VirtualIdentity: a Privacy-Preserving User Profiling Service

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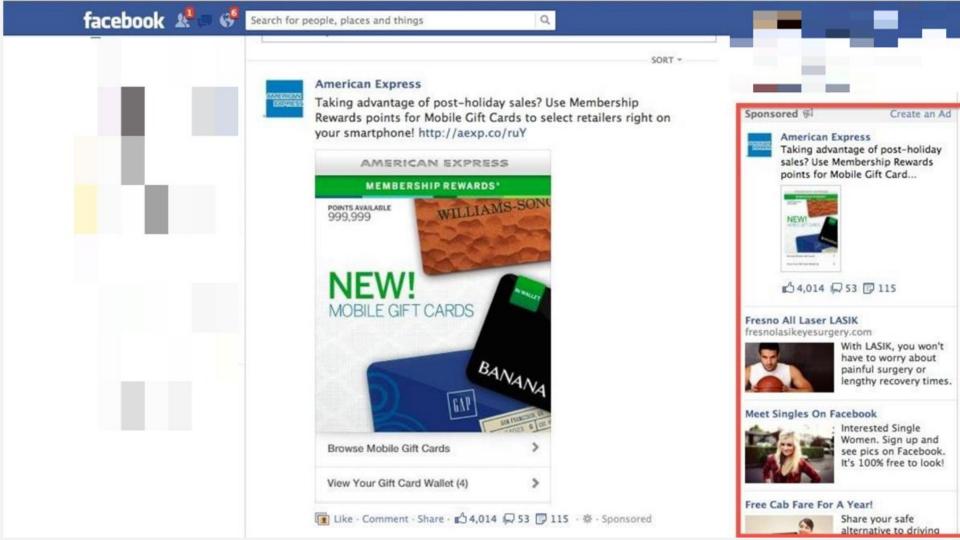
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User Profiling



Machine Learning

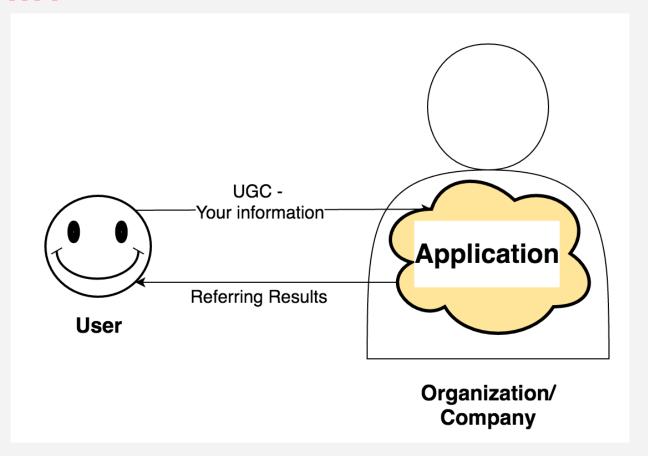
Gender, Age,
Personality,
Religion, Sexual
Orientation,
Interest,...



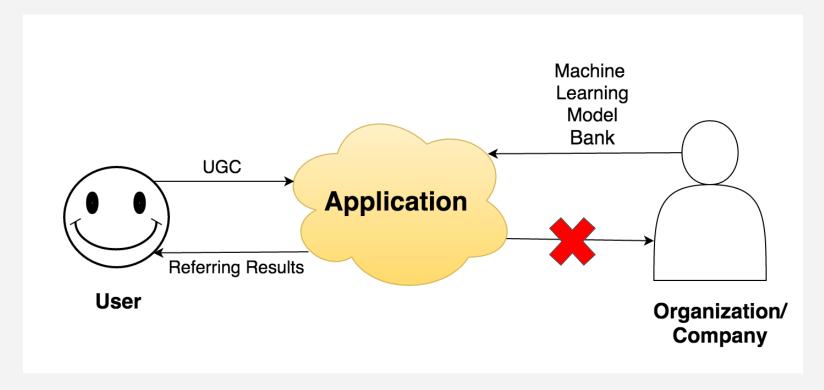
Image Credit: The New Yorker cartoon modified by Eric Blattberg / VentureBeat



