ТЕМАТИЧЕСКАЯ КЛАССИФИКАЦИЯ



Тематическая классификация

Зачем?

- релевантность (в т.ч. рекламы)
- вертикальные поиски

Данные

- список запросов
- коллекция текстов
- веб
- клики



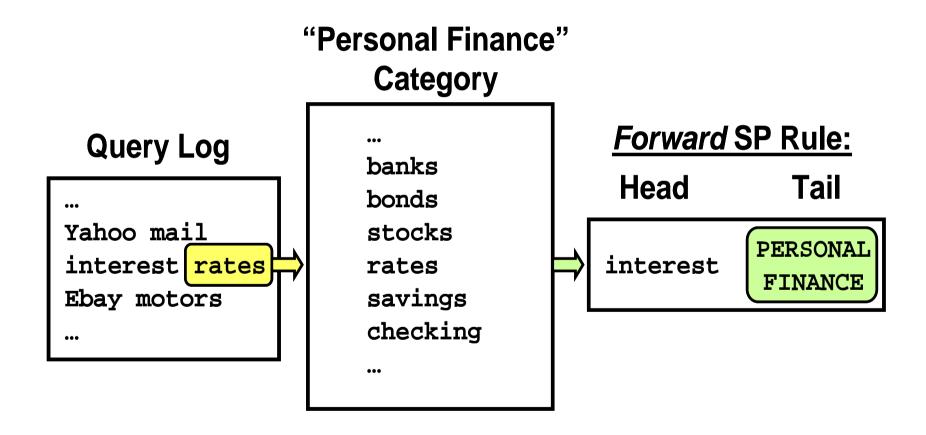
- Obtain a large log of unlabeled web queries
- View each query as pairs of lexical units:
 - <head, tail>
 - Only applicable to queries of 2+ terms
 - Queries with n terms form n-1 pairs
 - Example: "directions to DIMACS" forms two pairs:
 - <directions, to DIMACS> and <directions to,DIMACS>

- Obtain a set of manually labeled queries
- Check the heads and tails of each pair to see if they appear in the manually labeled set

Павел Браславский - Анализ запросов

- Convert each <head, tail> pair into:
 - <head, CATEGORY> (forward preference)
 - <CATEGORY, tail> (backward preference)





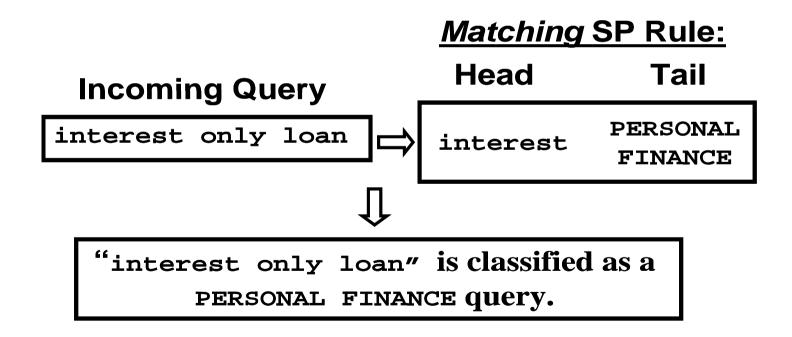
 Score each preference using Resnik's Selectional Preference Strength formula:

$$S(x) = D(P(U|x) || P(U))$$

$$= \sum_{u} P(u|x) \log_{2} \left(\frac{P(u|x)}{P(u)}\right)$$

- ☐ Where u represents a category, as found in Step 2.
- ☐ S(x) is the sum of the weighted scores for every category associated with a given lexical unit

 Use the mined preferences and weighted scores from Steps 3 and 4 to assign classifications to unseen queries



