

OB-PWS: Obfuscation-Based Private Web Search

Ero Balsa, Carmela Troncoso and Claudia Diaz

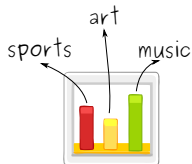
ESAT/COSIC, IBBT - KU Leuven

Wednesday, 23 May 2012

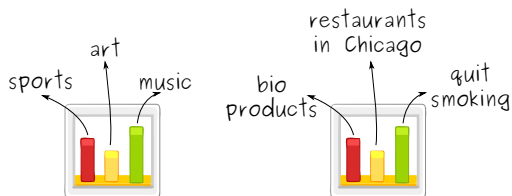
COSIC



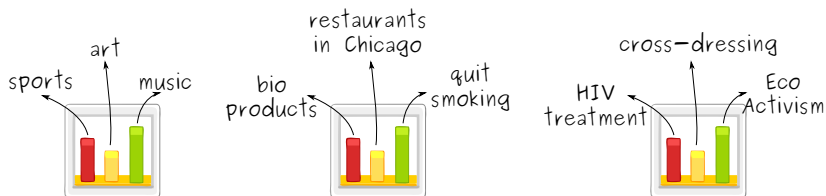
The Privacy Problem



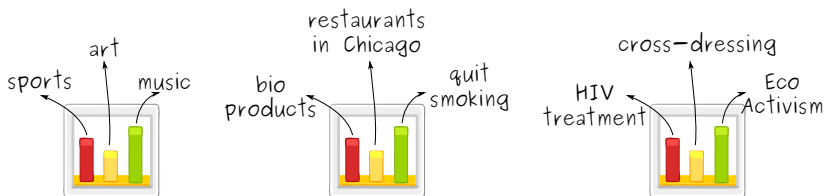
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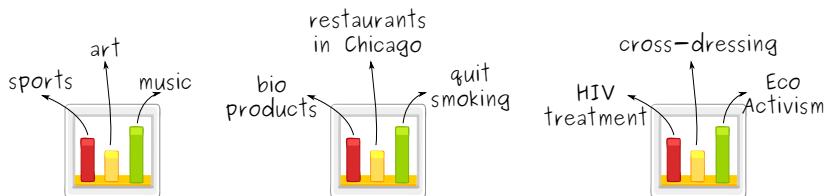


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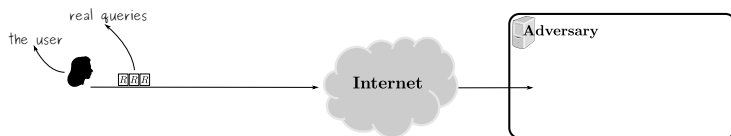
- Some solutions:

- Anonymous communications
- PIR
- **OB-PWS** \Rightarrow Prevent **profiling** and provide query **deniability**.

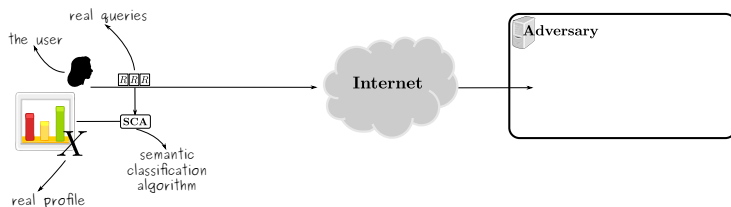
Our contribution

- General model.
- Evaluation framework
 - ⇒ with relevant privacy properties (details in the paper).
- Analysis of 6 existing systems (4 in this talk).

