

Agenda – 21.11.2022

Presentation and Discussion: Exercise Of Last Lecture

ETL Workflow: Airflow

Practical Exam Discussion

Grading, Timeline, Deliverables, Presentation Procedure, Exam Topics

Work On Exam

01

Work on practical exam.



03

02

Schedule

	Lecture Topic	HandsOn
10.10.2022 13:00-15:45 Ro. N/A	About This Lecture, Introduction to Big Data	Setup Google Cloud, Create Own Hadoop Cluster and Run MapReduce
17.10.2022 13:00-15:45 Ro. N/A	(Non-)Functional Requirements Of Distributed Data-Systems, Data Models and Access	Hive and HiveQL
24.10.2022 13:00-15:45 Ro. N/A	-	-
31.10.2022 13:00-15:45 Ro. N/A	Challenges Of Distributed Data Systems: Partitioning	HiveQL via JDBC, Data Partitioning (with HDFS and Hive)
07.11.2022 13:15-15:45 Ro. N/A	Challenges Of Distributed Data Systems: Replication	Spark, Scala, PySpark and Jupyter Notebooks
14.11.2022 13:00-15:45 Ro. N/A	ETL Workflow and Automation & Batch and Stream Processing	Airflow
21.11.2022 13:00-15:45 Ro. N/A	Practical Exam	Work On Practical Exam
28.11.2022 13:00-15:45 Ro. N/A	Practical Exam	Work On Practical Exam
05.12.2022 13:00-15:45 Ro. N/A	Practical Exam Presentation	
12.12.2022 13:00-15:45 Ro. N/A	Practical Exam Presentation	





Practical Exam - Grading

Grade/Percentage:

grade will be mixed with grade of another lecture

therefore, the rating won't be a grade (1-6) but a

percentage:

Percentage	Grade	
100%	1.0	
50%	4.0	

Timeline:

02.12.2022 23:59 All deliverables (next slide) will be delivered to following email address (DropBox, Google-Drive...):

contact@marcel-mittelstaedt.com

bitte eintrager bitte eintrage 60 Veranstaltung: bitte eintragen Matrikelnummer Normiert

Studiengang Informatik Klausurergebnisse (Punkte)

05.12./12.12.2022 13:15-15:45 Presentation of Practical Exam



Practical Exam - Deliverables

Deliverables:

- A simple Documentation:
 - Explanation of whole ETL Workflow
 - List of Jobs/Transformations in Case of PDI or DAGs and Steps in Case of Airflow
 - Short description of the purpose of each job/transformation or task and applied business rules
 - all PDI Jobs, Transformations and related files (ktr, kjb, kettle.properties, shared.xml... files)
 - All Airflow DAGs and tasks
- All Scripts (e.g. Download) or other external applications called within PDI
- All Airflow DAGs, Python Files etc.
- All DDLs (CREATE Table...):
 - One file for each table
 - Table name = File Name, e.g.:



- Depending on Exam Type:
 - Code of Frontend Application and related Database (DDLs) or
 - Calculated KPIs



Practical Exam - Presentation

Procedure:

- Start ETL Workflow
- 2. During execution:
 - Quickly explain data source
 - API
 - Data Structure
 - Approach for gathering data
 - Quickly Explain whole ETL Workflow
 - Explain Idea and purpose of each Job/Transformation
 - External ressources/scripts (e.g. download)
 - Explain Data Model (Raw Layer, Final Layer, simple Frontend)
- After execution:
 - Depending on Exam:
 - Demo of simple Frontend application or
 - Explanation of calculated KPIs







Spotify provides an API for basic track information:

- ID, e.g. 2IEcSduKEXEK5KJ9hJzlCz
- name, e.g. Gloana Bauer (Teenage Dirtbag)
- artist, e.g. D' Hundskrippln
- ...

as well as related audio features:

_	energy , e.g. 0.454	Energy is a measure from 0.0 to 1.	.0 and represents a perceptual measure of

intensity and activity. Typically, energetic tracks feel fast, loud, and noisy.

For example, death metal has high energy, while a Bach prelude scores low on

the scale.

speechiness, e.g. 0.0388 Speechiness detects the presence of spoken words in a track. The more exclu

sively speech-like the recording (e.g. talk show, audio book, poetry),

the closer to 1.0 the attribute value.

loudness, e.g. -8.758 The overall loudness of a track in decibels (dB). Loudness values are averaged.

tempo, e.g. 97.532 The overall estimated tempo of a track in beats per minute (BPM)

A confidence measure from 0.0 to 1.0 of whether the track is acoustic.

- ...

https://developer.spotify.com/documentation/web-api/reference/#/operations/get-audio-features



acousticness, e.g. 0.268

We want to make use of those audio features to automatically assign each track to a certain category:

MetalClassicRockElectroPodcastSoul

Workflow:

- Query data from Spotify API
- Save raw data (JSON files) to HDFS
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Calculate categories (Metal, Classic, Rock, ...)
- Join track information and audio festures and save everything to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple HTML Frontend which reads from end-user database and displays result
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically

Vocal HipHop

j	Communities	A41 4	0-4
טו	Song Title	Artist	Category
42	Savior	Rise Against	Rock
11	Ham Kummst	Seiler u Speer	Rock
13	No One	Alicia Keys	Soul
37	I Try	Macy Gray	Soul
55	Metalingus	Alter Bridge	Metal
56	The Trooper	Iron Maidon	Metal
77	Du	Cro	HipHop
88	Solala	Blumentopf	HipHop