Selectional Preference Rule Examples

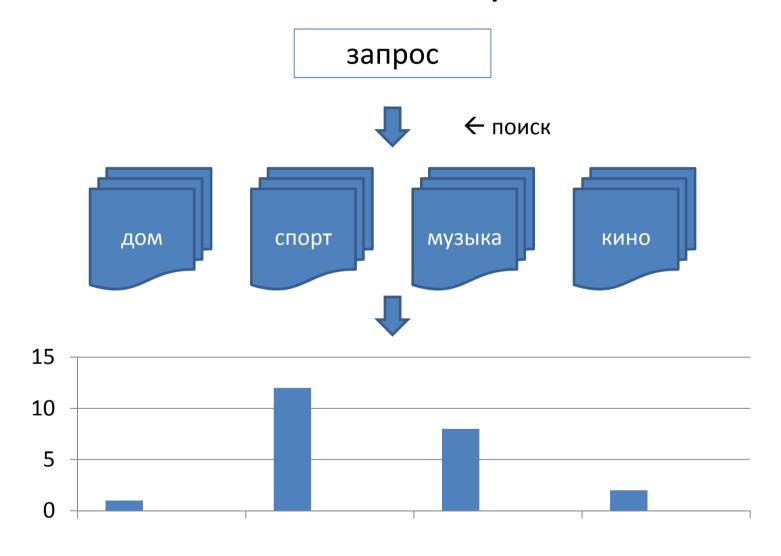
Forward Rules

- harlem club X
 - ENT->0.722
 - PLACES->0.378
 - TRAVEL->1.531
- harley all stainless X
 - AUTOS->3.448
 - SHOPPING->0.021
- harley chicks with X
 - PORN->5.681

Backward Rules

- X gets hot wont start
 - AUTOS->2.049
 - PLACES->0.594
- X getaway bargain
 - PLACES->0.877
 - SHOPPING->0.047
 - TRAVEL->0.862
- X getaway bargain hotel and airfare
 - PLACES->0.594
 - TRAVEL->2.057

На основе коллекции текстов

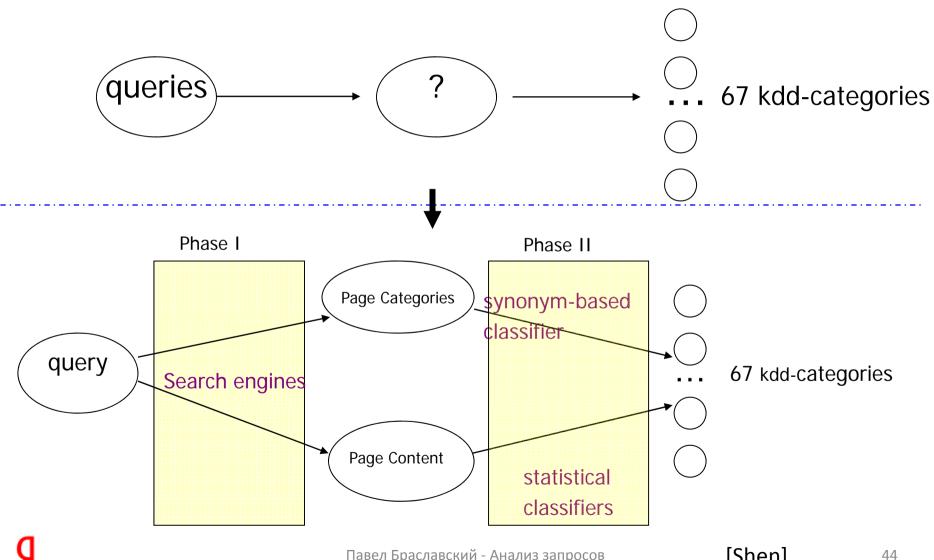


KDD Cup 2005

- Классификация запросов
- 800,000 запросов, 67 категорий
- есть примеры, нет обучающего множества
- нет подробного описания категорий
- ответ системы: до 5 категорий
- оценка: 800 запросов оцениваются тремя асессорами
- метрика: F1



Nature of Problem



Phase I: From queries to pages and categories

- Input:
 - A query: Q_i,
- Output:
 - <Page list_{i.} Category list_i >
- Approach:
 - through Search Engines (SE)

- We collected
 - 40 million entries
 - 50GB
- Search engines
 - Lumur (CMU open source)
 - Google
 - ODP
 - Looksmart

Computerworld

Category: <u>Computers > News and Media > Magazines and E-zines > Magazines</u>

News and product coverage for information technology managers.