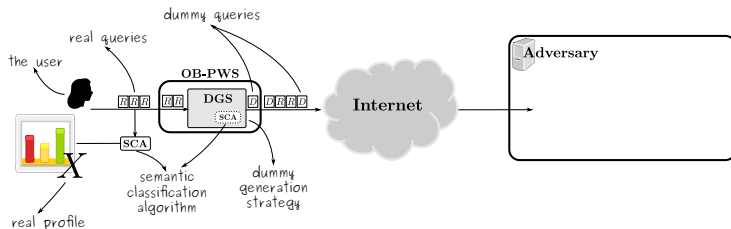
An abstract model for OB-PWS



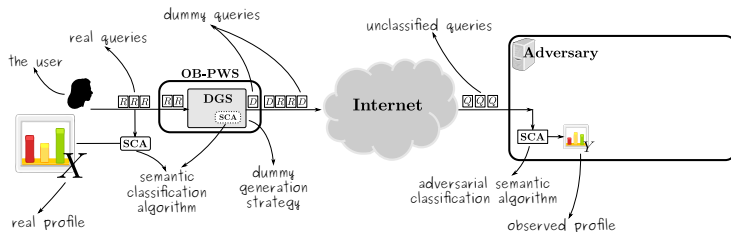
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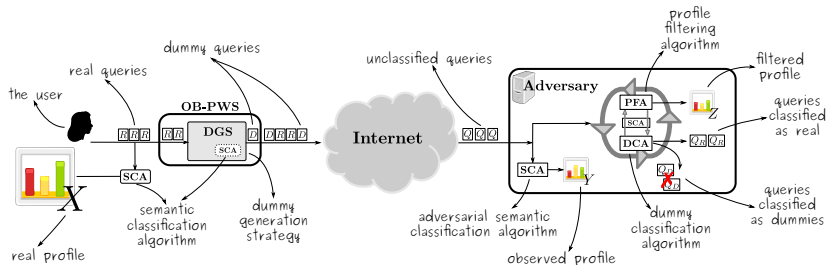
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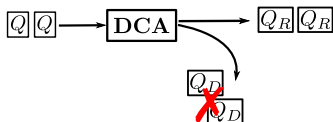
An Evaluation framework for DGS

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Query-Based Analysis

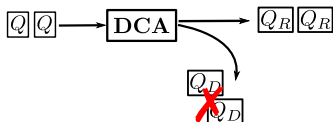


- Exploit vulnerabilities in the DGS to distinguish real from dummy queries.

An Evaluation framework for DGS

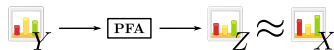
A dual analysis is required:

Query-Based Analysis



- Exploit vulnerabilities in the DGS to distinguish real from dummy queries.

Profile-Based Analysis



- Exploit vulnerabilities in the DGS to filter observed profile and recover the real profile.

GooPIR $h(k)$ -Private Information Retrieval

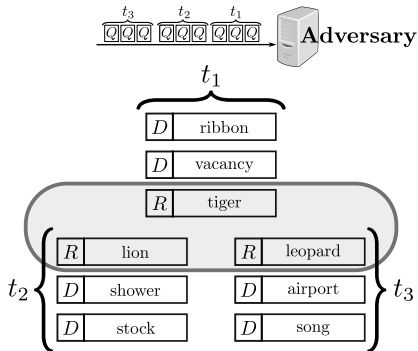
from Privacy-Uncooperative Queryable Databases [1]

- A k -anonymity inspired approach.

- Prevents attacks based on:

- Timing/metadata.
- Popularity of queries.
- Statistical disclosure.

- However does not consider the *topic* of the queries.



⇒ No dummy indistinguishability

PDS Plausibly Deniable Search [2]

Lion **CATS**

Leopard **CATS**

Tiger **CATS**

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Shower **(dummy)**
BATHROOM

Stock **(dummy)**
BUSINESS

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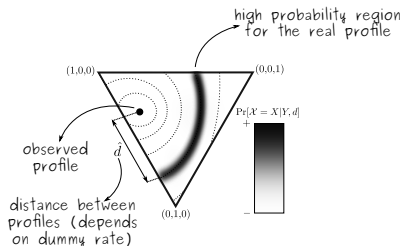


PRAW (A PRivAcy model for the Web) [3]

- Privacy = Dissimilarity.
- Dissimilarity \propto amount of dummy queries.

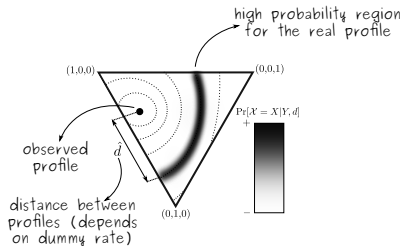
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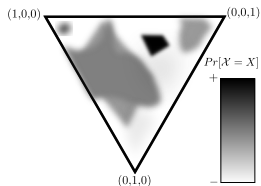


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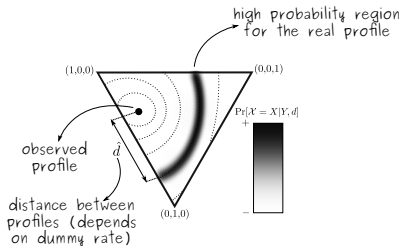