Dataflow: 1. Get Track Information

curl -X "GET" "https://api.spotify.com/v1/tracks/2IEcSduKEXEK5KJ9h JzICz?market=DE" -H "Accept: application/json" -H "Content-Type: application/json" -H "Authorization: Bearer BBB0v7vCOHXxaUAshFUVIF bozLDO_ysq8cPb4wYR3oko_JfDcrUSEsy0Mq6P4cu5vvS0ljrn6R24ra ME8o4qa2XNy02IhGGCufMgwgPtf43s2OoAcbfJUfcsXA1-dpW19_x_ 3rG75ADnA4dlr25"



"href": "https://api.spotify.com/v1/tracks/2IEcSduKEXEK5KJ9hJzlCz", "id": "2IEc



/user/hadoop/spotify/track_data/raw/... /user/hadoop/spotify/audio_features/raw/...

https://developer.spotify.com/documentation/web-api/reference/#/operations/get-track

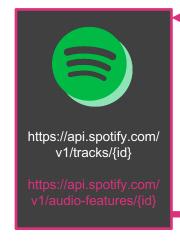
SduKEXEK5KJ9hJzlCz

"release date": "2016-07-01"



Dataflow: 2. Get Track Audio Features

curl -X "GET" "https://api.spotify.com/v1/audio-features/**2IEcSduKEXE K5KJ9hJzICz**" -H "Accept: application/json" -H "Content-Type: application/json" -H "Authorization: Bearer BBB0v7vCOHXxaUAshFUVIFbozL
DO_ysq8cPb4wYR3oko_JfDcrUSEsy0Mq6P4cu5vvS0ljrn6R24raME8
o4qa2XNy02lhGGCufMgwgPtf43s2OoAcbflJUfcsXA1-dpW19_x_3rG7
5ADnA4dlr25"



```
{
  "danceability": 0.685,
  "energy": 0.454,
  "loudness": -8.758,
  "speechiness": 0.0388,
  "acousticness": 0.268,
  "instrumentalness": 0,
  "liveness": 0.202,
  "valence": 0.833,
  "tempo": 97.532,
  "id": "2IEcSduKEXEK5KJ9hJzICz",
  "uri": "spotify:track:2IEcSduKEXEK5KJ9hJzICz",
  "analysis_url": "https://api.spotify.com/v1/audio-analysis/2IEcSduKEXEK5KJ9hJzICz",
  "duration_ms": 226852,
  "time_signature": 4
}
```



/user/hadoop/spotify/track_data/raw/... /user/hadoop/spotify/audio_features/raw/...





Dataflow: 3. Raw To Final Transfer



/user/hadoop/spotify/track_data/**raw**/... /user/hadoop/spotify/audio_features/**raw**/...



- move data from raw to final directory
- optimize and reduce data structure for analytical/query purposes (JSON to tabular, only needed attributes etc.)
- remove duplicates if necessary



/user/hadoop/spotify/track_data/final/...
/user/hadoop/spotify/audio features/final/...



Dataflow: 4. Run Analysis and Save Results



/user/hadoop/spotify/track_data/final/... /user/hadoop/spotify/audio_features/final/...



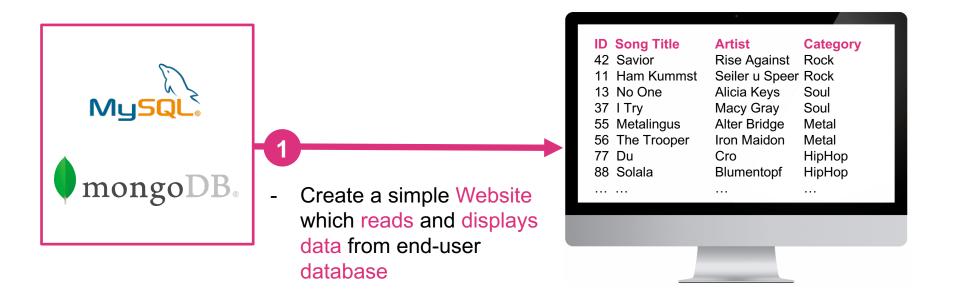
- calculate categories for each track, using *Hive*, *Python*, *Spark* or *PySpark*
- join track and audio feature data
- save everything to a enduser database (e.g. MySQL, M ongoDB)







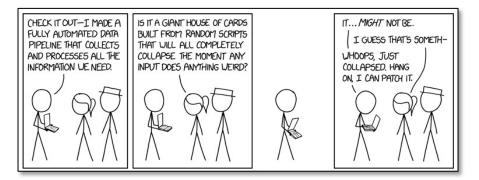
Dataflow: 5. Provide Simple Web Interface





XKCD provides regularly comics:

- https://xkcd.com/
- JSON API: https://xkcd.com/2054/info.0.json



https://xkcd.com/2054/

```
{
    "month": "10",
    "num": 2054,
    "link": "",
    "year": "2018",
    "news": "",
    "safe_title": "Data Pipeline",
    "transcript": "",
    "alt": "\"Is the pipeline literally running from your laptop?\" \"Don't b
    e silly, my laptop disconnects far too often to host a service we rel
    y on. It's running on my phone.\"", "img": "https://imgs.xkcd.com/comi
    cs/data_pipeline.png",
    "title": "Data Pipeline",
    "day": "3"
}
```

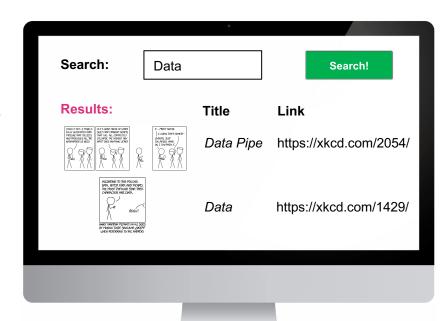
2054.json



We want to make use of this data to build a searchable database for XKCD comics.