www.computerworld.com/ - Cached - Similar pages

Computer Associates

Category: Computers > Security > Authentication > Single Sign-On
Makers of eTrust Single Sign-On. Automates access to authorized Web services and
enterprise applications.

www.ca.com/ - <u>Cached</u> - <u>Similar pages</u>

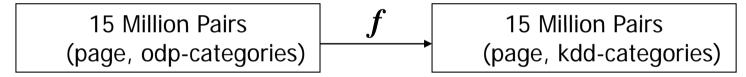
Phase II.a Synonym-based Classifier: using directories

- 67 KDD-categories in KDDCUP
- 172,565 in ODP/Google,272,405 in Looksmart
- For each of the KDDCUP category
 - Apply Wordnet to find the corresponding synonyms in the categories of ODP (Google) and Looksmart, respectively
- This produces one mapping function f for each directory
 - Also returns a rank by matching frequency

- Advantage
 - Fast,
 - Precise
- Disadvantage
 - Many of the 172K and 272K categories from ODP/Google and Looksmart do not map to KDDCUP categories
 - This may result in low recall

Phase II.b: Statistical Classifiers

- Statistical Classifiers
 - Support Vector Machine (SVM): mapping pages to KDDCUP categories
- Training Data
 - 15 million pages with categories from ODP Directory
 - Apply the mapping f from Phase II.a, to build training data.



- Application of the classifier
 - Construct a virtual document for each query by combining the snippets from the returned pages given in Phase I.
 - Classifier returns category and rank

Component Classifier Integration

- We follow an ensemble learning approach
 - Each classifier returns the category and rank
 - The two kinds of classifiers have the similar performance.
 - We integrate the different classifiers together by a weighted sum of the ranks
 - Weights can be determined by validation data set:
 - Based on the performance on the 111 sample data;
 - Assign different weight values for a classifier on different categories
 - The higher the precision, the higher the weight value
 - We have also tried to use equally weighted component classifiers



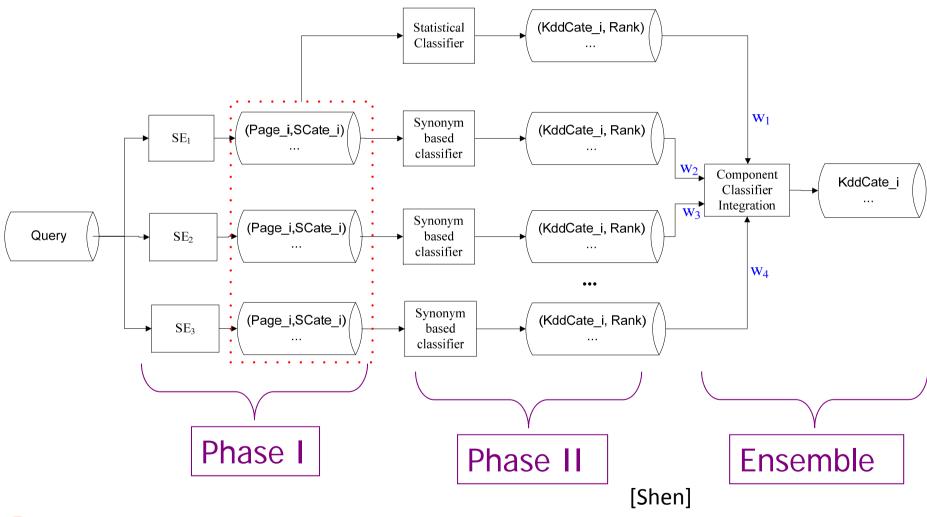
Final Result Generation

- Two Solutions: One for each evaluation criteria
 - S1: Using the validation data set is expected to achieve better precision measure
 - Since each component classifier is highly weighted on the classes where it achieves high precision.
 - 52: Equally weighted combination is expected to achieve higher F1 performance
 - Since the recall is relatively high
 - Evaluation Results (http://www.acm.org/sigs/sigkdd/kdd2005/kddcup.html)