Considering prior information $\Pr[\mathcal{X} = X]$:

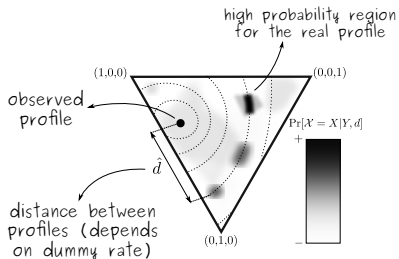


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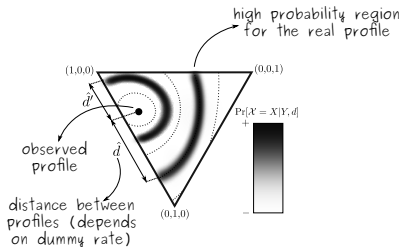


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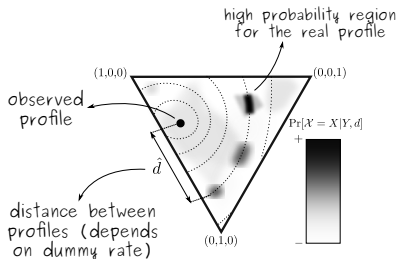


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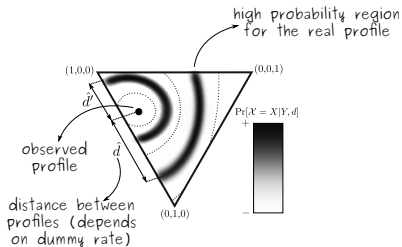


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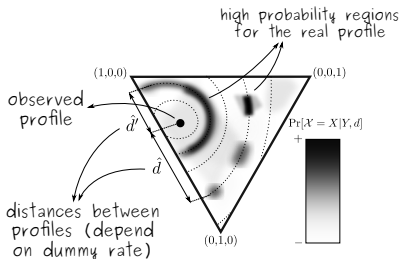


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QQF-PIR Optimized Query Forgery for Private Information Retrieval [4]

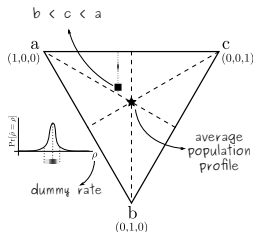
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- Exploitable features:
 - Known target profile.
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 - Waterfilling-based DGS.

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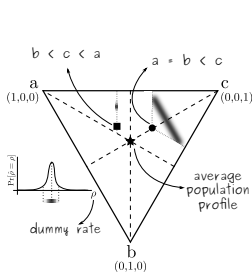
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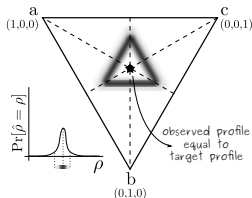
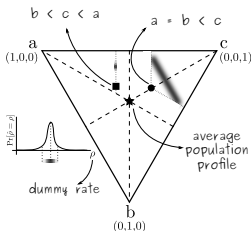
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Systems' Analysis Summary

- Two main categories of DGS:
 - Query based.
 - Profile based.
- Different definitions of what privacy means:
 - k -deniability.
 - The (dis)similarity of profiles.
- Ad-hoc analyses and evaluations.

Open problems and future work

- Plausibility of dummy queries, e.g., The dictionary issue.
- Adversarial modelling, e.g., Adversarial SCA issue.

Conclusions

- Abstract model for OB-PWS systems.
- Analysis framework
⇒ Definition and formalization of relevant privacy properties.
- Analysis of 6 existing OB-PWS systems (4 in this talk).

- Both profile and query based analyses are needed!

Thank you.

Questions?

Main references:

- [1] Josep Domingo-Ferrer, Agusti Solanas, and Jordi Castellà-Roca.
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Online Information Review, 33(4):720–744, 2009.
- [2] Mummoorthy Murugesan and Christopher W. Clifton.
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In *Proceedings of the Workshop on Secure Knowledge Management (SKM 2008)*, November 2008.
- [3] Bracha Shapira, Yuval Elovici, Adlay Meshiach, and Tsvi Kuflik.
PRAW - A PRivAcy model for the Web.
JASIST, 56(2):159–172, 2005.
- [4] David Rebollo-Monedero and Jordi Forné.
Optimized query forgery for private information retrieval.
IEEE Transactions on Information Theory, 56(9):4631–4642, 2010.