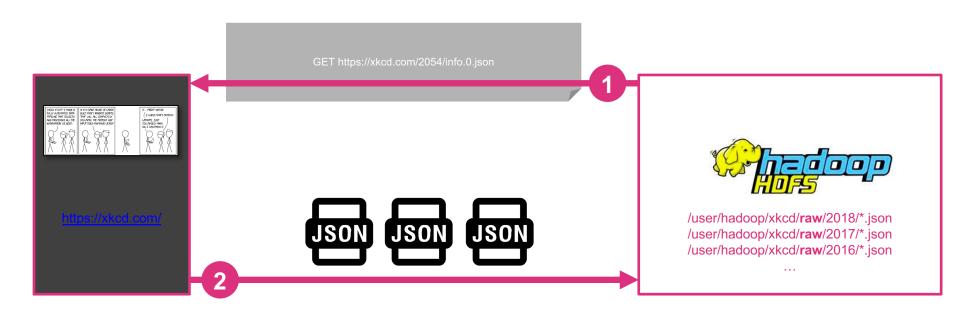
Workflow:

- Gather data from xkcd.com
- Save raw data (JSON files) to HDFS (partitioned by year, e.g. 2018, 2017, 2016...)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Export xkcd data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple **HTML Frontend** which is able to:
 - read from end-user database
 - process user input (search phrase...)
 - checks against xkcd data in enduser database
 - Display result (comics containing search phrase)
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically





Dataflow: 1. Get XKCD Data



Dataflow: 2. Raw To Final Transfer

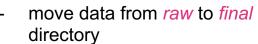


/user/hadoop/xkcd/**raw**/*.csv









- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/xkcd/**final**/*



Dataflow: 3. Enhance Data And Save Results





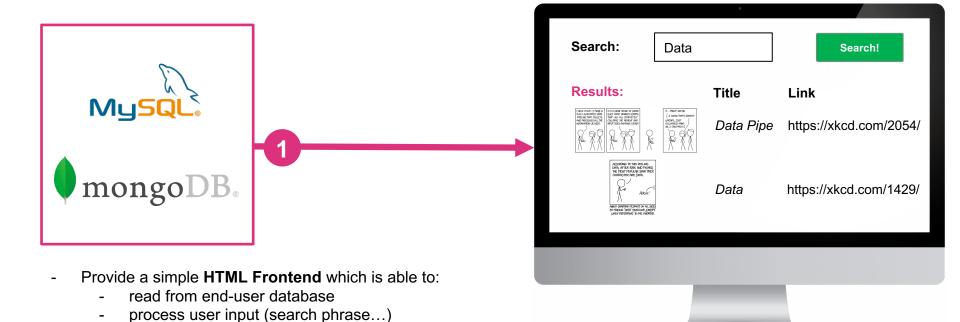
- enhance data (e.g. for later querying)
- use *Hive*, *Spark* or *PySpark*
- save everything to a enduser database (e.g. MySQL, MongoDB)







Dataflow: 4. Provide Simple Web Interface



user database

checks against xkcd data in end-

Display result (comics containing search phrase)



Maxmind.com provides regulary exports of worldwide IP and Geolocation data:

https://dev.maxmind.com/geoip/geolite2-free-geolocation-data

```
curl -s http://ifconfig.me
88.130.59.75

netw_rk,geoname_id,registered_country_geoname_id,represented_country_geoname_id,is_anonymous_proxy,is_satellite_provider,postal_code,latitude,longitude,accura cy_ra_lius
88.130.59.0/24,2939623,2921044,,0,0,85221,48.2600,11.4340,50
[...]
```

GeoLite2-City-Blocks-IPv4.csv

```
geonar e_id,locale_code,continent_code,continent_name,country_iso_code,country_name,subdivision_1_iso_code,subdivision_1_name,subdivision_2_iso_code,subdivision_2_name,city_name,metro_code,time_zone,is_in_european_union
32_535,de,EU,Europa,DE,Deutschland,BY,Bayern,,Höhenkirchen-Siegertsbrunn,,Europe/Berlin,1
2939623,de,EU,Europa,DE,Deutschland,BY,Bayern,,Dachau,Europe/Berlin,1
3207410,de,EU,Europa,DE,Deutschland,BY,Bayern,,Rödental,,Europe/Berlin,1
3207412,de,EU,Europa,DE,Deutschland,BY,Bayern,,Röslau,,Europe/Berlin,1
3208324,de,EU,Europa,DE,Deutschland,BY,Bayern,,,Asbach-Bäumenheim,,Europe/Berlin,1
[...]
```

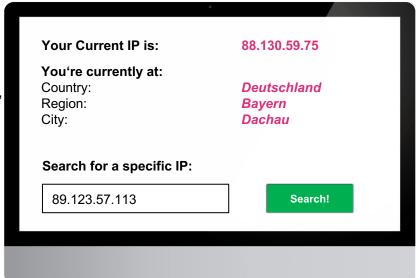
GeoLite2-City-Locations-[XX].csv



We want to make use of this data to build a real time IP-Geolocation resolution as well as a searchable database for lps and related Geolocations.

