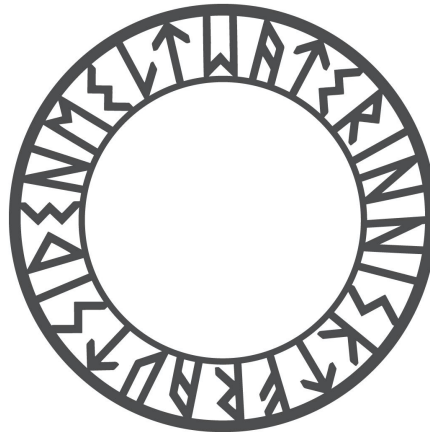# Wrapping Millions of Documents Per Day

and How that's Just the Beginning

Meltwater

Background: About me

Lecturer in Databases and Co-investigator VADA

Fellow of Oxford Martin School & OMI

Co-founder and CTO

Senior Research Manager

Meltwater

Mission

Records

FEATURED
2482 Bound Brook Ln
Yorktown Heights NY 10598 | Map
3 bd 1 ba 1,741 sqft / 0.5 acres
Single-Family Home
Houlihan Lawrence Somers Brokerage

$372,000
$1,742/mo Get Pre-Approved
Terrific split level. Fabulous great room addition with a wall of built-ins, vaulted ceiling, beams, skylights, e ...
View Details

Data Areas

NEW HOMES
Preserve at Ardsley
Scarsdale NY 10583
4 Bd 3,377+ sqft
3.5 Ba New Community
Toll Brothers

From **$1,404,995**
Preserve at Ardsley is a community of 11 single-family ... More

Fields

FEATURED
3468 Carol Ct
Yorktown Heights NY 10598 | Map
3 bd 2 ba 1,602 sqft / 0.38 acres
Single-Family Home
Houlihan Lawrence Somers Brokerage

$379,000
$1,774/mo Get Pre-Approved
LAKELAND SCHOOLS with LOW TAXES. Don't let this one slip away. Nestled away on a quiet cul-de-sac, backing u ...
View Details

Descriptions

FEATURED
2906 Hickory St
Yorktown Heights NY 10598 | Map
3 bd 2 ba 1,290 sqft / 0.26 acres
Single-Family Home
Houlihan Lawrence Yorktown Brokerage

$329,000
$1,540/mo Get Pre-Approved
Incredible updates, since last on the market. Renovated kitchen with luxurious counters and back splash, renovate ...
View Details

Mission

BEACH HAVEN II
600 N. Bay Avenue,
Beach Haven, NJ 08008
609-492-6300
Send Address To E-Mail|Mobile
8.22 miles
Get Driving Directions
Gift Cards Sold, Gift Cards Accepted

BORGATA
1 Borgata Way,
Atlantic City, NJ 08401
609-317-8206
Send Address To E-Mail|Mobile
18.04 miles
Get Driving Directions
Scoop Website

Catering, Gift Cards Sold, Gift Cards Accepted

30TH STREET STATION
30TH STREET STATION, Amtrak - 30th Street Station
Philadelphia, PA 19104
215-222-2996
Send Address To E-Mail|Mobile
48.70 miles
Get Driving Directions
Scoop Website

Address or Postcode

Bridge Rd, Little Egg Harbor Township,

United States of America

SEARCH

☐ Offer catering   ☐ Offer ice cream cakes

© OpenStreetMap contributors

1 BEACH HAVEN II
600 N. Bay Avenue,
Beach Haven, NJ 08008
609-492-6300
Send Address To E-Mail|Mobile
8.22 miles
Get Driving Directions
Gift Cards Sold, Gift Cards Accepted

2 BORGATA
1 Borgata Way,
Atlantic City, NJ 08401
609-317-8206
Send Address To E-Mail|Mobile
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NAME	STREET ADDRESS	LOCALITY	STATE	POSTCODE	PHONE
BEACH HAVEN II	600 N. Bay Avenue	Beach Haven	NJ	08008	609-492-6300
BORGATA	1 Borgata Way	Atlantic city	NJ	08401	609-317-8206
30TH STREET STATION	30TH STREET STATION	Philadelphia	PA	19104	215-222-2996

White House press secretary Sean Spicer speaks in the media briefing room in Washington, D.C., on Saturday. (Olivier Douliery/Abaca Press/TNS)

White House vows to fight media 'tooth and nail' over Trump coverage; says it presented 'alternative facts'

By Doina Chiacu and Jason Lange
Reuters

WASHINGTON — The White House vowed on Sunday to fight the news media "tooth and nail" over what it sees as unfair attacks, with a top adviser saying the Trump administration had presented "alternative facts" to counter low inauguration crowd estimates.

