19 September 2017

The Best Privacy Defense is a Good Privacy Offense Obfuscating a Search Engine User's Profile

Jörg Simon Wicker S The University of Auckland J

Stefan Kramer

Johannes Gutenberg University Mainz









Privacy Defence

- Privacy on the internet is an important and unsolved issue
- Privacy preserving data mining addresses this from a service provider perspective
- From a user's perspective, what can the user do to ensure the protection of his or her data?









davidsbeenhere.com/2015/12/30/10-places-you-must-visit-in-macedonia/ -

Dac 20, 2015 - Other must do's in Obrid include seeing the freecose of Saint Sofia (Suati Sofia) Chu



houses and monuments, and tourism is predominant. It is located southwe

Search Engines

Google	what to visit in ma	cedonia		ų) Q					
	All Images Maps News Videos More			Settinas	Tools					
	About 40,100,000 res	ults (0.94 seconds)		Ь	ohrid maced	onia				م
	Visit Macedonia Ad www.traveltheur Discover The Charms Financially Insured - H Services: Group Tours	 Small Group & Tai iknown.com/Macedonia/Ta Of Macedonia - Highly Rate ighly Recommended - Stuni , Adventure Holidays. Privat 	Iormade Tours ours ▼ ed Tours - Trusted Tour Agenc ning Destinations - Fly From A e Tours, Tallormade Tours	y ny	Web Ima	ages Videos	News	My sav	ves	
	The Balkan Odys Balkan Explorer	sey Tour The Balkans	Balkans Archaeology		969,000 RESUL	.TS Date 🔻	Language	e ▼ Re	gion 👻	
	Macedonia (FYR	OM) / Top sights			Car Renta Prices	al in Ohrid, N	lacedonia	- Best Car	r Rental	1
			all the second		Ad · www.Ren Search, Comp Brands: Toyota "Travel Standa	talcars.com/Mace are and Save Using I, Ford, Chevrolet, C rd of Excellence - V	donia/Ohrid the World's Bigg Chrysler, Jeep, H VebAward 2016"	gest Online Car I yundai, Mercede – WebAward 20	Hire Service. es)16	
	Lake Ohrid Lakes, monasteries, and churches	Matka Canyon Lakeside hotel with a wine cellar	Church of St. John at Kaneo Grand historic church with lake views	R C	Ohrid - W https://en.wiki	ikipedia pedia.org/wiki/Ohri	d -			
	I More Macedor	ia (FYROM) sights			Overview	Contents	Name	History	Demographics	
	10 Places You M	fust Visit in Macedor	nia - David's Been He	re	largest city on inhabitants as	Lake Ohrid and the of 2002. Ohrid is no	eighth-largest cit table for once ha	iving had 365 ch	, with over 42,000 nurches, one for each	1

davidsbeenhere.com/2015/12/30/10-places-vou-must-visit-in-macedonia/ -



Google	what to visit in macedonia		.♥ Q			
	All Images Maps News Video	s More Settin	as Tools			
	About 40,100,000 results (0.94 seconds)	Ь	ohrid macedonia	م		
	Visit Macedonia - Small Group & Ta Marking www.traveltheunknown.com/Macedonia/T Discover The Charms Of Macedonia - Highly Rat Financially Insured - Highly Recommended - Stur Services: Group Tours, Adventure Holidays, Prive	ilormade Tours burs • xed Tours - Trusted Tour Agency ining Destinations - Fly From Any te Tours, Tailormade Tours	Web Images Videos News My saves			
	The Balkan Odyssey Tour The Balkan Explorer Balkans	# Balkans Archaeology	969,000 RESULTS Date - Language - Region -			
	Macedonia (FYROM) / Top sights		Car Rental in Ohrid, Macedonia - Best Car Rental			
		all and	PTICES Ad - www.Rentalcars.com/Macedonia/Ohrid Search, Compare and Save Using the World's Biggest Online Car Hire Service. Brands: Toyota, Ford, Chevrolet, Chrysler, Jeep, Hyundai, Mercedes			
	Laka Obrid Matka C	-OO! ohrid macedonia	X Search			
	Lakes, monasteries, Lakeside Web wine cella	Ads related to: ohrid	. macedonia .nia - Top 10 Hotels (with Prices) - tripadvisor.co.pz			
	More Macedonia (FYRON Answers	www.tripadvisor.co Save money & book Millions of hotel revi	advisor.co.nz/ wy & book with TripAdvisor® (the world's largest travel website). hotel reviews - Easy price comparison - Candid traveller photos Dhrid Municipalit			
	10 Places You Must Visit i davidsbeenhere.com/2015/12/30	380 Hotels in C www.booking.com/	Jhrid - Lowest price guarantee - booking.com Ohrid-Hotels ▼ The city is rich in Dirid-Hotels ■	42,000 1e for each picturesque		