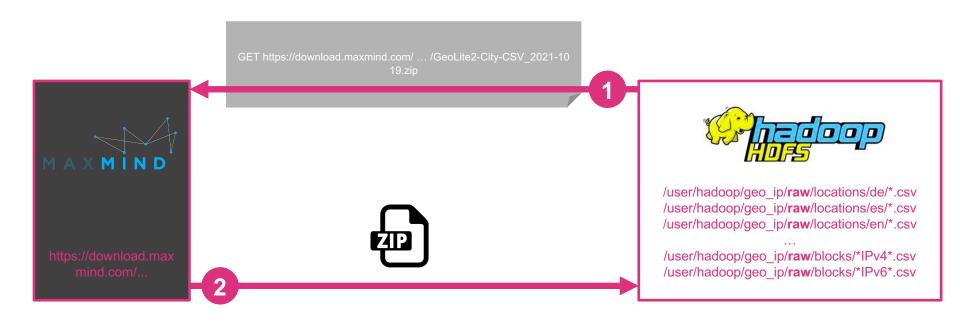
Workflow:

- Gather data from maxmind.com
- Save raw data (CSV files) to HDFS (partitioned by country code,
 e.g. de, es, en...)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Export Geolite2 data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple **HTML Frontend** which is able to:
 - determine a user's IP address, lookup and show
 Geolocation
 - process user input (IP...) and check against enduser database
 - Display result Geolocation
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically





Dataflow: 1. Get Geolite2 Data





Dataflow: 2. Raw To Final Transfer



/user/hadoop/geo_ip/**raw**/locations/de/*.csv /user/hadoop/geo_ip/**raw**/locations/es/*.csv /user/hadoop/geo_ip/**raw**/locations/en/*.csv

/user/hadoop/geo_ip/**raw**/blocks/*IPv4*.csv /user/hadoop/geo_ip/**raw**/blocks/*IPv6*.csv









- move data from raw to final directory
- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/geo_ip/final/locations/de /user/hadoop/geo_ip/final/locations/es/ /user/hadoop/geo_ip/final/locations/en/

/user/hadoop/geo_ip/final/blocks/*IPv4* /user/hadoop/geo_ip/final/blocks/*IPv6*



Dataflow: 3. Enhance Data And Save Results



/user/hadoop/geo ip/final/locations/de /user/hadoop/geo ip/final/locations/es/ /user/hadoop/geo ip/final/locations/en/

/user/hadoop/geo ip/final/blocks/*IPv4* /user/hadoop/geo ip/final/blocks/*IPv6*









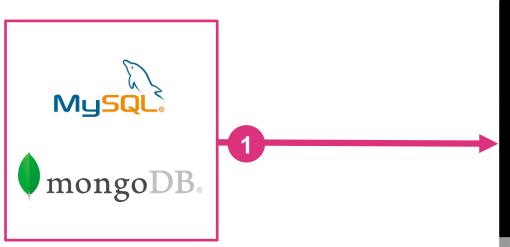
- enhance data (e.g. for later querying)
- use Hive, Python, Spark or **PySpark**
- save everything to a enduser database (e.g. MySQL, MongoDB)





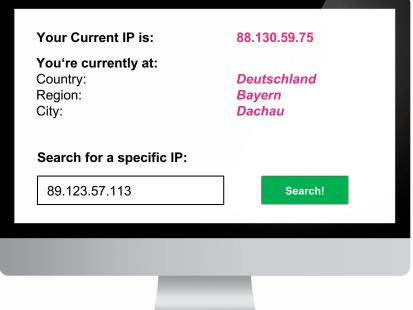


Dataflow: 4. Provide Simple Web Interface



Provide a simple **HTML Frontend** which is able to:

- determine a user's IP address, lookup and show Geolocation
- process user input (IP...) and check against enduser database
- Display result Geolocation

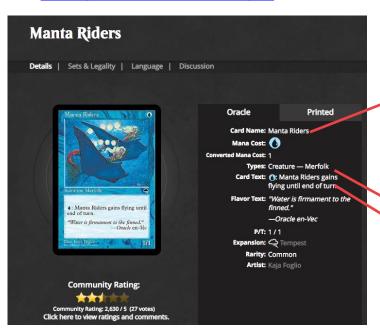






magicthegathering.io provides up-to-date information regarding all MTG trading cards available:

https://gatherer.wizards.com/Pages/



http://gatherer.wizards.com/Pages/Card/Details.aspx?multiverseid=4711

```
<div class="smallGreyMono" style="margin-top: 5px;">
  <br/>
<br/>
b class="ft"><b></b></b>
  <div id="ctl00 ctl00 ctl00 MainContent SubContent SubContent nameRow" class="row">
    <div class="label">
      Card Name:</div>
    <div class="value">
    Manta Riders</div>
  <div id="ctl00 ctl00 ctl00 MainContent SubContent SubContent manaRow" class="row manaRow">
    <div class="label" style="line-height: 25px;">
      Mana Cost </div>
    <div class="value">
      <imq src="/Handlers/Image.ashx?size=medium&amp;name=U&amp;type=symbol" alt="Blue" align="absbottom" /></div>
  <div id="ctl00 ctl00 ctl00 MainContent SubContent SubContent cmcRow" class="row">
    <div class="label" style="font-size: .7em;">
      Converted Mana Cost:</div>
    <div class="value">
      1</div>
  <div id="ctl00 ctl00 ctl00 MainContent SubContent SubContent typeRow" class="row">
    <div class="label">
      Types:</div>
    <div class="value">
      Creature — Merfolk</div>
  <div id="ctico ctl00 ctl00 MainContent SubContent SubContent textRow" class="row">
    <div class="lahel">
      Card Text:</div
    <div class="value">
      <div class="cardtextbox style="padding-left:10px;"><img src="/Handlers/Image.ashx?size=small&amp;name=U&amp;type=sy</p>
mbol" alt="Blue" align="absbottom" />: Manta Riders gains flying until end of turn.</div></div>
  </div>
```

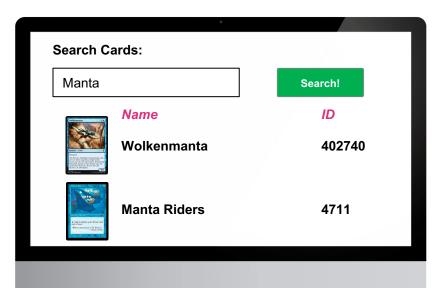


We want to make use of this data to build a searchable database of all MTG trading cards.

Workflow:

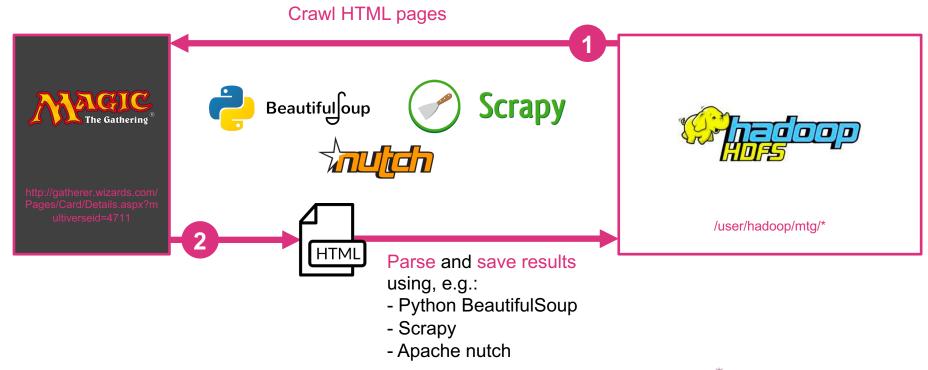
- Crawl data from gatherer.wizards.com
- Parse required information and save them to HDFS (queryable through Hive)
- Export MTG data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple HTML Frontend which is able to:
 - read from end-user database
 - process user input (card name, text or artist)
 - display search results







Dataflow: 1. Get MTG Data



Dataflow: 3. Enhance Data And Save Results





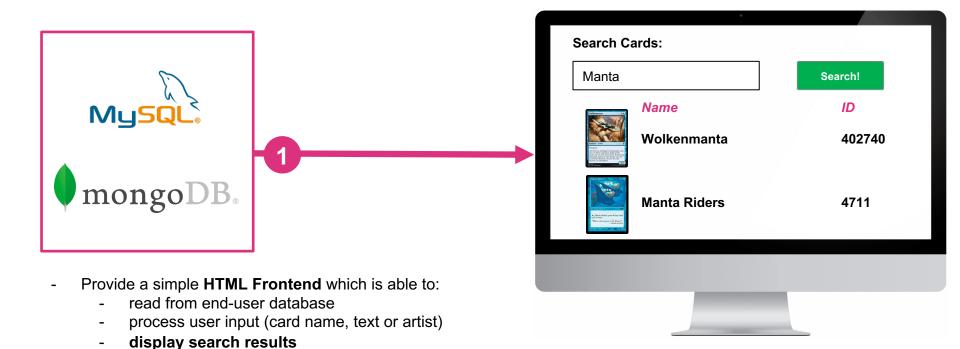
- enhance data (e.g. for later querying)
- use Hive, Python, Spark or PySpark
- save everything to a enduser database (e.g. MySQL, MongoDB)







Dataflow: 4. Provide Simple Web Interface





magicthegathering.io provides up-to-date information regarding all MTG trading cards available:

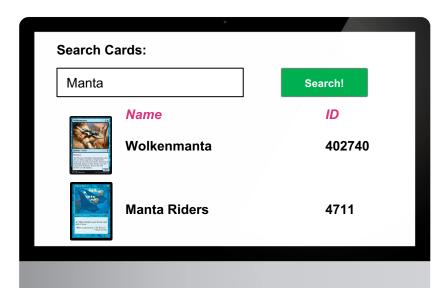
https://docs.magicthegathering.io/ "card":{ "name":"Manta Riders", E.g. https://api.magicthegathering.io/v1/cards/4711 "manaCost":"{U}", "cmc":1. "colors": "Blue" "colorIdentity":[Manta Riders "type": "Creature — Merfolk", "types":["Creature" "subtypes": "rarity":"Common", "set":"TMP". Summon Merfolk "setName":"Tempest", "text":"{U}: Manta Riders gains flying until end of turn.", . Manta Riders gains flying until "flavor":"\"Water is firmament to the finned.\"\n—Oracle en-Vec", end of turn. "artist":"Kaja Foglio", "Water is firmament to the finned." "power":"1", -Oracle en-Véc "toughness":"1", "layout": "normal", "multiverseid":4711. "imageUrl": "http://gatherer.wizards.com/Handlers/Image.ashx?multiverseid=4711&type= card",



We want to make use of this data to build a searchable database of all MTG trading cards.