On his first full day as president, Trump said he had a "running war"

---

change

Exclusive: The five-yearly assessment of what will happen to the UK as the world warms says one of an array of potential threats is the 'significant risk' to supplies of food

Ian Johnston, Tom Batchelor Environment Correspondent | 8 hours ago | 💬230 comments

In the report, the Government said high-risk issues that needed to be addressed included the damage expected to be caused by flooding
Getty

The Government has been accused of trying to bury a major report about the potential dangers of global warming to Britain – including the doubling of the deaths during heatwaves, a "significant risk" to supplies of food and the prospect of infrastructure damage from flooding.

The UK Climate Change Risk Assessment Report, which by law has to be produced every five years, was published

DEMO GODS

HELP US PLEASE

memegenerator.net

Demo

https://youtu.be/j_0IZdNJ-aw
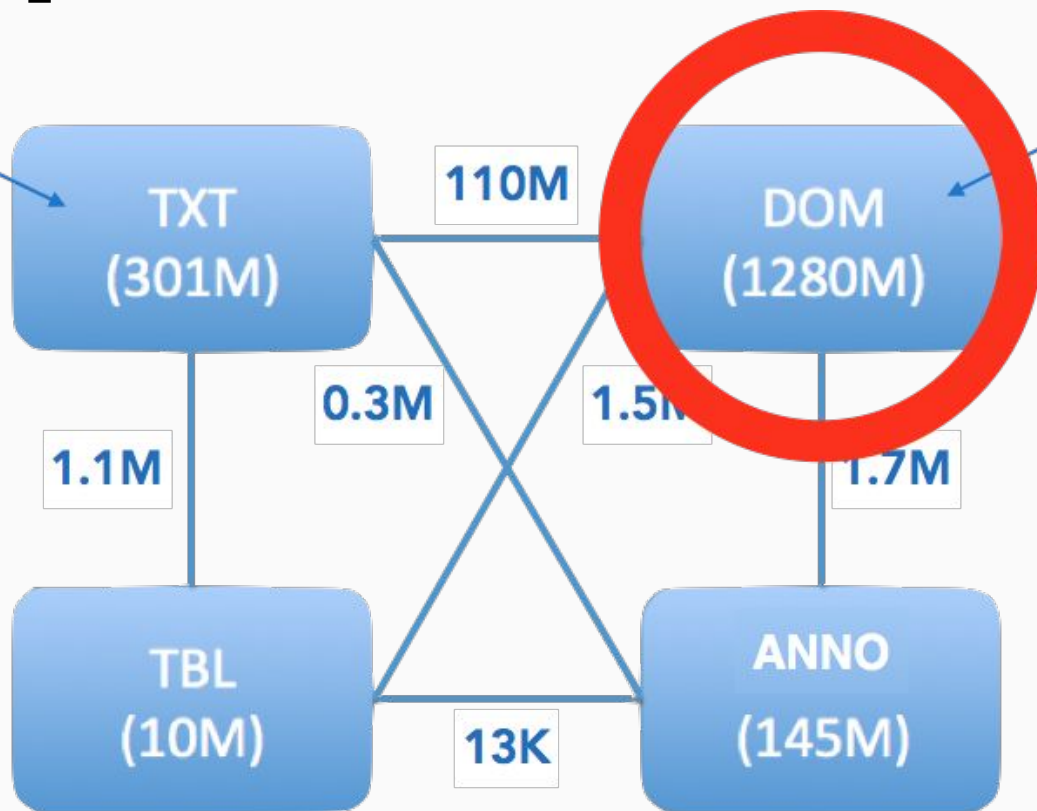
World class <span style="color:red">crawling</span> platform

to largely automate

<span style="color:red">outside content</span> collection

Mission

Crawling Space & Volume

Source: Xin Luna Dong (Google) - PVLDB '14

Crawling Coverage

For many kinds of information one has to extract from **thousands of sites** in order to build a **comprehensive** database

Source: Nilesh Dalvi (Yahoo!) et al. – VLDB 2012

Mission

Vertical 1 & 2:

Real Estate & Used Cars, UK

Results

Results

10,493
Sites from real-estate and used-car

92%
Effective wrappers
(where we get all the data)