	Submission ID	Precision	F1
<i>S1</i>	37	0.423741	0.426123
<i>S2</i>	22	0.414067	0.444395

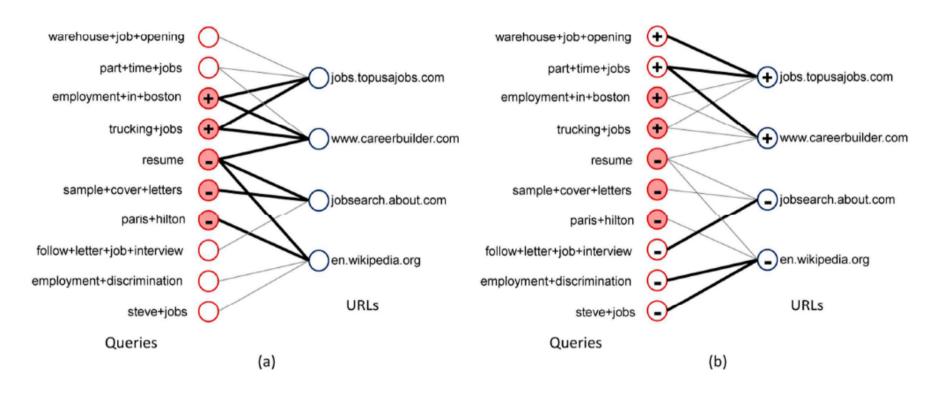
The Results are generated automatically.

Putting them together





Классификация на основе кликов



- 1. Расширение обучающего множества
- 2. Классификация на основе лексических признаков
- 3. Комбинация методов

Li et al. SIGIR2008

Классификация запросов

$$p_{\lambda}(y|x) = \frac{\exp^{\sum_{j} \lambda_{j} \phi_{j}(x,y)}}{\sum_{y} \exp^{\sum_{j} \lambda_{j} \phi_{j}(x,y)}}$$

x — запрос, y — класс (0/1) Признаки $\varphi(x, y)$ - n-граммы [britney spears] \rightarrow britney, spears, <s> britney, britney spears, spears </s>, <s> britney spears, britney spears </s>

Распространение меток

- W матрица mхn, w_{ij} количество кликов на документ j по запросу i
- F матрица mх 2 , f_{iy} вероятность принадлежности запроса i классу y
- F^{θ} первоначальная разметка $B=D^{-1/2}W$
- Итерации:

$$H^{i} = B^{T} F^{i-1};$$

$$F^{i} = \alpha B H^{i} + (1 - \alpha) F^{0}$$

Комбинация

```
Input: matrix F<sup>0</sup> and matrix B = D<sup>-1/2</sup>W
Output: F* and λ*
1: Initialize F* = F<sup>0</sup>, and initialize λ as random;
2: repeat
3: Find λ* = argmin Q(F*, λ) using stochastic gradient descent;
4: Find F* = argmin Q(F, λ*) using Algorithm 1, where the input are F<sup>c</sup>(λ*) and B;
5: until the value Q(F*, λ*) converges
```

БЛИЗКИЕ ЗАПРОСЫ



михаил булгаков мастер и маргарита мастер и маргарита фильм владимир бортко тарас бульба фильм богдан ступка михаил боярский д'артаньян три мушкетера александр дюма



Подробнее

- 1. Уточнение: золотое кольцо → золотое кольцо с бриллиантом
- 2. Снятие неоднозначности: ягуар → ягуар животное
- 3. Расширение: золотое кольцо → ювелирные украшения
- 4. Синоним:
 японская вишня → сакура

Еще подробнее

- 5. Другой запрос на ту же тему: $мерседес \rightarrow aydu$
- 6. Смена поисковой цели: купить санки → детский мир
- 7. Перевод: коралловый клуб → coral club



Близкие, но «про другое»

- Опечатки: курсовая робота курсовая работа
- Транслитерация: *золото zoloto*
- Раскладка: *lbvf* ,*bkfy* дима билан
- Реникса: otbeptka oтвертка



Хороший запрос – это непросто

Назовите глагол из вопроса, помещенного на борту транспортного средства подопечных Фатиха Терима на первенстве континента?