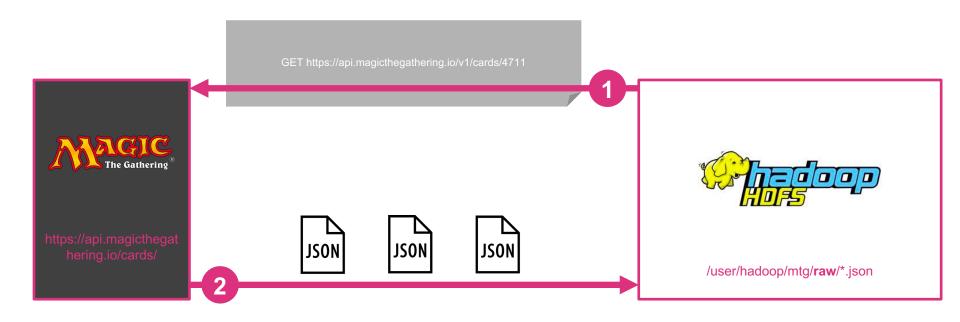
Workflow:

- Gather data from api.magicthegathering.io
- Save raw data (JSON files) to HDFS
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Export MTG data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple HTML Frontend which is able to:
 - read from end-user database
 - process user input (card name, text or artist)
 - display search results
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically





Dataflow: 1. Get MTG Data



Dataflow: 2. Raw To Final Transfer





- move data from raw to final directory
- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/mtg/final/*



Dataflow: 3. Enhance Data And Save Results





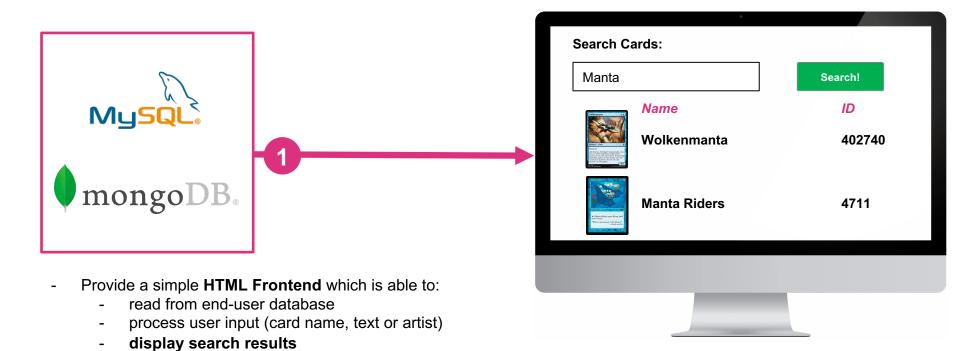
- enhance data (e.g. for later querying)
- use Hive, Spark or PySpark
- save everything to a enduser database (e.g. MySQL, MongoDB)







Dataflow: 4. Provide Simple Web Interface





NYC.gov provides monthly exports of NYC yellow taxi trip records:

- https://www1.nyc.gov/site/tlc/about/tlc-trip-record-data.page
- Latest Full Dumps:
 - https://d37ci6vzurychx.cloudfront.net/trip-data/yellow_tripdata_2022-01.parquet
 - https://d37ci6vzurychx.cloudfront.net/trip-data/yellow_tripdata_2022-02.parquet
 - https://d37ci6vzurychx.cloudfront.net/trip-data/yellow_tripdata_2022-03.parquet
 - ..

	VendorID	tpep_pickup_datet	time t	pep_dropoff_datetime	passenger_count	trip_distance	RatecodeID :	store_and_fwd_flag	mta_tax	tip_amount	tolls_amount	improvement_surcharge	total_amount	congestion_surcharge	airport_fee
0				2022-01-01 00:53:29	2.0	3.80	1.0			3.65	0.0	0.3	21.95		0.0
1		2022-01-01 00:33	3:43	2022-01-01 00:42:07	1.0	2.10	1.0	N N	0.5	4.00	0.0	0.3	13.30	0.0	0.0
2		2022-01-01 00:53		2022-01-01 01:02:19	1.0	0.97	1.0		0.5	1.76	0.0	0.3	10.56	0.0	0.0 0.0
3		2022-01-01 00:25		2022-01-01 00:35:23	1.0	1.09	1.0		0.5	0.00	0.0	0.3	11.80		0.0
4		2022-01-01 00:36	5:48	2022-01-01 01:14:20	1.0	4.30	1.0	N N	0.5	3.00	0.0	0.3	30.30		0.0
				2022 04 24 22 42 54					: : :	2.39	0.0	0.3	43.60		
2463926		2022-01-31 23:36		2022-01-31 23:42:51	NaN		NaN	None	0.5				13.69	NaN	NaN
2463927		2022-01-31 23:44		2022-01-31 23:55:01	NaN	4.19	NaN	None	0.5	4.35	0.0	0.3	24.45	NaN	NaN
2463928		2022-01-31 23:39		2022-01-31 23:50:00	NaN	2.10	NaN	None	0.5	2.00	0.0	0.3	16.52	NaN	NaN
2463929		2022-01-31 23:36	5:42	2022-01-31 23:48:45	NaN	2.92	NaN	None	0.5	0.00	0.0	0.3	15.70	NaN	NaN
2463930		2022-01-31 23:46	5:00	2022-02-01 00:13:00	NaN	8.94	NaN	None	0.5		0.0	0.3	35.06	NaN	NaN
50															
[2463931	rows x 19	ocolumns]													
												·			