96%
Precision of extracted primary attributes

20
Days (one expert) to adjust system to a new domain

VLDB 2015

	wrapper		
	effective	**wrong or missing data**	**no data**
UK real estate	**91%**	7%	2%
Oxford real estate	**90%**	6%	4%
ViNTs[10]	**4%**	5%	91%
UK used cars	**93%**	4%	3%
US real estate	**90%**	5%	5%

```
doc('http://www.wwagency.com/')//label[@for='sale_type_id']/following-sibling::select/{0 /}
        //form/div[@class='formbtn-ctn'][last()]/button[@class='formbtn']/{click /}
   /.:<data_area>[?.//div[@class='pagenumlinks'][1]//span/text():<number_results=.>]
   /(///div[contains(@class,'proplist_wrap')]/following-sibling::div//a[@class='pagenum'][last()]/{nextclick /})*
        //div[contains(@class,'proplist_wrap')]:<record>[? .:<origin_url=current-url()>]
        [? .//span[@class='prop_price']/text():<price=normalize-space(.)> ]
        [? .//span[.='Type:']/following-sibling::strong/text():<property_type=normalize-space(.)> ]
        [? .//div[@class='prop_statuses']//text():<property_status=normalize-space(.)> ]
        [? .//span[.='Bathrooms:']/following-sibling::strong/text():<bathroom_number=normalize-space(.)> ]
        [? .//span[.='Bedrooms:']/following-sibling::strong/text():<bedroom_number=normalize-space(.)> ]
        [? .//strong[@class='orange']/preceding-sibling::text():<location_raw=string(.)> ]
        [? .//strong[@class='orange']/text():<postcode=normalize-space(.)> ]
        [? .//strong/preceding-sibling::strong/text():<street_address=normalize-space(.)> ]
        [? .//@src:<image=normalize-space(.)> ]
        [? .//div[@class='prop_statuses']/following-sibling::a/@href:<url=normalize-space(.)> ]
        [? .//div[@class='prop_maininfo']:<description=normalize-space(.)> ]
```

Vertical 3:

News Articles

Exclusive: The five-yearly assessment of what will happen to the UK as the world warms says one of an array of potential threats is the 'significant risk' to supplies of food

Ian Johnston, Tom Batchelor | Environment Correspondent | 8 hours ago | 230 comments
AUTHOR_NAME AUTHOR_NAME

Click to follow
The Independent Online

White House press secretary Sean Spicer speaks in the media briefing room in Washington, D.C., on Saturday. (Olivier Douliery/Abaca Press/TNS)

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In the report, the Government said high-risk issues that needed to be addressed included the damage expected to be caused by flooding
Getty

White House vows to fight media 'tooth and nail' over Trump coverage; says it presented 'alternative facts'

TITLE

By Doina Chiacu and Jason Lange
AUTHOR_NAME AUTHOR_NAME
Reuters

WASHINGTON — The White House vowed on Sunday to fight the news media "tooth and nail" over what it sees as unfair attacks, with a top adviser saying the Trump administration had presented "alternative facts" to counter low inauguration crowd estimates.

On his first full day as president, Trump said he had a "running war"

The Government has been accused of trying to bury a major report about the potential dangers of global warming to Britain – including the doubling of the deaths during heatwaves, a "significant risk" to supplies of food and the prospect of infrastructure damage from flooding.

BODY_INCLUDE

The UK Climate Change Risk Assessment Report, which by law has to be produced every five years, was published

~40,000

US/UK news sources from Meltwater media intelligence

85%

Effective wrappers (where we get all the data)

89%

Precision of extracted ADICT+ attributes

90

Days to adjust system to this vertical (why? 30min refresh)

Results

```
"site":{ ⊟
    "name":"The Guardian (U.S. edition)",
    "url":"http://www.theguardian.com/us",
    "socialHandlers":{ ⊟
        "twitter":"@guardian"
    }
},
"startUrls":[ ⊟
    "http://www.theguardian.com/us"
],
"sectionTpls":[ ⊕ ],
"articleTpls":[ ⊟
    { ⊟
        "urlPatterns":[ ⊟
            "(?<wordset>([a-zA-Z]{1,}[:\\-./]{1,}){1,5}[a-zA-Z]{1,})/(?<numberset>([0-9]{1,}){1,1})/(?<wordset1>([a-zA-Z]{1,}){1,1})/(?<numberset1>([0-
        ],
        "canonicalUrlXpath":"wrty:normalize-url(//link[@rel='canonical']/@href)",
        "titleXpath":"wrty:normalize-space(//h1[contains(@class,'content__headline')])",
        "alternativeUrls":[ ⊕ ],
        "bylineXpath":"//span[@itemprop='name']",
        "datePublished":{ ⊟
            "datelineXpath":"//meta[@property='article:published_time']/@content"
        },
        "ingressXpath":"wrty:normalize-space(//div[@class='hide-on-mobile']//p[1])",
        "keywordsXpath":"//head//@content",
        "contentXpath":{ ⊟
            "includeXpath":"wrty:normalize-space(wrty:string-join(//div[@class='content__article-body from-content-api js-article__body']//node()[self:
        },
        "engagementPatterns":[ ⊟
            { ⊟
                "valueXpath":"//span[@class='commentcount2__value tone-colour js_commentcount_actualvalue']",
                "type":"comments"
            },
            { ⊕ }
```

Why different wrapper format?