IX Кубок Яндекса по поиску (2008) http://kubok.yandex.ru



Работа мысли

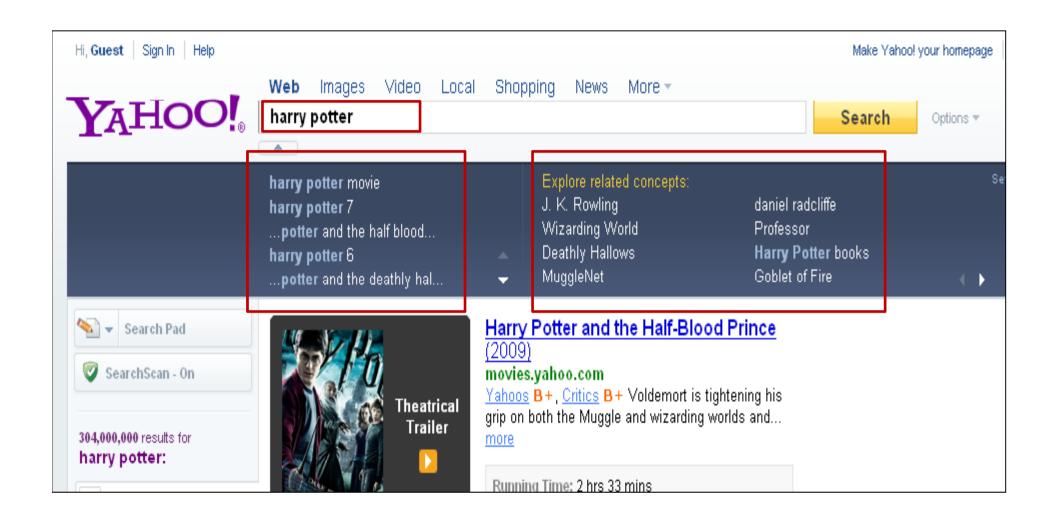
- Фатих Терим
- сборная турции на борту самолета
- сборная турции "на борту" самолета
- сборная турции надпись "на борту" самолета
- сборная турции надпись "на борту" автобуса
- сборная турции надпись на автобусе
- футбол "сборная турции" надпись на автобусе
- чемпионат европы футбол "сборная турции" надпись на автобусе



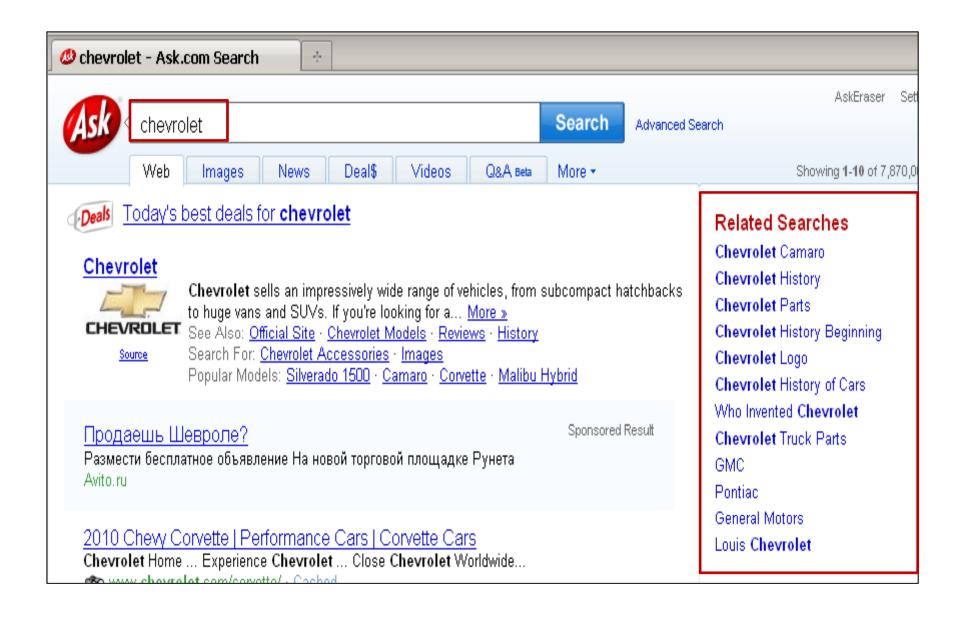
Ответ

Вместит ли автобус всю страсть Турции?