yellow_tripdata_2022-01.parquet



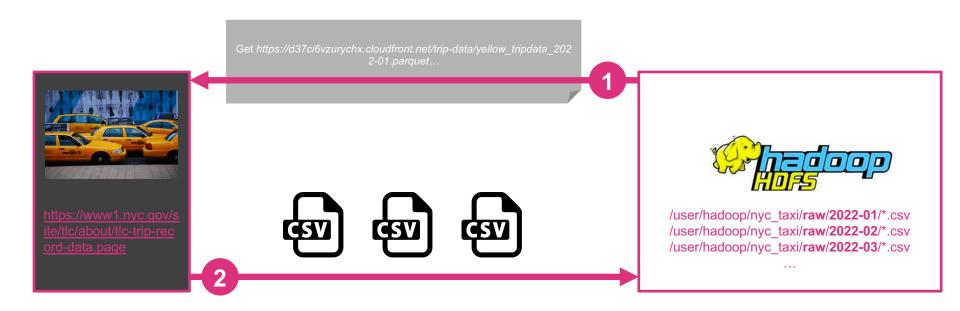
We want to make use of this data to calculate some KPIs

Workflow:

- Gather data from https://www1.nyc.gov/site/tlc/about/tlc-trip-record-data.page
- Save raw data (CSV files) to HDFS (partitioned by YYYY-MM)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Calculate KPIs and Export them to an Excel File
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically



Dataflow: 1. Get TLC NYC Taxi Data



Dataflow: 2. Raw To Final Transfer

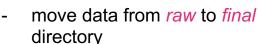


/user/hadoop/nyc_taxi/**raw/2022-01/***.csv /user/hadoop/nyc_taxi/**raw/2022-02/***.csv /user/hadoop/nyc_taxi/**raw/2022-03/***.csv









- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/nyc_taxi/final/2022-01/*
/user/hadoop/nyc_taxi/final/2022-02/*
/user/hadoop/nyc_taxi/final/2022-03/*

...



Dataflow: 3. Calculate And Export KPIs





- calculate KPIs and export them to Excel
- use Hive, Spark or PySpark



Dataflow: 4. KPIs To Calculate

Calculate per Month:

- Average Trip Duration (in minutes)
- Average Trip Distance (in miles)
- Average total amount (in USD)
- Average tip amount (in USD)
- Average passenger count (as Number)
- Usage Share by payment type (credit card, cash... in percent)
- Usage share per timeslot (in percent):
 - 00:00-06:00
 - 06:00-12:00
 - 12:00-18:00
 - 18:00-24:00





kaggle.com provides monthly exports of Hubway bike sharing trip records:

- https://www.bluebikes.com/
- Latest Full Dumps: https://www.kaggle.com/acmeyer/hubway-data

```
"tripduration", "starttime", "stoptime", "start station id", "start station name", "start station latitude", "start station longitude", "end station id", "end station na
me", "end station latitude", "end station longitude", "bikeid", "usertype", "birth year", "gender"
"133","2015-12-01 00:01:52","2015-12-01 00:04:06","9","Agganis Arena - 925 Comm Ave.","42.351246","-71.115639","41","Packard's Corner - Comm. Ave. at Brighto
n Ave.","42.352261","-71.123831","199","Customer","1995","1"
"1522","2015-12-01 00:05:30","2015-12-01 00:30:53","41","Packard's Corner - Comm. Ave. at Brighton Ave.","42.352261","-71.123831","54","Tremont St / West St","4
2.354979"."-71.063348"."876"."Customer"."1983"."1"
"153","2015-12-01 00:07:46","2015-12-01 00:10:20","75","Lafayette Square at Mass Ave / Main St / Columbia St","42.36346469304347","-71.10057324171066","67","
MIT at Mass Ave / Amherst St", "42.3581", "-71.093198", "757", "Subscriber", "1995", "1"
"435","2015-12-01 00:07:48","2015-12-01 00:15:04","68","Central Square at Mass Ave / Essex St","42.36507","-71.1031","29","Innovation Lab - 125 Western Ave. at
Batten Way", "42.363732", "-71.124565", "853", "Subscriber", "1988", "1"
"1208","2015-12-01 00:12:15","2015-12-01 00:32:23","36","Boston Public Library - 700 Boylston St.","42.349673","-71.077303","110","Harvard University Gund Hall at
Quincy St / Kirkland S","42.376369","-71.114025","437","Customer","1982","1"
"1117","2015-12-01 00:16:31","2015-12-01 00:35:09","31","Seaport Hotel","42.348833","-71.041747","67","MIT at Mass Ave / Amherst St","42.3581","-71.093198","11
61", "Subscriber", "1988", "1"
"1287","2015-12-01 00:16:50","2015-12-01 00:38:18","10","B.U. Central - 725 Comm. Ave.","42.350406","-71.108279","23","Mayor Martin J Walsh - 28 State St","42.3
5892"."-71.057629"."565"."Subscriber"."1966"."1"
```