- **OXPath**: perfect for interactive, search engine style websites

- However: in **media intelligence** – freshness of data is critical

  - 30min maximum between publishing and indexing time

  - (almost) every article has an indexable, unique URL

  - large variety of different article templates

- **Decompose** OXPath wrapper into **single page segments**

  - memorise set of section pages encountered in a run

  - recrawl stored section pages in next run

    - to find new article (& section) pages

Vertical 4:

Company Extractors

Results

PERSON_IMAGE
Jorn Lyseggen
RECORD_VALUE AGE
Chief Executive Officer and Founder
PERSON_ROLE   PERSON_ROLE

PERSON_IMAGE
Martin Hernandez
RECORD_VALUE AGE
Chief Financial Officer
PERSON_ROLE

PERSON_IMAGE
Kaveh Rostampor
RECORD_VALUE AGE
Executive Director, Americas
PERSON_ROLE

PERSON_IMAGE
Paal Larsen
RECORD_VALUE AGE
Executive Director, EMEA
PERSON_ROLE

PERSON_IMAGE
John Box
RECORD_VALUE AGE
Executive Director, APAC
PERSON_ROLE

PERSON_IMAGE
Niklas de Besche
RECORD_VALUE AGE
Executive Director, Products
PERSON_ROLE

Results

Ann Arbor
2300 Traverwood Dr.
Ann Arbor, MI 48105
United States
Phone: +1 734-332-6500
Directions

Atlanta
10 10th Street NE
Atlanta, GA 30309
United States
Phone: +1 404-487-9000
Directions

Austin
9606 North MoPac Expressway
Austin, TX 78759
United States
Phone: +1 512-343-5283
Directions

Birmingham
325 North Old Woodward
Birmingham, MI 48009
United States
Directions

Boulder
2600 Pearl Street
Boulder, CO 80302
United States
Phone: +1 303-245-0086
Directions

Cambridge
355 Main Street
Cambridge, MA 02142
United States
Phone: +1 617-575-1300
Directions

Chapel Hill
200 West Franklin Street

Chicago
320 N. Morgan, Suite 600

Company Extraction: **Goals**

- Given **only** a company website

  - Extract as much relevant information from <span style="color:red">structured sources</span>

    - executive team, locations, subsidiaries, ...

  - Identify unstructured sources

    - press releases, financial reports, ...

- **Scale** to millions of companies in multiple languages

Results

Company Extraction: **Results**

% of companies			
CEO & execs	description	location	logo
85%	**74%**	**82%**	**89%**
58%	**79%**	**69%**	**74%**

recall
present

Restaurant locations

Don't believe us? You aren't the first – major US technology company

- **Need:** US restaurant locations (including chains) for check-ins

- **Problem:** existing location databases incomplete and full of errors

- **Want:** Get that data from the "authoritative" source, i.e.,

  - the restaurant (chain) websites

- They evaluated state-of-the-art – most solutions to crude

  - Settled on scrapy, but: 2 months for top 20 US chains

  - Very worried about maintenance

25 LOCATIONS FOUND NEAR YOU

1. BEACH HAVEN II
600 N. Bay Avenue,
Beach Haven, NJ 08008
609-492-6300
Send Address To E-Mail|Mobile
8.22 miles
Get Driving Directions
Gift Cards Sold, Gift Cards Accepted

2. BORGATA
1 Borgata Way,
Atlantic City, NJ 08401
609-317-8206
Send Address To E-Mail|Mobile
18.04 miles
Get Driving Directions
Scoop Website

Catering, Gift Cards Sold, Gift Cards

Address or Postcode

| Bridge Rd, Little Egg Harbor Township. | United States of America | SEARCH |

☐ Offer catering   ☐ Offer ice cream cakes

© OpenStreetMap contributors

NAME	STREET ADDRESS	LOCALITY	STATE	POSTCODE	PHONE
BEACH HAVEN II	600 N. Bay Avenue	Beach Haven	NJ	08008	609-492-6300
BORGATA	1 Borgata Way	Atlantic city	NJ	08401	609-317-8206
30TH STREET STATION	30TH STREET STATION	Philadelphia	PA	19104	215-222-2996

Restaurant location: **Results**

- After <span style="color:red">1 month</span> applying Wrapidity technology:

    - over 300 US chains, over <span style="color:red">100k</span> websites

    - more than <span style="color:red">3M</span> locations in total

    - fully automated maintenance for those sources

coverage **85%**   precision **95%**

- But: *they still didn't believe*

    - hired Accenture to assess quality of the data

    - result: **over 97% precision**

Results

Restaurant location: **Independent Evaluation**

835	**Present**	and **correct** data & extraction	but **wrong** extraction	but **wrong** data	but **raters** disagree
city	100%	**99.3%**	**0.7%**	0.0%	0.0%
street	100%	**96.4%**	**1.7%**	1.9%	0.0%
postcode	99%	**97.1%**	**0.1%**	0.0%	2.8%
latlong	89%	**99.7%**	**0.1%**	0.0%	0.1%
hours	47%	**98.2%**	**0.0%**	1.3%	0.5%
name	100%	**99.5%**	**0.5%**	0.0%	0.0%
phone	86%	**98.7%**	**1.3%**	0.0%	0.0%
category	100%	**98.9%**	**0.0%**	0.0%	1.1%
	90%	**98.5%**	**0.5%**	0.4%	0.6%

This evaluation is done by independent, external evaluators on a sample of 1000 locations.