Video results for hermione granger



Hermione Granger 3 min 23 sec - 4 May 2006 www.youtube.com



Emma Watson/Hermione
Granger *Lucky*
3 min 26 sec - 28 Jan 2008
www.youtube.com

Searches related to hermione granger

hermione granger pictures

ginny weasley

<u>harry potter quiz</u>

hermione granger costume

ron weasley draco malfoy

<u>harry potter</u>

emma watson

G000000000gle >
1 2 3 4 5 6 7 8 9 10 Next

hermione granger

Search within results - Language Tools - Search Help - Dissatisfied? Help us in

Яндекс картинки

Поиск Почта Карты Маркет Новости Словари Блоги Видео Картинки ещё▼

гарри поттер

в найденном

Любые

Обои Большие Средние Маленькие

Портреты







Поищите также:

гермиона рон уизли хогвартс

harry potter дамблдор снейп

джинни уизли драко малфой ден рэдклифф

Яндекс.Видео



Гарри Поттер 3 Еще по запросу: «гарри поттер» 71145



Гарри Поттер и Орден феникса (Harry Potter and the Order of the...

240×215 ... 699×631

www.kino-govno.com

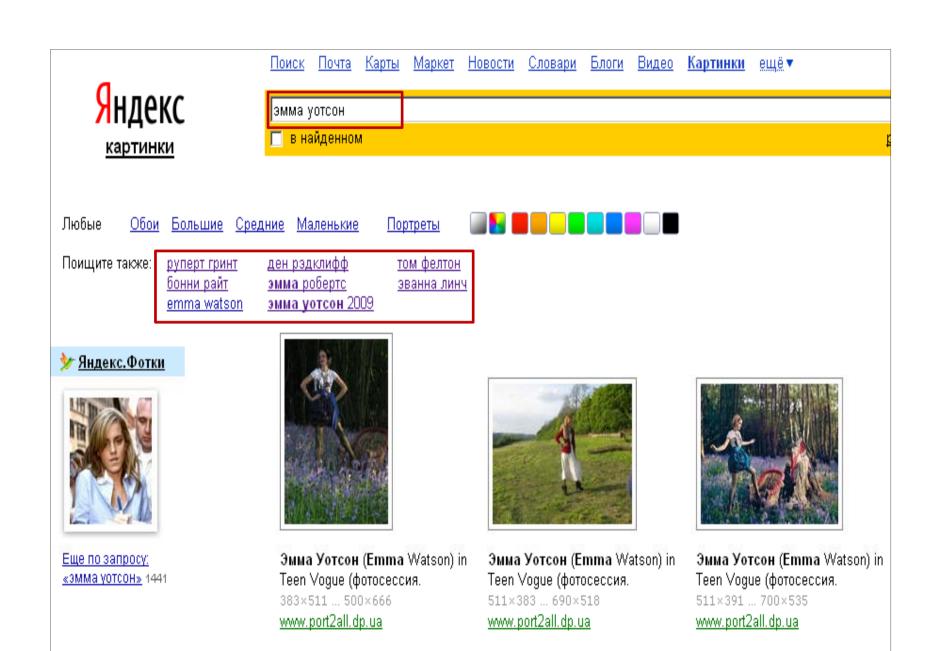


73. Harry Potter 5 / Гарри Поттер 5Просмотров 62×100 ... 850×1360 gallery.galaxy-portal.ru



Гарри Поттер и орден Феникса / Harry Potter and the Order of the...

