

# Filling Context-Ad Vocabulary Gaps with Click Logs

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Yahoo Japan Corporation

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- Our paper pdf in the proceedings seems to be broken...
- Please refer to the followings and get the pdf
  - <http://dl.acm.org/citation.cfm?id=2623334>
  - [http://research-lab.yahoo.co.jp/ml/20140701\\_81.html](http://research-lab.yahoo.co.jp/ml/20140701_81.html)

- For contextual advertising, we propose a method of translating contextual information into the textual features of ads by using past click data
  - In this study, we focus on increasing the click-through rate(CTR)
- We applied our approach to a real ad serving system and achieved an improvement over the existing production system

- Contextual advertising for Japanese market
  - Developed by **Yahoo! JAPAN**
  - Annual revenue: 30 billion JPY (300 million USD)

Example on a Web page of trip for Hawaii (ハワイ)



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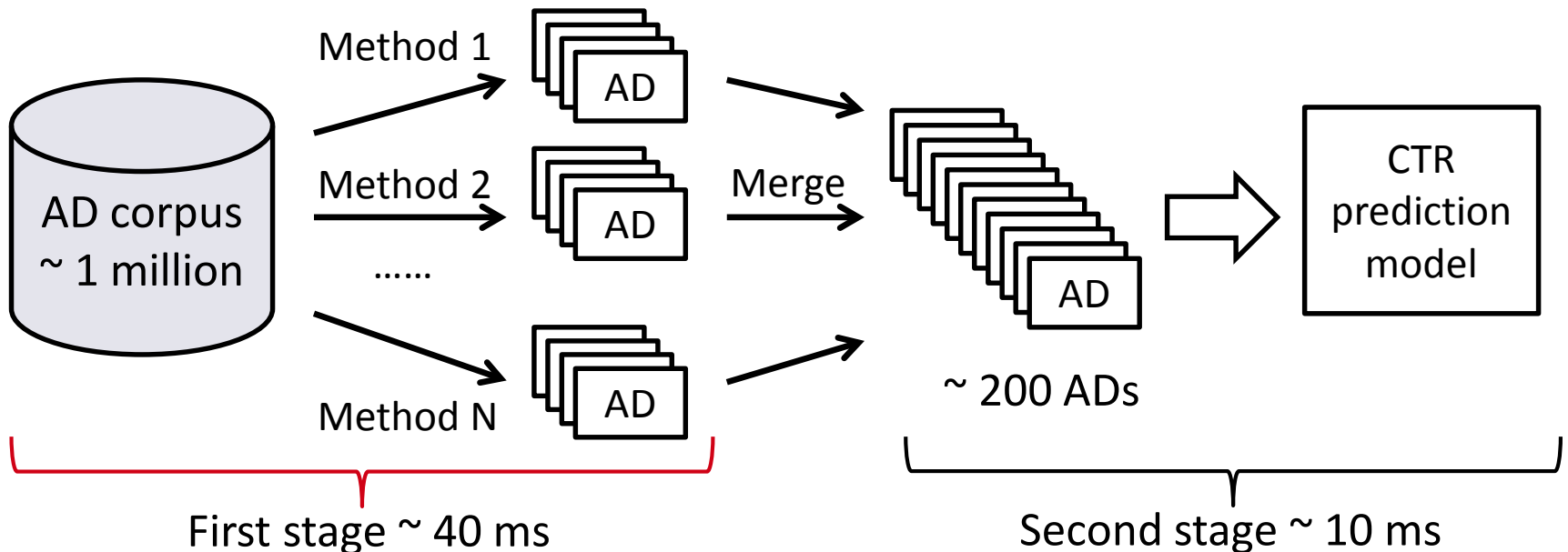
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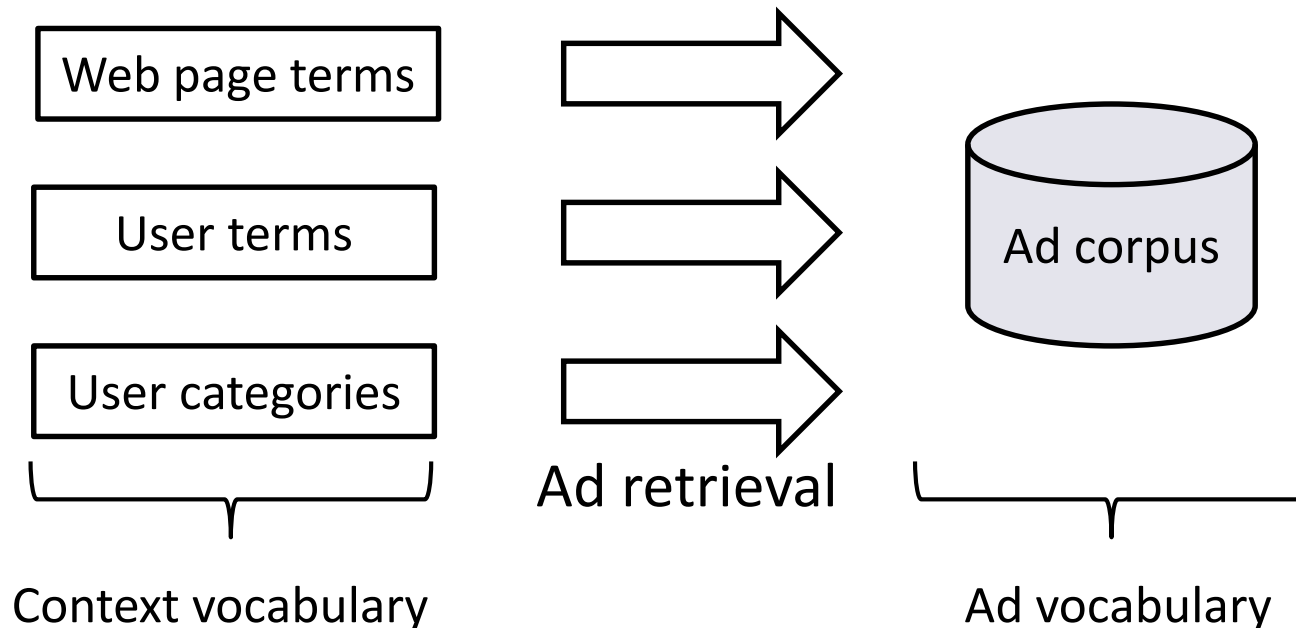
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- Two-stage approach in our ad serving system
  1. Ads are retrieved by multiple methods using inverted index
  2. The ads are merged and passed to click-through rate(CTR) prediction model