Results

Summary

~50k sources, 30 min recrawl interval

87%

90%

over 300 US chains, over 100k websites;
more than 3M locations

FORK, KNIFE & GLASS
DINNER CLUB

over 30 attributes, where present; 1M+ company's site crawl ongoing

91%

COMPANY NAME
Company Slogan

Approach

Web-Scale Wrapper Induction

- We need to **scale to the web**

  - minimize supervision per source

- But: we can afford **prior knowledge**

  - about entities and attributes

  - mostly in form of known **labels** & **instances** and "**appearance**"

    - expressed as Gazetteers or rules for local, textual information

    - higher-level rules or classifiers for complex structures

Web-Scale Wrapper Induction

- **Problem:** application of prior knowledge is costly & noisy

  - wrapper induction to generalise to other pages of site

  - "template" hypothesis

- **Solution:** Generate "wrapper" program from examples

  - then apply to all pages of a site

  - when to apply which extractor

- **Full site extraction** needs to also deal with

  - Interactivity such as pagination & form filling (deep web)

  - Detecting complex structures such as lists, tables, …

Approach

Desired Schema

Source URL

Ontology Design
5% | D

Instance Collection
15% | T.D

JAPE Design
0% | E

Heuristic Adaptation
0% | E

Source URL

Wrty Ontology

Web Acess Layer
95% | -

SNER
70% | T

Block classification
93% | T

Exploration
90% | T

Form filling
95% | D

Segmentation & alignment
92% | D

Induction
95% | D

Validation loop test & training corpus

Wrapper

Extraction
100% | -

Cleaning
90% | D

Repair
90% | D

Text Analysis
70% | D

Clean Data

Monitoring Data

Exception Data

Re-Scheduling

Re-Induction

Reporting

Exploration: Self-Adaptive

- Self-adaptive, dynamic exploration plans
  - planers expressed as guarded FSTs
  - with Datalog rules as guards
  - 1000's of unique exploration plans

NER for DOMs: Labels, structure, …

Labels and instances, visible and invisible (HTML structure, Javascript values)

Approach

```
<div class="icon first">
  <img src="…/bdes.jpg" alt="Bedrooms" title="Bedrooms">
  <br>8
</div>
<div class="icon">
  <img src="…/bath.jpg" alt="Bathrooms" title="Bathrooms">
  <br>4
</div>
```

Form understanding

- Sub-problems: form **labeling**, form **segmentation**, **classification**

- **Combines** structural, textual, visual, and semantic clues

  - structural = structure of the DOM, e.g., distance

  - visual = rendering of the form, e.g., for alignment

  - textual = detectors for a vertical's types (e.g., "LHR")

  - semantic = class, id, ... with semantic labels (e.g. "finput_dest")

- **Polynomial time** labeling, grouping, and classification algorithm

ICQ dataset	HA [14]	ExQ [41]	StatParser [36]	DIADEM [17]
F_1 for labeling	92%	96%	96%	98%

Approach

Pick a Path: Wrapper induction

- **Pick** robust, "semantic" paths
  - less affected by changes
  - over time and within a template
- Suitable as "foundation" for
  - **template discovery**
- E.g.: Select the director
  - Firebug ("canonical" XPath)

```
/html[1]/body[1]/ ... /div[4]/a[1]/span[1]
```

  - Ours:

```
//div[starts-with(.,"Director:")]//span[(@class="itemprop")
```

Wrapper Repair

Postcode	Phone	Locality	State	Street Address
San Diego, CA 92101	619-234-1802	San Diego, CA 92101	<NULL>	471 Horton Plaza, near Westland park
Boise, ID 83702	208-342-1992	Boise, ID 83702	<NULL>	103 North 10th Street
Portland, OR 97209	503-796-3033	Portland, OR 97209	<NULL>	301 NW 10th Avenue, near the Fish Market

Postcode	Phone	Locality	State	Street Address
92101	619-234-1802	San Diego	CA	471 Horton Plaza
83702	208-342-1992	Boise	ID	103 North 10th Street
97209	503-796-3033	Portland	OR	301 NW 10th Avenue

San Diego, CA 92101

Boise, ID 83702

Portland, OR 97209

Locality **State** **Postcode** **Phone**

Wrapper Repair

- **Joint repair** of wrapper and output data (relation)

- Problem related to table segmentation problem

  - generally NP-complete

  - but we show that it's polynomial under atomic misplacement

- **Atomic misplacement:** attribute value is either

  - entirely misplaced, or

  - its fragments are in adjacent fields

Results

① ROSeAnn (VLDB'14)
Entity extraction from DOMs

② OPAL (WWW'12, VLDBJ'13)
Form understanding & filling

③ AMBER (ICWE'11)
Record identification for lists

④ OXPath (VLDB'11, VLDBJ'13)
Extraction language

⑤ Robust XPaths (SIGMOD'16)
Change-resistant wrappers

⑥ Oxtractor (Coling'16)
Attribute extraction

⑦ WaDaR (ICDE'16)
Joint **wrapper & relation repair**

⑧ VADA (EDBT'16)
Wrangling of extracted data (in progress)

⑥ DIADEM (VLDB'14)
World-first accurate, automatic **full-site extraction system**

Meltwater: Who are we?

