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Measures for assessing the data freshness in Open Data portals

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Problem/Challenge



- How up-to-date are resources in Open Data portals?
- Required information for such a metric:
 - Change history of documents in portals
- Challenge:
 - 1. Collect available change history
 - 2. Estimated next change time to assess up-to-dateness
- Two scenarios:
 - *Portal provider*: wants to add freshness measure to metadata
 - Data consumer: updating of application, DB, etc..



Open Data Portals



- Single point of access
- Local and external resources
- Meta data
 - Title
 - Modification date
 - ...



Typical software:





Sources of change information in OD portals





- **Push-based** history:
 - Data provider push change information to portal
 - If *local*, by uploading new version
 - If *external*, by updating a specific metadata field
 last_modified: "2013-09-25T00:00:00"
- Pull-based history:
 - Age sampling:
 - Access to latest change time of a resource (i.e., last-modified timestamp in *HTTP Header*)

```
Last-Modified: Mon, 04 Nov 2013 13:00:08 GMT
ETag: "21096456bff7d72268dc99b3bf082565"
```

- Comparison sampling:
 - Detect changes by monitoring and comparing the resources



Open Data Portal Watch



http://data.wu.ac.at/portalwatch/

- Periodically monitoring over 260 Open Data portals
- Metadata quality assessment
 - Uniform handling of metadata (using DCAT mapping)
- Evolution tracking & archiving
 - Meta data
 - Data

≡	Open Data Portal Info Portals	API Quality Measures A	About
	Portal List containing 261 portals		
	Q Filter		
	SORT BY URL SORT BY ISO3 SORT BY SOFTWARE SORT BY #DATASETS SORT BY #RESOURCES		
	http://opendata.paris.fr.opendatasoft.com		
	SEE PORTAL DETAILS		
	http://datos.argentina.gob.ar/		
	SEE PORTAL DETAILS		



Available change information





- CKAN: age- and comparison-sampling required
- Socrata & OpenDataSoft: push-based possible



Local vs external resources on CKAN

- 130 CKAN portals:
 - 27 portals host all resources externally, 9 all locally
 - Majority of all URLs (~88%) belong to 54 portals with <25% local resources
 - HDX portal: 9574 URLs, 8833 distinct, 2114 local (~24%)

external					local	
ratio	0	< 0.25	< 0.5	< 0.75	< 1	1
p % of $ r $	27 5.76%	54 88.48%	9 0.38%	7 0.05%	27 1.12%	9 4.21%



Estimation of next updates



- Evaluating three change estimation heuristics:
 - Poisson process
 - Cho and Garcia-Molina (2003) propose Poisson process model to estimate updates in the context of Web sites
 - Markov chain approach
 - Umbrich et al. (2015) use Markov chains to schedule next crawl times for URLs based on previous observed changes
 - Empirical distribution
 - Build empirical distribution of changes based on intervals



Estimation of next updates (cont'd)

Age sampling

(last-modified timestamp in HTTP Header)

- Poisson distribution
 - $X/T \ (= \frac{number \ of \ changes}{monitoring \ period})$ as estimator for Poisson parameter
 - Compute next change time by considering *p*-quantiles
- Empirical distribution
 - Use intervals between the observed last-modified times
 - *p*-quantiles of empirical distribution





- Markov chain approach
 - Probability of next change based on previous state, e.g.:







 Extend approach by considering the last k states for computing the probabilities:

i \i+1	1	0	TOTAL
00	2	1	3
01	1	1	2
10	1	1	2
11	0	1	1



P(1|00) = 2/3



Evaluation Summary

Controlled environment:

- Evaluation using revision histories of Wikipedia articles
 - 1562 randomly Wiki articles with >3 years history and >30 revisions
 - Wiki change history does not follow Poisson distribution
- Different confidence values:
 - For fixed p, we report the ratio of successfully predicted updates
- Conclusion:
 - Markov chain approach best for comparison-based sampling
 - Empirical distribution best for pushbased and age-based sampling

COMPARISON SAMPLING RESULTS.								
Estimator	Estimator All		Regular		Irregular			
p = 0.7 S = 10d								
$C_{EmpDist}$ $C_{ChoNaive}$ $C_{ChoImpr}$	0.59 0.67 0.66	40d 36d 35d 42d	0.66 0.67 0.62	40d 35d 34d 41d	0.60 0.63 0.61	90d 83d 82d 96d		
p = 0.7 S = 50d								
$egin{array}{c} C_{EmpDist} \ C_{ChoNaive} \ C_{ChoImpr} \ C_{UmbMarkov} \end{array}$	0.54 0.65 0.27 0.58	40d 37d 43d 39d	0.57 0.36 0.31 0.59	40d 40d 36d 40d	0.57 0.63 0.47 0.68	84d 78d 76d 82d		
p = 0.9 S = 10d								
$egin{array}{c} C_{EmpDist} \ C_{ChoNaive} \ C_{ChoImpr} \ C_{UmbMarkov} \end{array}$	0.81 0.71 0.57 0.88	66d 38d 36d 84d	0.87 0.70 0.66 0.94	70d 37d 35d 85d	0.80 0.67 0.60 0.90	145d 85d 83d 184d		



Thank you for your attention

Goal

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- Data Freshness estimation in Open Data
- Challenge
 - Collecting change history (push vs pull)
- Approach
 - Estimators for different scenarios
 - Empirical evaluation

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