- Ad serving system selects ads that are relevant to the page content and/or user information
  - Advertisers are primarily interested in targeting relevant users
  - Publishers, which is a Web page owner, are concerned with keeping the user experience pleasant

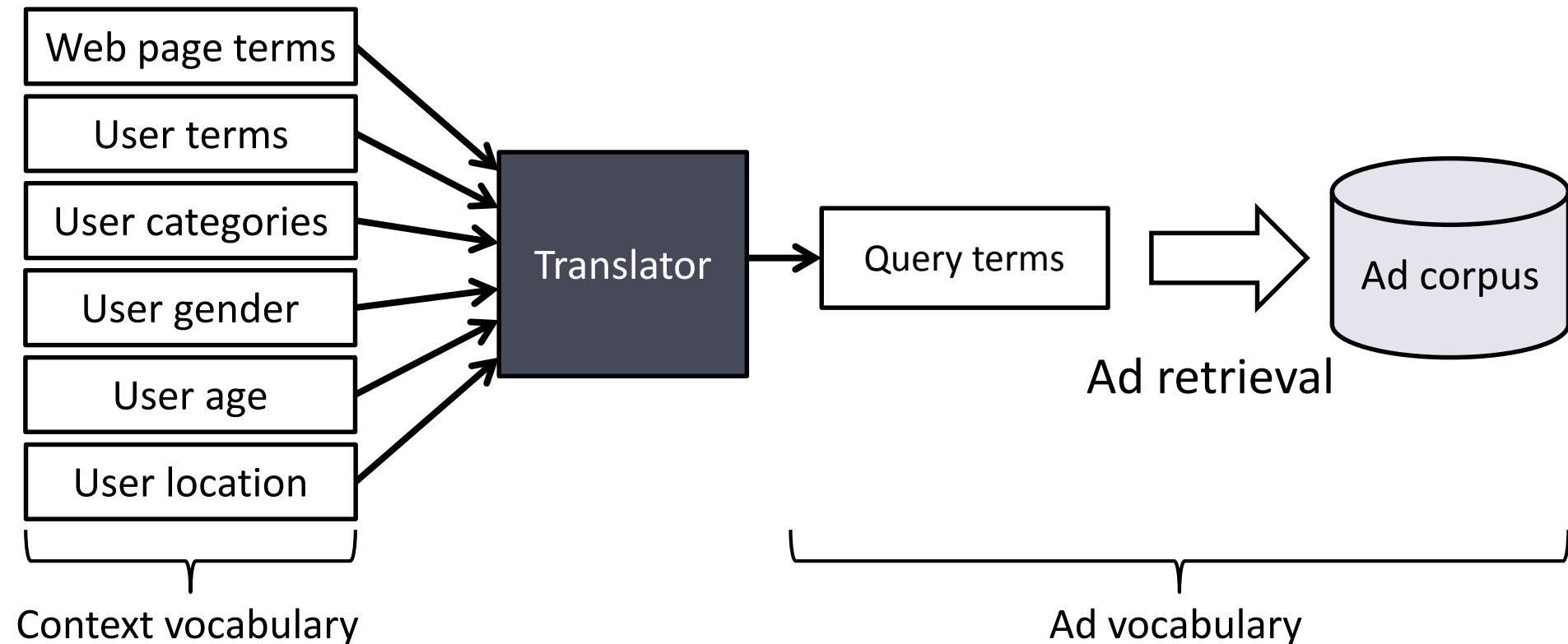
- To measure the word overlap between two types of information
  - In practice, these vocabularies are different  
→ **context-ad vocabulary gaps**