FOUNDED 2001
in Oslo, Norway

HEADQUARTERS
in San Francisco

Fairhair

Montreal
Toronto
Chicago
Manchester
Boston
New York
Washington, DC
San Francisco
Santa Monica
San Diego
Charlotte
Atlanta
Austin
Miami

Edinburgh
Amsterdam
Cardiff
London
Paris

Oslo
Stockholm
Helsinki
Gothenburg
Copenhagen
Hamburg
Berlin
Munich
Vienna
Budapest

Beijing
Tokyo
Shanghai
New Delhi
Hong Kong
Dubai
Kuala Lumpur
Singapore
Accra

Sao Paulo
Buenos Aires
Cape Town

Brisbane
Sydney
Melbourne

1500 employees
worldwide

26,000 Business customers
in **108** countries

60 offices across
27 Countries

Meltwater: **Media Intelligence**

Sources: Editorial, Social, Broadcasts

media exposure

trends

influencers

sentiment analysis

More than 300k different types of user queries

Meltwater: **In Numbers**

~200B indexed documents

- Crawlers fetch ~**3.3M** articles/day from 190k editorial sources
  - re-crawled every 30 minutes
- With the social fire hoses we go up to 30M docs/ day.

Name	Country	Language	Documents ▼
Notiradar	Mexico	Spanish	40700
福建东南新闻网	China	Chinese (simpl.)	23182
中工网	China	Chinese (simpl.)	20953
매일경제	Korea, Republic Of	Korean	20191
جستجوگر اخبار تی نیوز	Iran (islamic Republic Of)	Persian	18055
Match 生活網	Taiwan	Chinese (trad.)	17512
47NEWS	Japan	Japanese	16966
Nambia Press Agency	Namibia	English	9521
中金在线 - 外汇网	China	Chinese (simpl.)	7957
Onet.pl	Poland	English	7127

cluster 1 (Scandinavia), cluster 2 (EU, E.Asia), cluster 3 (UK), cluster 4 (...), cluster 6 (N. America), cluster 7 (N. America), cluster 8 (AU, S. America)

Meltwater: **Existing Technology Stack**

Ingestion:

o Social media hoses (partnerships)
o Editorial/News (partnerships + web crawling)
o Broadcasts (views on the above)

Enrichments (15 languages):

o Text categorization (*topic*, *language*)
o NERD (*person*, *location*, *organization*, …)
o NED (https://en.wikipedia.org/wiki/Tim_Cook)
o Sentiment Analysis

Storage and search:

o Elastic search
o Rabbit MQ (distributed queues)
o AWS

Media Intelligence applications (Custom)

o Boolean queries (*keywords / entities*)
o Counters
o Aggregates
o Drill downs / pivoting

| Sources | Hoses + crawlers | Enrichments | Elastic search | Intelligence apps |

Vision: **Insight** Building on **Outside** Data

Build a world class AI platform for a new software category

Outside Insight

Fairhair: People & Community

5 Data Science **Research offices**

University **collaborations**

6 Data Science Hubs (co-working spaces)

- London
- San Francisco
- Singapore
- Sydney
- Berlin
- New York

Meltwater Entrepreneurial School of Technology

- campuses in Ghana and Nigeria
- it's a school for African entrepreneurs
- it's an incubator (33 startups)
- it's a networking hub

Fairhair's AI First Approach

Step 1: Outside Data acquisition & making it available in a form that's crunchable.
Step 2: Make Data Science (data, algorithms, infrastructure, tools) power everything
Step 3: We can't foresee all uses of data and insights → Developer APIs & integrations

APIs & Services	**Search, Alerting, Analytics, Reporting** Building blocks to leverage the platform
Context Building	**Knowledge graph** Enable cognitive applications on top of our Data by connecting the dots
Enrichments & Analysis	**Data science platform** Enrich, Analyze & Build Insights by interoperating with all major players
Data acquisition	**AI driven crawling** Bring high quality Outside Data to our repository with minimal human effort

Fairhair's AI Crawlers

Traditional scraping requires a **huge human effort:**

- **Code** wrappers for each source, e.g., in Scrapy or MW's source configurations
- **Visually** testing and support tool (ala Connotate, Mozenda, …)
- **Automatic** scraping for small number of fixed data types (ala Diffbot), e.g., Microdata
- Meltwater (old): ~50 "source engineers" maintaining manual wrappers
  - sources failing at a rate of 100's per week, 1-2h to fix each source effectively

80-90% lower
human **effort**

without loss in quality
compared with
state-of-the-art

3-10x more
attributes

and domains than existing
automated solutions and
affordable supervised one

10-100x
more sources

e.g., 100k+ restaurant websites,
300k+ news sources, 1M+ of
company websites

Data and Content Lake

Factual information: <span style="color:red">wherefrom</span>?

