Improving Users' Demographic Prediction via the Videos They Talk about

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- Mean Girls, Pretty Woman, The Devil Wears Prada
- House of Cards, Mission Impossible, NBA





- "Will the Big Bang Theory last into the next century?"
- "Sheldon is so cool, I love him!"

"Jim Parsons was nominated for another Emmy Award"



**YOUKU**优酷

iQIYI爱奇艺

搜狐视频

腾讯视频

V.00.COM



#### 《星球大战:原力觉醒》彩蛋和花絮总汇

三刷影片后,确认了不少彩蛋,下面尽量按照时间顺序排列。 2187 Finn的暴风兵编号为FN-2187,2187是《新 希望》中帝国军关押Leia公主的牢房号码。这个数字本来就是一个彩蛋,《21-87》是由加拿大先锋导演Arthur…

2

豆瓣

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### Data

Normal User

Verified User

User followed by Verified User



Attribute	<b>Completion Rate</b>	Categories		
Gender	95.019%	Male, Female		
Age	18.604%	Teenage (<18), Youngster (18-24), Young (25-34), Mid-age(>34)		
Education BG	17.443%	University, Non-University		
Marital Status	2.203%	Single, Non-Single		

**Table 1:** Demographic attributes and corresponding categories

### Data

	Video	Actor	Keyword
Variety show	344	1007	2925
Movie	306	741	2049
TV	197	515	1302
Total	847	1422	4094

**Table 2:** Statics of video relevant information (There is anoverlap between the three collections of actors and keywords.)

# Discover Indirect Relationships

- Unobvious relationship
  - N\*M pairs
- Direct relationship
  - User 2, "Will the Big Bang Theory last into the next century?"
- Indirect relationship
  - User 3 posts, "Sheldon is so cool, I love him!"











# **Discover Indirect Relationships**

### Step 1

 $P(v_n) = \frac{num(users \ watched \ the \ n_{th} \ video)}{num(users)}$   $P(w_{ni}|v_n) = \frac{num(users \ watched \ the \ n_{th} \ video \ and \ mentioned \ the \ n_{th} \ keyword)}{num(users \ watched \ the \ n_{th} \ video)}$   $P(a_{nj}|v_n) = \frac{num(users \ watched \ the \ n_{th} \ video \ and \ mentioned \ the \ n_{jth} \ actor)}{num(users \ watched \ the \ n_{th} \ video)}$ 

#### Step 2

$$\begin{split} P(v_n | W_m, A_k) &= \frac{P(W_m, A_k | v_n) * P(v_n)}{P(W_m, A_k)} \\ &= \frac{\prod_{w_{ni} \in W_m} P(w_{ni} | v_n) * \prod_{a_{nj} \in A_k} P(a_{nj} | v_n) * P(v_n)}{P(W_m, A_k)} \end{split}$$

# **Discover Indirect Relationships**



#### Two Baseline Model

Two Indirect Relationship Based Model



### **Discriminant Model**

- Matrix Factorization<sup>1</sup>, K=20
- LR<sup>2</sup>, SVM<sup>2</sup>, GBDT<sup>3</sup>



1 libFFM 2 liblinear 3 XGBoost

### Generative Model

- Calculate video demographic tendency
- Calculate user demographic attribute
- Smooth the result

# Evaluation

		Precision	Recall	F1	AUC
Gender	Dis-Baseline	0.720	0.714	0.717	0.730
	Dis-Model	0.786	0.779	0.783	<b>0.812</b> ↑ 11.2%
	Gen-Baseline	0.701	0.687	0.694	0.707
	Gen-Model	0.799	0.802	0.801	<b>0.825</b> ↑ 16.7%
	Dis-Baseline	0.569	0.541	0.554	*
Age	Dis-Model	0.642	0.653	<b>0.648</b> ↑ 16.8%	*
2	Gen-Baseline	0.529	0.504	0.516	*
	Gen-Model	0.663	0.645	<b>0.654</b> ↑ 26.7%	*
Education BG	Dis-Baseline	0.707	0.716	0.711	0.730
	Dis-Model	0.788	0.801	0.795	<b>0.809</b> ↑ 11.1%
	Gen-Baseline	0.680	0.659	0.669	0.690
	Gen-Model	0.790	0.808	0.799	<b>0.812</b> ↑ 17.7%
Marital Status	Dis-Baseline	0.565	0.549	0.557	0.571
	Dis-Model	0.657	0.640	0.648	<b>0.659</b> ↑ 15.4%
	Gen-Baseline	0.572	0.550	0.560	0.581
	Gen-Model	0.682	0.691	0.687	<b>0.696</b> ↑ 19.8%

Table 3: Prediction accuracy based on users' video describing words. Classes have been balanced.

### Evaluation



### Evaluation



Figure 5: Results of Fusion Model evaluation (Macro-F1).

# Conclusion

- Our motivation is that user's video related behavior is usually under-utilized on demographic prediction tasks.
- With the help of third-party video sites, we detect the direct and indirect relationships between users and video describing words, and demonstrate this effort can improve the accuracy of users' demographic predictions.
- To our knowledge, this is the first work which explores demographic prediction by fully using users' video describing words.
- This framework has good scalability and can be applied on other concrete features, such as user's book reading behaviors and music listening behaviors.

# Thanks!