120×90 ... 1500×1125 relax.grad66.ru



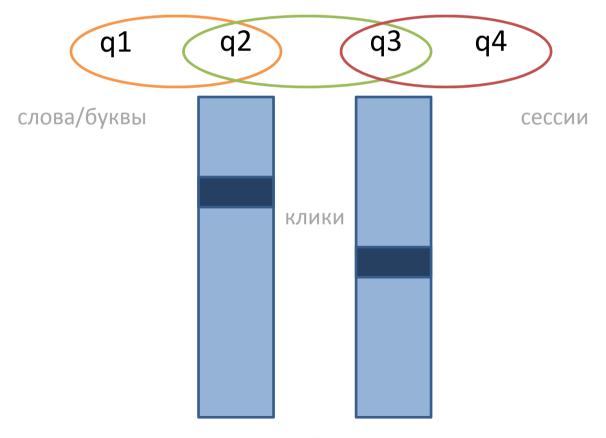
Источники данных

- Лог запросов
- Текст ссылок
- Корпус текстов

```
ford → ford focus, ford fusion, ford mondeo карта → карта памяти, карта города машина → стиральная машина, швейная машина
```



Близость запросов





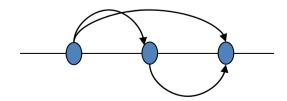
Метод

запрос=<timeStamp, userId, queryText, ckicks>

- О. ЧИСТКА ЛОГА (порно, Яндекс, «подсказки», язык запросов, опечатки)
- 1. выделение сессий (по времени)



2. выделение пар (в.ч. транзитивно)



Метод – 2

- 3. нормализация запросов (стоп-слова, капитализация, лемматизация, сортировка слов + выбор лучшего обратного преобразования)
- 4. борьба со флешмобами и «событийными» ассоциациями
- 5. матрица частоты переходов «запрос-запрос» (пороги для пользователя, ограничение на абсолютную частоту)
- 6. weight(q1 \rightarrow q2)= f(freq_{q1}, freq_{q2q1})*f(freq_{q2q1})
- 7. ранжирование, отсечение по порогу
- 8. индекс: $q \rightarrow q1, q2, q3...$ (оригинальные запросы)

Оценка

- «оценка глазами»
- «классы объектов»
- кластеры запросов Яндекс.Директа
- кликабельность



Результат

« манго » — связанные запросы			« умные фразы» — связанные запросы			
<u>mango</u>	манго одежда аво	<u>кадо</u>	<u>умные мысл</u>	<u>ій</u> <u>умны</u>	е слова	<u>смешные</u> фразы
<u>папайя</u>	<u>манго фрукт</u> инх	кир	<u>умные цитат</u>	<u>гы умны</u>	е статусы	фразы о жизни
<u>финик</u>	<u>магазин манго</u> пол	<u>нело</u>	<u>умные вопр</u>	<mark>осы</mark> фразь	ы о любви	мудрые фразы
«хочу похудеть» — связанные запросы			« смайлики» — связанные запросы			
хочу люб	<u>ви</u> <u>хочу умереть</u>	<u>смайлы для qip анимашки smiles</u>				
<u>хочу ребенка</u> <u>хочу денег</u>			<u>улыбка анимации иконки</u>			
			<u>позитив</u>			
« люблю » — связанные запросы			« утконос » — связанные запросы			
<u>любимом</u> у	<u>у смс скучаю се</u>	<u>одце</u>	<u>ехидна</u>	<u>коала</u>	кенгуру	
<u>поцелуй</u>		<u>выхухоль</u>	<u>эму</u>	муравьед		
			<u>вомбат</u>	<u>динго</u>	броненосец	
« лук» — связанные запросы			« ива» — связанные запросы			
рыцарь в	<u>ампир</u> <u>лукашенко</u>	мечи	<u>дуб</u> <u>то</u>	<u>поль</u> <u>бер</u> с	<u>e3a</u>	
<u>капуста</u>	<u>морковь</u>	<u>свекла</u>		<u>сна</u> <u>клен</u>		
<u>огурец</u>	<u>картофель</u>	<u>арбалет</u>	<u>ель</u> <u>ка</u>	<u>штан</u> <u>рябі</u>	<u>ина</u>	