Need to **restrict the domain**: focus on the <span style="color:red">corporate domain</span>, i.e., companies, people, products, …

Editorial	Social Media	Job Postings
Patents & Trademarks	Gov Data	Company Websites
Credit Ratings	SEO Filings	● ● ●

Disk used on data nodes

513 TiB

elasticsearch

APACHE Spark™

HIVE

WolframAlpha

EC2 cluster with 2.8k vCPU, 21TB RAM, 630TB SSD

Data and Content Lake

Linguistic enrichments to support semantic retrieval and fact extraction

Summary:
- Scalable and distributed dynamic enrichment workflows
- CRFs for NER, PageRank (variant) for NED, CNNs and LSTMs for Relation/Event extraction, sentiment analysis
- TensorFlow and GPUs for training infrastructure

Enrichments

We can't foresee all uses of our data: Developer APIs to Integrate and orchestrate third party tools.

Personalization is key in Data Science: A flexible data wrangling infrastructure is required.

- Interoperate with state-of-the-art external enrichments
- Chain multiple external enrichments
- Train your own models!

Connectors to Internal Systems

Goal is to join Outside Insights with Internal Data and workflows

Data Ingestion & Insights Delivery by setting up simple schema mappers

Knowledge Graph

Wait... did you say PageRank, triples? So do you have a (Knowledge) Graph?

Content:
- o   Companies
- o   Brands
- o   Products
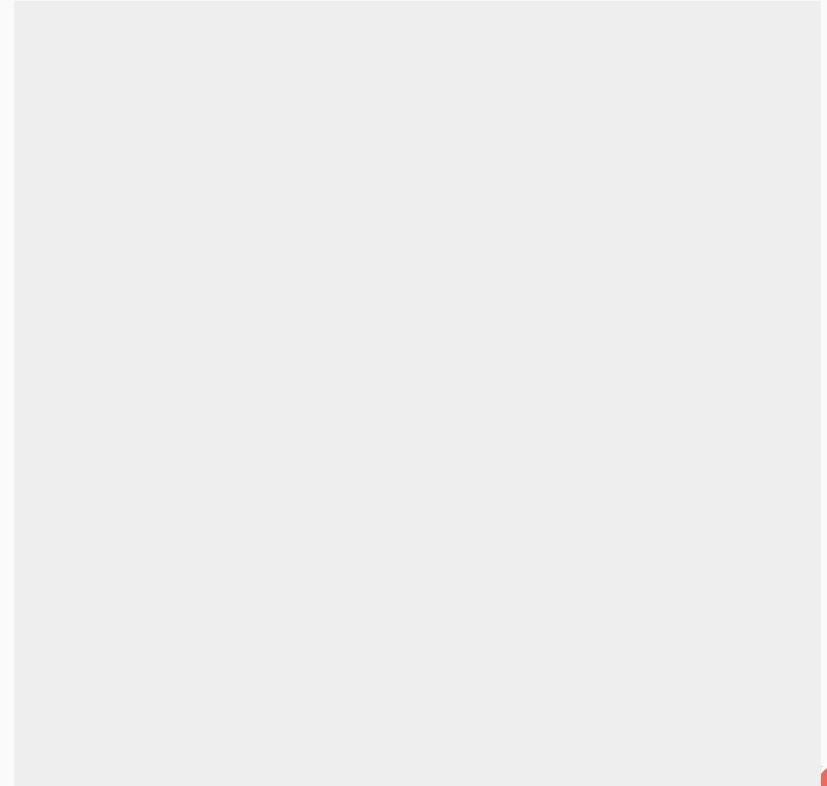- o   Key people
- o   Influencers

Goals:
- o   Relate facts
- o   Data mining
- o   Cognitive applications (higher-order reasoning)

Challenges:
- o   Data Cleaning
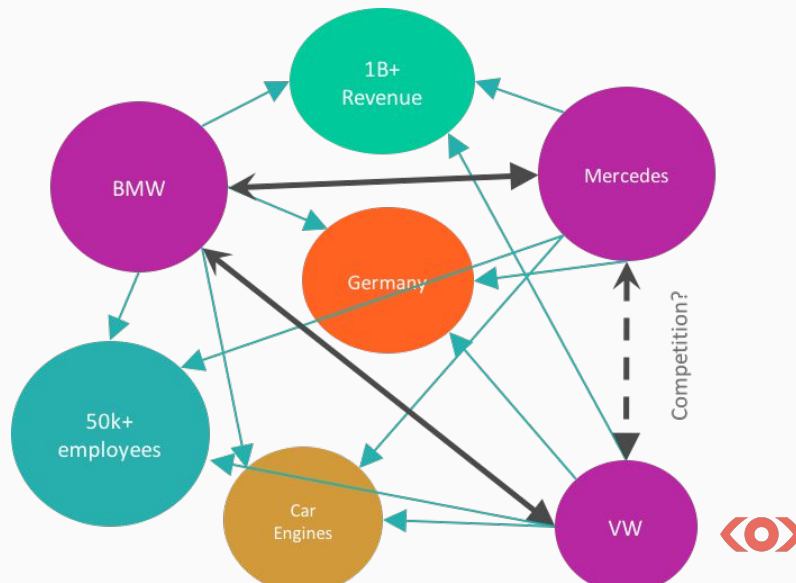- o   Data deduplication / integration
- o   Truth Finding

