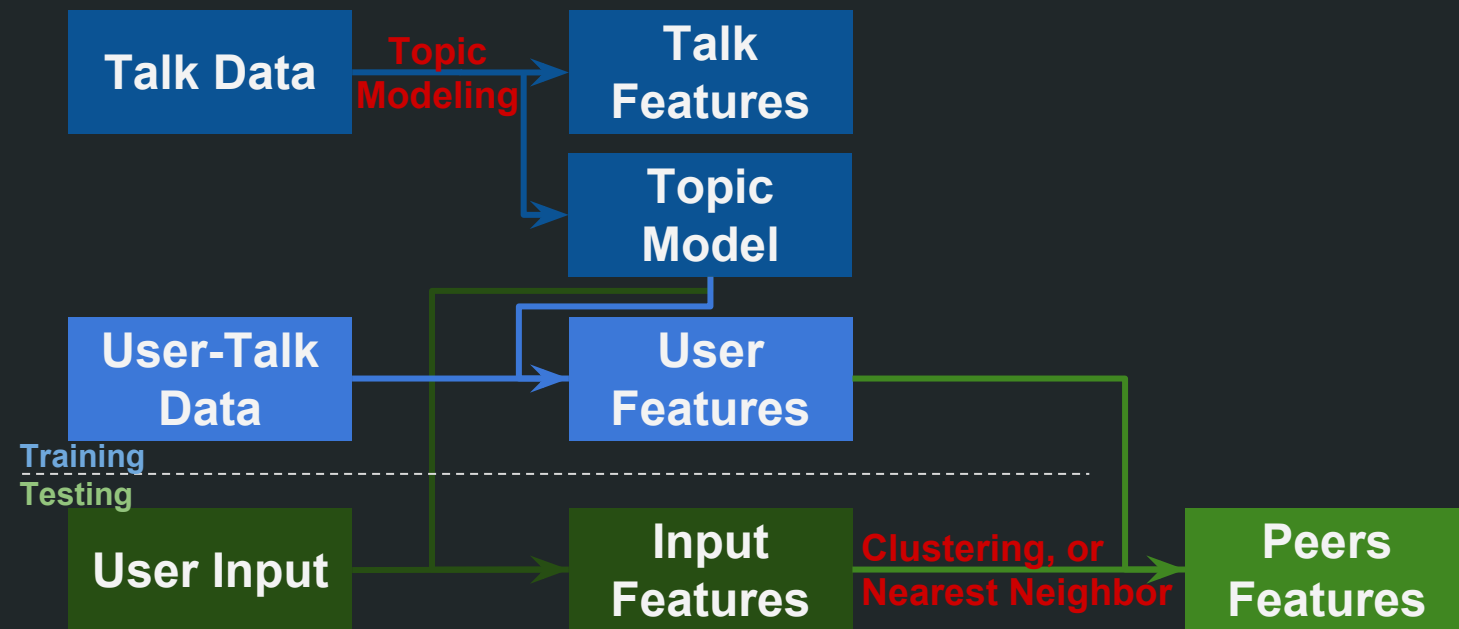
# Learn Wider: User-User Recommender

Recommend talks in peers' next favorite topics and closest to a user's interests



# Learn Wider: User-User Recommender

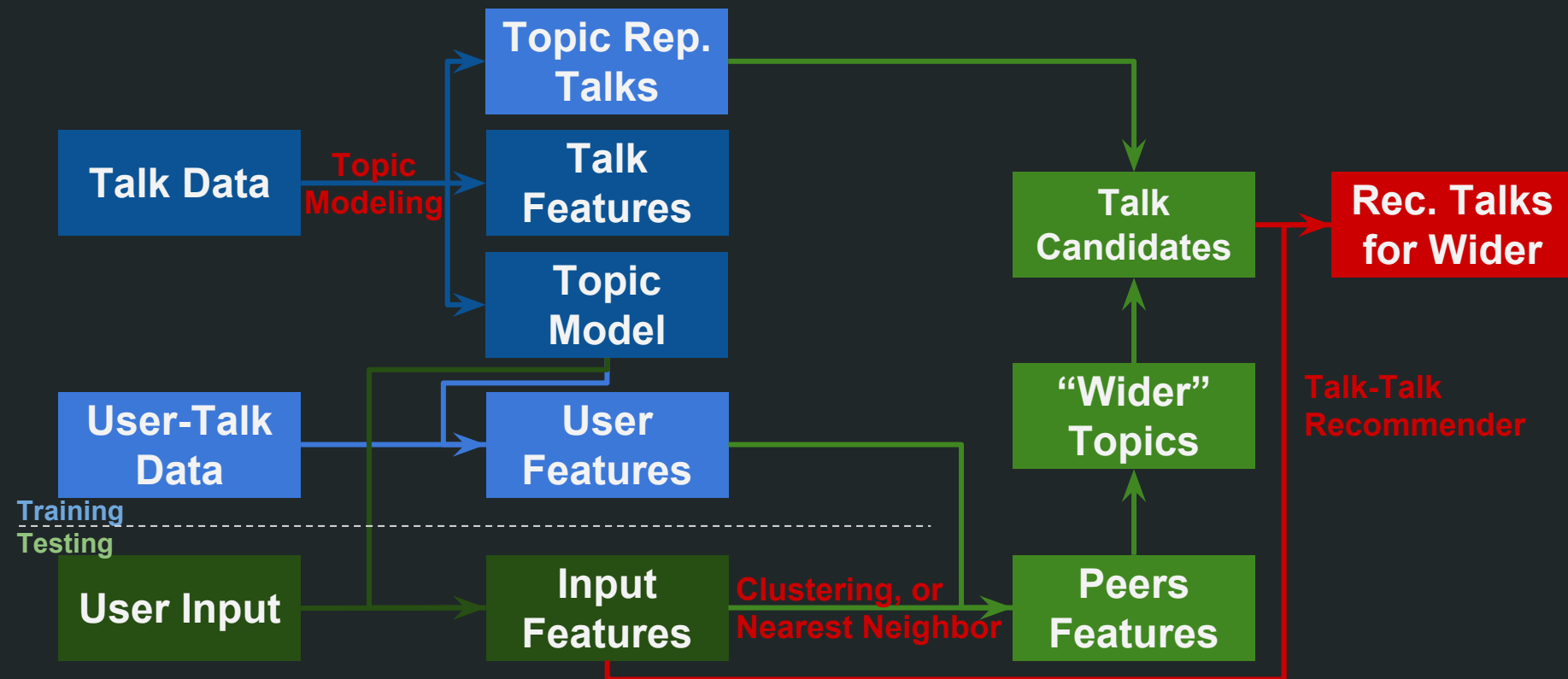
Recommend talks in peers' next favorite topics and closest to a user's interests





# Learn Wider: User-User Recommender

Recommend talks in peers' next favorite topics and closest to a user's interests



# Model Selections

**Natural Language Processing (NLP)**

**Latent Dirichlet Allocation (LDA)**

**Nearest Neighbor**

**Less preferred models:**

- **Non-negative matrix factorization (sparse data)**
- **Graphlab matrix factorization (sparse data)**
- **K-mean clustering (inter- vs. intra- distances)**

# Evaluation

**Compared to random selections,  
are recommended talks closer to a user's favorite talks?**

→ Random: 1.01 | Deeper Only: 0.84 | Deeper+Wider: 0.89 (smaller = better)

**Compared to “deeper” topics only,  
do “wider” topics help cover a user's favorite talks?**

→ Deeper Only: 1.17 | Deeper+Wider: 1.11 (smaller = better)

# Future Work

Transcript is noisy but informative

More usage information for better prediction



## Acknowledge

Nikolaos Pappas, Andrei Popescu-Belis, "Combining Content with User Preferences for TED Lecture Recommendation", 11th International Workshop on Content Based Multimedia Indexing, Veszpré Hungary, IEEE, 2013 [PDF](#) [Bibtex](#)

**THANK YOU &  
HAPPY LEARNING !**

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<https://github.com/liviachang/DiscoverTED>