Google	what to visit in macedo	onia		پ م			
	All Images Maps	News Videos More	Settinas	Tools			
	About 40,100,000 results (0.	.94 seconds)	Ь	ohrid macedonia			
	Agit Macedonia - Sn Agit www.traveltheunknow Dis ver The Charms Of Ma Financially Insured - Highly I Services: Group Tours, Adve	nall Group & Tailormade T vn.com/Macedonia/Tours acedonia - Highly Rated Tours - Tru: Recommended - Stunning Destinatie enture Holidays, Private Tours, Tailor	ours sted Tour Agency ons - Fly From Any made Tours	Web Images Videos News My saves			
	The Balkan Odyssey Balkan Explorer	Tour The Balkans Balkans Archaeology		969,000 RESULTS Date • Language • Region •			
	Macedonia (FYROM) / Top sights			Car Rental in Ohrid, Macedonia - Best Car Rental Prices Ad - www.Rentalcars.com/Macedonia/Ohrid Search, Compare and Save Using the World's Biggest Online Car Hire Service. Brands: Toyota, Ford, Chevrolet, Chrysler, Jeep, Hyundai, Mercedes			
	Lake Obrid	Aatka C NEW ZEALAND	ohrid macedonia	X Search			
	Lakes, monasteries, La and churches wi	akeside Web	Ads related to: ohrid macedonia				
	Images Onn'to Mi News Save money Answers Millions of ho			Ia - Iop 1U Hotels (with Prices) - tripadvisor.co.nz y/ * y/ * ThirtpAdvisor® (the world's largest travel website). pry Demographics s - Easy price comparison - Candid traveller photos Dhrid Municipality. It is the			
	10 Places You Must Visit i davidsbeenhere.com/2015/12/30 Sports		380 Hotels in Of www.booking.com/O	rrid - Lowest price guarantee - booking.com hrid-Hotels ▼ The city is rich in picturesque			











Personalised Advertisement

- Which ad is displayed depends on
 - 1. The submitted query
 - 2. The user profile
- Ads are assigned to categories
- Users are assigned to categories







Privacy Offense

To implement a method to defend privacy, we need





Privacy Offense

To implement a method to defend privacy, we need
 A way to measure the privacy, i.e. objective function



Privacy Offense – Goal

Objective function:

$$\sigma(\kappa_i, \boldsymbol{P}) = \sum_{\boldsymbol{p}_j \in \boldsymbol{P}, \kappa_j \in \boldsymbol{T}} \boldsymbol{p}_j \boldsymbol{d}_{\boldsymbol{T}}(\kappa_i, \kappa_j)$$

- User interest category κ, distribution of probabilities P, category tree T, tree distance d_T
- Score σ is the weighted distance between user interest category and current category the user is assigned to





Privacy Offense

To implement a method to defend privacy, we need

- A way to measure the privacy, i.e. objective function
- Method to use the feedback (ads)



Category Prediction of an Ad

- Search engines provide example queries for each category
- Use sample queries of category tree as input and train independent classifiers – one for each category
- Classifiers can be applied to queries, as well as any other text
- Predictions on ads work very well due to similar structure of the text







Privacy Offense

To implement a method to defend privacy, we need

- A way to measure the privacy, i.e. objective function
- Method to use the feedback (ads)
- A set of actions