Cognitive Applications

Infer high-level insights from a set of extracted events/facts.

- Competitor
- Customer
- Investment
- Lawsuit/Litigation
- Partnership

- Supplier
- Acquisition
- Out/under performance
- Expanding Operations
- Compliance

- Funding Developments
- Leadership Changes
- New Offerings
- Bankruptcy
- Restructuring, Cost Cutting

Insight discovery:

- Rule/Graph mining (data cleaning)
    - GPAR (VLDB '15)
    - RUDIK (internal, paper submitted)
- Link prediction (data enrichment, fact checking)
    - Path Ranking Algorithms (PRA)
    - Probabilistic Soft Logic (PSL)
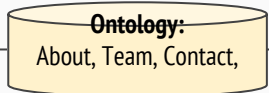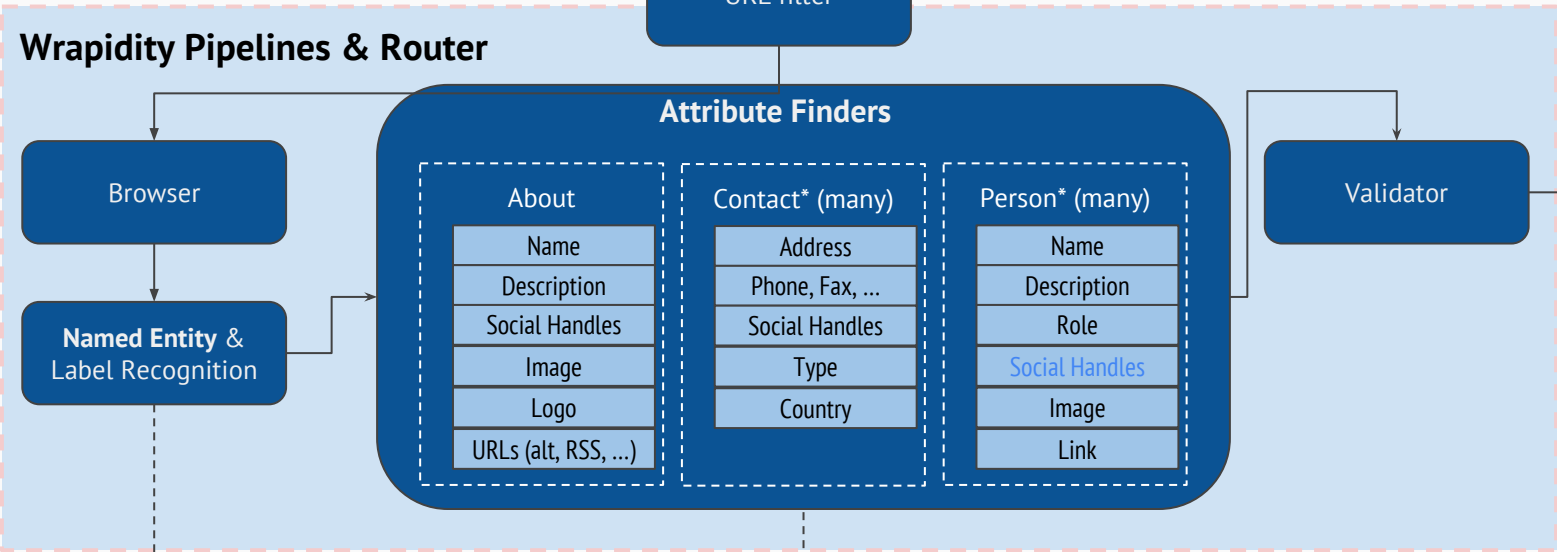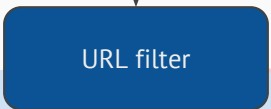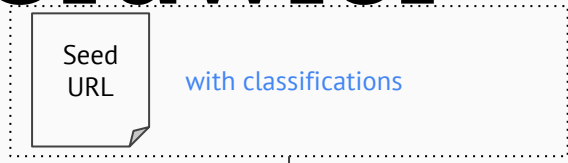   - Snorkel (Stanford Collaboration)

Questions

More?

Company Crawler