Литература

- Broder A. A Taxonomy of Web Search. SIGIR 2002.
- Broder A. et al. Robust classification of rare queries using web knowledge.
 SIGIR 2007.
- Risvik K. M. et al. Query Segmentation for Web Search. WWW2003.
- Bergsma S. & Wang Q. I. Learning Noun Phrase Query Segmentation. EMNLP-CoNLL 2007.
- Hagen M. et al. The Power of Naïve Query Segmentation. SIGIR 2010.
- Beitzel, S.M. et al. Temporal analysis of a very large topically categorized web query log, JASIST, vol. 58, no. 2, 2007.
- Beitzel, S.M., et al. Automatic classification of web queries using very large unlabeled query logs. ACM Trans. Inf. Syst., 25(2):9, 2007.
- Shen D. et al. Q2C@UST: Our Winning Solution to Query Classification in KDDCUP 2005, SIGKDD Explorations 7(2).
- Li X. et al. Learning Query Intent from Regularized Click Graphs. SIGIR 2008.

Ссылки

- История про лог AOL см. <u>http://en.wikipedia.org/wiki/AOL search data scandal</u>
- Поиск по логу AOL2006 http://www.aolstalker.com/
- Query Log Analysis Workshop @ WWW2007, http://querylogs2007.webir.org/
- KDD Cup 2005, http://www.sigkdd.org/kdd2005/kddcup.html
- WSCD09: Workshop on Web Search Click Data 2009, http://research.microsoft.com/users/nickcr/wscd09/
- Microsoft Web N-gram Services, http://research.microsoft.com/en-us/collaboration/focus/cs/web-ngram.aspx
- Jiang D. et al. Web Search/Browse Log Mining: Challenges, Methods, and Applications, http://research.microsoft.com/en-us/people/djiang/web_search_and_browse_log_mining.pdf

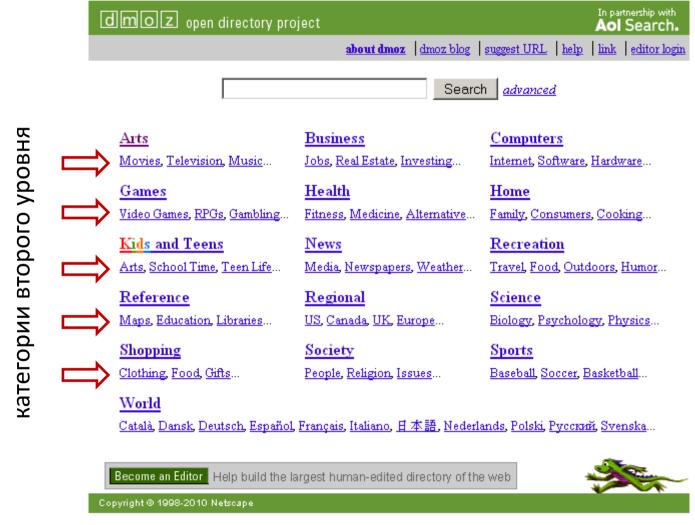
ПРАКТИКА



Практика

- Задачи
 - Сегментация
 - Тематическая классификация
 - Близкие запросы
- Данные AOL2006
 http://www.gregsadetsky.com/aol-data/
 - seed ~2000 запросов,
 http://www.kansas.ru/querylog_analysis/2000queries.txt
 - + описания DMOZ для категорий второго уровня http://narod.ru/disk/371552001/dmoz_data.zip.html
- Можно использовать любые внешние ресурсы (не человеческие ;)

DMOZ



4,771,467 sites - 89,784 editors - over 1,001,177 categories

Примеры

Сегментация

- at the vet | norman rockwell | litho
- big weenie | by eminem

Тематическая классификация

• big weenie by eminem --> Arts_Music /t Art_Television До трех категорий, упорядоченных по уменьшению уверенности

Кластеризация

 1996 mitsubishi mirage --> 2001 subaru impreza wagon /t toyota corolla /t toyota sienna hybrid

До 10 запросов из большого лога (сначала – самый близкий)

- Присылайте по почте с темой querylog_analysis_results
- segmentation_имя_фамилия.txt
- classification_имя_фамилия.txt
- clustering_имя_фамилия.txt
- имя_фамилия.pdf краткое описание методов





Павел Браславский pb@yandex-team.ru