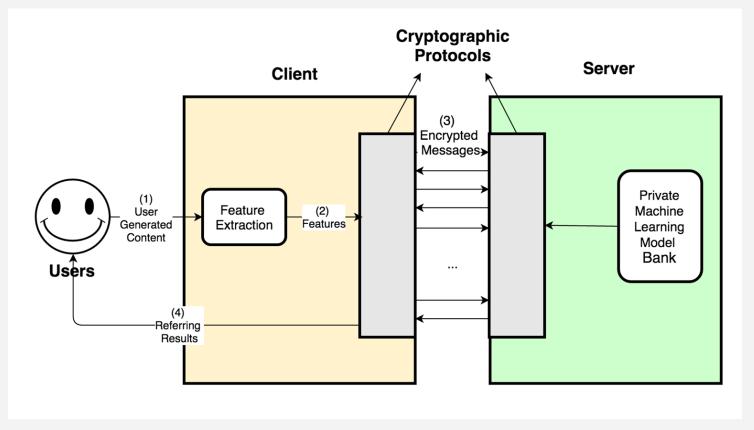
Problem?



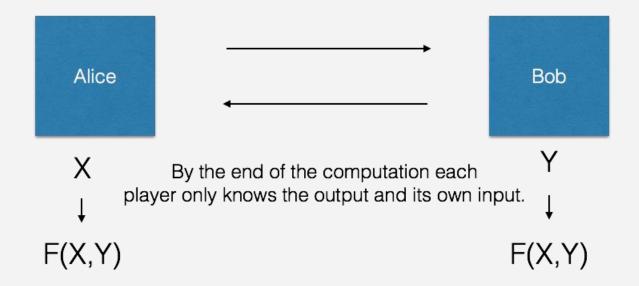
Privacy-Preserving



VirtualIdentity -- Overview



Secure Two Party Computations





A. Yao. How to Generate and Exchange Secrets. In 27th FOCS, pages 162-167, 1986.

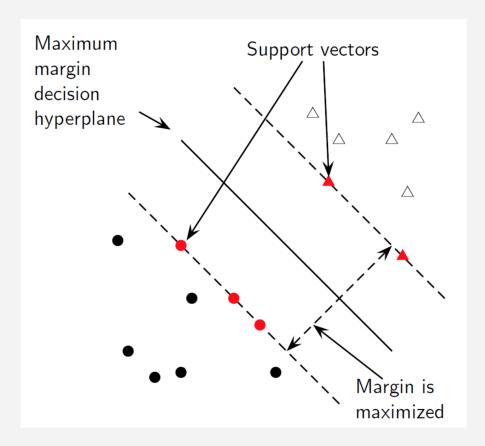
Cryptographic Protocols

Decompose **Machine Learning Scoring** operation into smaller and simpler operations

Use **Secure Multi-party Computation** to make each simple operation privacy-preserving

Combine the Secure Multi-party computations to get the Machine Learning Scoring privacy-preserving

Support Vector Machine Scoring



Scoring for new instance x_q:

$$f(x_q) = \operatorname{sign}\left(\sum_{i} \alpha_i y_i K(x_q, x_i)\right)$$

- Multiplication
- Addition
- Comparison
- ...

Privacy-Preserving Comparison -- by Multi-party Computation

Let ℓ be the bit length of the integers to be compared. The trusted initializer pre-distributes the correlated randomness necessary for the execution of all instances of the distributed multiplication protocol. The parties have as inputs shares $[x_i]_2$ of each bit of x and shares $[y_i]_2$ of each bit of y. The protocol proceeds as follows:

- 1. For $i = 1, ..., \ell$, compute $\llbracket d_i \rrbracket_2 \leftarrow \llbracket y_i \rrbracket_2 (1 \llbracket x_i \rrbracket_2)$ using the multiplication protocol π_{DM} and locally compute $\llbracket e_i \rrbracket_2 \leftarrow \llbracket x_i \rrbracket_2 + \llbracket y_i \rrbracket_2 + 1$.
- 2. For $i = 1, ..., \ell$, compute $[\![c_i]\!]_2 \leftarrow [\![d_i]\!]_2 \prod_{i=i+1}^{\ell} [\![e_j]\!]_2$ using the multiplication protocol π_{DM} .
- 3. Compute $[w]_2 \leftarrow 1 + \sum_{i=1}^{\ell} [c_i]_2$ locally.

We hope to achieve highly practical results that allow the benefits of machine learning to be unlocked without the cost of individual privacy.

Check our website: http://secureml.insttech.washington.edu/

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