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_{-select}(\kappa_r \in K)$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$ Sibling: $a_{sibling}(T, \kappa_{ref}) = sibling(\kappa_{ref} \in K)$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$ Sibling: $a_{sibling}(T, \kappa_{ref}) = sibling(\kappa_{ref} \in K)$ Most general: $a_{general}(T, \kappa_{ref}) = max_parent(\kappa_{ref} \in K)$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$ Sibling: $a_{sibling}(T, \kappa_{ref}) = sibling(\kappa_{ref} \in K)$ Most general: $a_{general}(T, \kappa_{ref}) = max_parent(\kappa_{ref} \in K)$ Most specialized of sibling: $a_{specialized}(T, \kappa_{ref} = lowest_child(all_siblings(\kappa_{ref} \in K))$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$ Sibling: $a_{sibling}(T, \kappa_{ref}) = sibling(\kappa_{ref} \in K)$ Most general: $a_{general}(T, \kappa_{ref}) = max_parent(\kappa_{ref} \in K)$ Most specialized of sibling: $a_{specialized}(T, \kappa_{ref} = lowest_child(all_siblings(\kappa_{ref} \in K)))$ Distance-based: $a_{dist}(T, \kappa_{ref}) = \kappa_r : \forall \kappa_t \in K, d(\kappa_r, \kappa_{ref}) \ge d(\kappa_t, \kappa_{ref})$



Definition of actions to choose one category κ in the set of categories *K* using category tree *T* based on reference category κ_{ref}

Random: $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$ Same: $a_{same}(T, \kappa_{ref}) = \kappa_{ref}$ Sibling: $a_{sibling}(T, \kappa_{ref}) = sibling(\kappa_{ref} \in K)$ Most general: $a_{general}(T, \kappa_{ref}) = max_parent(\kappa_{ref} \in K)$ Most specialized of sibling: $a_{specialized}(T, \kappa_{ref} = lowest_child(all_siblings(\kappa_{ref} \in K)))$ Distance-based: $a_{dist}(T, \kappa_{ref}) = \kappa_r : \forall \kappa_t \in K, d(\kappa_r, \kappa_{ref}) \ge d(\kappa_t, \kappa_{ref})$

- Jaccard Distance
- Normalized Mutual Information

















































 $a_{random}(T, \kappa_{ref}) = random_select(\kappa_r \in K)$









on	line shopp	ing
Google	anline shapping	<u>ه</u> و
	# Progen News Videos Dooks More	Semip Ton
	And Standard Human () 13 warms) Cardonics Carbon () Weners 15 March (Carbon) 45 Fashib Information () Weners 15 March () Carbon) 45 Fashib Information () Weners 15 March () Weners 15 March () March () Weners 15 March () Weners 15 March () March () Weners 15 March () Weners 15 March () Weners 15 Weners 15 March () Weners 15 March () Weners 15 Weners 15 March () Weners 15 March () Weners 15 Weners 15 March () Weners 15 March () Weners 15 Weners 15 March () Weners 15 March () Weners 15 March () Weners 15 March () March () Weners 15 March () Weners 15 March () Weners 15 Marc	on Online ene every der tere dessen. Ing EnExy NZ men & lick heiters an well









IGU













Experiments

- Users are given one interest category and either
 - Use the proposed method, or
 - Submit queries from random categories, or
 - Submit queries from the category that is the furthest away from their interest category
- All users submit in 10% of the cases random queries from their interest category

20 categories where used:

antiques and collectibles car video cosmetic procedures desktop computers erectile dysfunction game systems and consoles make up and cosmetics real estate listings timeshares and vacation properties vitamins and supplements bicycles and accessories computer components dating and personals drugs and medications family laptops and notebooks motorcycles sexual enhancement toys weight loss



Results



All users, queries not in interest category





SCIENCE

Conclusion

Does it work?





SCIENCE

Conclusion

Does it work? Maybe





Does it work?

- Maybe
- But: We can trigger and see a reaction





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category
 - Discard more aspects of the search engine's model, e.g., time and date of query





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category
 - Discard more aspects of the search engine's model, e.g., time and date of query
- Future work





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category
 - Discard more aspects of the search engine's model, e.g., time and date of query
- Future work
 - More sophisticated model





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category
 - Discard more aspects of the search engine's model, e.g., time and date of query
- Future work
 - More sophisticated model
 - Use more feedback than just the ads





- Does it work?
 - Maybe
 - But: We can trigger and see a reaction
- Simplified model
 - Only one interest category
 - Discard more aspects of the search engine's model, e.g., time and date of query
- Future work
 - More sophisticated model
 - Use more feedback than just the ads
 - Extend the use beyond search engines



Thank you for your attention! Any questions?

https://joerg-wicker.org