- To use semantic taxonomy or hidden classes
  - It is necessary to expand the ad retrieval system or build new index to handle it
  - In the operation, the following tasks are periodically required
    - A review of the number and hierarchical structure of categories
    - A re-creation of clusters
- In practice, these are not always easy to perform



- Our approach translates contextual information into the textual space of ads
- The translation table is learned with click logs





$$\begin{aligned} \text{score}(\mathbf{q}, \mathbf{a}) &= \mathbf{q}^T \mathbf{W} \mathbf{a} + \mathbf{w}_{basic}^T \mathbf{x}_{basic} \\ &= \sum_{i=1}^{D_q} \sum_{j=1}^{D_a} w_{ij} q_i a_j + \mathbf{w}_{basic}^T \mathbf{x}_{basic} \end{aligned}$$

$\mathbf{q} \in \mathbb{R}^{D_q}$  : contextual feature vector

$\mathbf{a} \in \mathbb{R}^{D_a}$  : ad feature vector

$\mathbf{x}_{basic}$  : features such as ad's display position

$\mathbf{w}_{basic}$  : weight vector corresponding to  $\mathbf{x}_{basic}$

$\mathbf{W}$  : translation matrix

} features

} parameters

- Feature selection using past click statistics
  - To learn the translation matrix efficiently
  - To make the transformed vector sparse
- The score is represented linear form

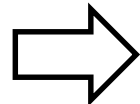
$$\begin{aligned} score(\mathbf{q}, \mathbf{a}) &= \sum_{(i,j) \in P} w_{ij} q_i a_j + \mathbf{w}_{basic}^T \mathbf{x}_{basic} \\ &= \mathbf{w}_{match}^T \mathbf{x}_{match} + \mathbf{w}_{basic}^T \mathbf{x}_{basic} \\ &= \mathbf{w}^T \mathbf{x} \end{aligned}$$

$$P = \{(i, j) \mid m_{ij} > T\}, \quad m_{ij} = \frac{ctr(q_i, a_j)}{\max(ctr(q_i), ctr(a_j))}$$

- In our setting, some ads are displayed on a page at the same time
- User clicks the ad because it is preferable in the listing

Clicked ad requests

AD1	✓
AD2	✗
AD3	✗
AD4	✓
AD5	✗



Pairwise preferences

AD1	>	AD2	AD4	>	AD2
AD1	>	AD3	AD4	>	AD3
AD1	>	AD5	AD4	>	AD5

- Defining a pairwise loss function
  - RankSVM (squared hinge loss)

$$L(\mathbf{w}) = \sum_{r \in R^+} \sum_{i: y_i^{(r)} = 1} \sum_{j: y_j^{(r)} = 0} \max(0, 1 - \mathbf{w}^T (\mathbf{x}_i^{(r)} - \mathbf{x}_j^{(r)}))^2$$

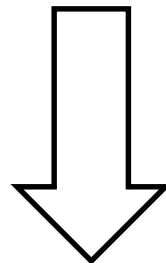
$R^+$ : a set of ad requests which includes clicked ad

- We add a regularization term to avoid overfitting and seek the parameters

$$\hat{\mathbf{w}} = \arg \min_{\mathbf{w}} \frac{1}{2} \|\mathbf{w}\|_2^2 + C \cdot L(\mathbf{w})$$

- With the learned matrix, the contextual feature vector is transformed into the input term vector of the ad retrieval system for each ad request

$$\mathbf{q}_{input} = \mathbf{q}^T \hat{\mathbf{W}}$$



Top- $k$  similarity search  
using inverted index  
(WAND algorithm)

$$(\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_N)$$

- Relative gain over existing method
  - Click-through rate(CTR)
    - Colored value represents “p-value < 0.05” on chi-squared test (website A, B, D, E, H)
  - Cost per click(CPC)
  - Revenue per request(RPR)

Metrics	Website							
	A	B	C	D	E	F	G	H
CTR	-3.67%	+4.60%	+0.48%	+2.82%	+2.47%	+1.42%	+3.27%	+4.02%
CPC	+3.63%	-2.00%	+1.62%	+1.31%	-1.01%	+7.51%	-2.42%	-2.94%
RPR	-0.18%	+2.51%	+2.10%	+4.17%	+1.44%	+9.04%	+0.77%	+0.97%

- Example of Web site B's table for user terms
  - "User terms" are extracted from user behavioral events

User term	Translated term	Weight
iPhone	iPhone	0.2114
	ケース(case)	0.1534
	iPad	0.0868
プリウス(Toyota Prius)	プリウス(Toyota Prius)	0.2600
	燃費(mileage)	0.0732
	HV(Hybrid Vehicle)	0.0607
温泉(hot spring)	温泉(hot spring)	0.1730
	旅館(Japanese inn)	0.1272
	露天風呂(outdoor hot spring)	0.0809



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# Questions?

Please speak clearly and slowly.

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