

Get to the Bottom: Causal Analysis for User Modeling



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Treated

High

High

High

High

Low

Low

Low

High

Motivation

Weather affects our mood and behavior, and through lacksquarethem many aspects of our lives



When it is sunny, people become happier and smile



When it rains, people get depressed

Dataset

- Australia-wide dataset of 10M catch-up TV viewing ulletevents from February to September 2012
 - More than 0.6M users all over Australia
 - More than **11K** unique programs



We present the first causal analysis of *how weather* affects TV content consumption patterns

Application

Causal analysis results can be used for better adaptive personalized recommendations of TV and video

Causality

- Setup:
 - Units $i \in |n|$ are TV viewing events
 - *Treatment* at event *i*, $T_i \in \{0, 1\}$, is an indicator of target weather (eg, precipitation) at *i*
 - Potential outcomes at $i_i(Y_i(0), Y_i(1))$, are indicators of watched content (eg, Drama programs) in treatment and control
- Goal: Estimate average treatment effect on treated (ATT)

$$ATT = \mathbb{E}_{i:T_i=1}[Y_i(1)] - \mathbb{E}_{i:T_i=1}[Y_i(0)]$$
Treatment group
Treatment outcome
Control outcome

Problem: Outcome of not treating on treated

				Weather attribute	
Category	Frequency	Category	Frequency	Temperature	
Drama	19.51%	Pre-school	19.31%	Feels-like temperature	
Children	17.01%	Comedy	11.37%	Wind speed	
Docs	10.61%	Lifestyle	8.06%	Cloud cover	
Panel	5.95%	News	4.10%	Pressure	
Arts	2.69%	Education	0.58%	Humidity	
Kids	0.50%	Sport	0.24%	Visibility	
Indigenous	0.05%	Shop	0.02%	Precipitation	

Empirical Results

Example: When pressure and precipitation is low (rain), the frequency of watching Dramas drops significantly. Possible explanation: children cannot play outside and watch TV at home





Significant changes in ATTs of all

- weather treatments on 8 most popular genres for different treated groups.
- Significant increase in red, significant decreases in blue, *insignificant effects*



 $\mathbb{E}_{i:T_i=1}[Y_i(0)]$, is not observed ($\mathbb{E}_{i:T_i=0}[Y_i(0)]$ does not approximate it because the treatments are not assigned at random)

- Key challenge: Balance the distributions of treated and control events
- Solution: Nearest neighbor matching on the covariates of events, and then

$$\operatorname{ATT} \approx \frac{1}{n_T} \sum_{i:T_i=1} (Y_i(1) - Y_{\pi(i)}(0))$$

Matched control event to treated event *i*

Covariate is the profile of the user and the time at *i* \bullet

Our causal findings confirms that there are causal relations between weather and users' TV watching patterns