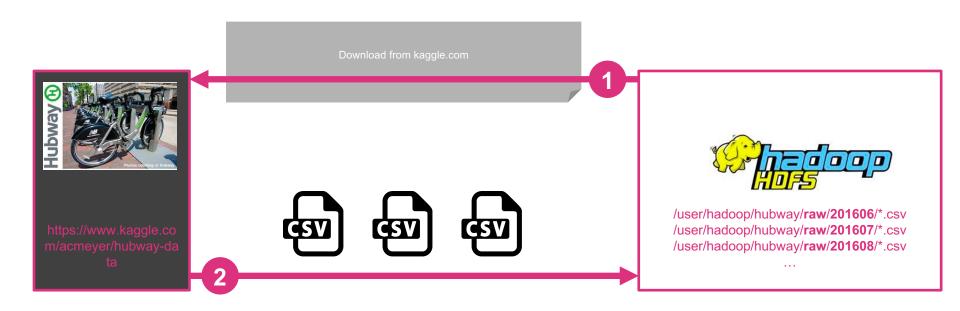
We want to make use of this data to calculate some Usage KPIs.

Workflow:

- Gather data from https://www.kaggle.com/acmeyer/hubway-data
- Save raw data (CSV files) to HDFS (partitioned by YYYYMM)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Calculate KPIs and Export them to an Excel File
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically



Dataflow: 1. Get Hubway Bike Sharing Data



Dataflow: 2. Raw To Final Transfer



/user/hadoop/hubway/**raw/201606/***.csv /user/hadoop/hubway/**raw/201607/***.csv /user/hadoop/hubway/**raw/201608/***.csv



- move data from raw to final directory
- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/hubway/final/201606/* /user/hadoop/hubway/final/201607/* /user/hadoop/hubway/final/201608/*

. . .



Dataflow: 3. Calculate And Export KPIs



/user/hadoop/hubway/final/201606/*
/user/hadoop/hubway/final/201607/*
/user/hadoop/hubway/final/201608/*

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- calculate KPIs and export them to Excel
- use Hive, Spark or PySpark



Dataflow: 4. KPIs To Calculate

Calculate per Month:

- Average Trip Duration (in minutes)
- Average Trip Distance (in km)
- Usage Share by gender (in percent)
- Usage Share by age (in percent)
- Top 10 most used bikes
- Top 10 most start stations
- Top 10 most end stations
- Usage share per timeslot (in percent):
 - 00:00-06:00
 - 06:00-12:00
 - 12:00-18:00
 - 18:00-24:00





OpenCellID.com provides regulary exports of worldwide cell data:

- https://www.opencellid.org
- Latest Full Dump and Diffs: https://www.opencellid.org/downloads.php

Full Database

cell_towers.csv.gz
 Updated: 2021-03-21 (982MB)

Differential

- OCID-diff-cell-export-2021-03-21-T000000.csv.gz (829KB)
- OCID-diff-cell-export-2021-03-20-T000000.csv.gz (1526KB)
- OCID-diff-cell-export-2021-03-19-T000000.csv.gz (1534KB)
- OCID-diff-cell-export-2021-03-18-T000000.csv.gz (1550KB)
- OCID-diff-cell-export-2021-03-17-T000000.csv.gz (1540KB)
- OCID-diff-cell-export-2021-03-16-T000000.csv.gz (1434KB)
- OCID-diff-cell-export-2021-03-15-T000000.csv.gz (545KB)

radio,mcc,net,area,cell,unit,lon,lat,range,samples,changeable,created,updated,averageSignal
UMTS,262,2,801,86355,0,13.285512,52.522202,1000,7,1,1282569574,1300155341,0
GSM,262,2,801,1795,0,13.276907,52.525714,5716,9,1,1282569574,1300155341,0
GSM,262,2,801,1794,0,13.285064,52.524,6280,13,1,1282569574,1300796207,0
UMTS,262,2,801,211250,0,13.285446,52.521744,1000,3,1,1282569574,1299466955,0
UMTS,262,2,801,86353,0,13.293457,52.521515,1000,2,1,1282569574,1291380444,0
UMTS,262,2,801,86357,0,13.289106,52.53273,2400,3,1,1282569574,1298860769,0
UMTS,262,3,1107,83603,0,13.349675,52.497575,3102,222,1,1282672189,1300710809,0
GSM,262,2,776,867,0,13.349711,52.497367,1000,214,1,1282672189,1300710809,0
GSM,262,3,1107,13971,0,13.349743,52.497437,1000,212,1,1282672189,1300710809,0
UMTS,262,3,1107,355,0,13.34963,52.497378,1000,198,1,1282672189,1300710809,0
UMTS,262,3,1107,329299,0,13.349223,52.497519,3041,186,1,1282672189,1299860879,0

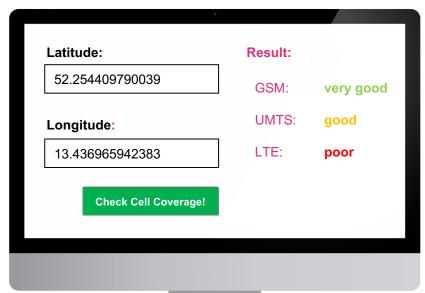
cell_towers.csv



We want to make use of this data to estimate the coverage of GSM, UMTS and LTE for a certain place (latitude, longitude).

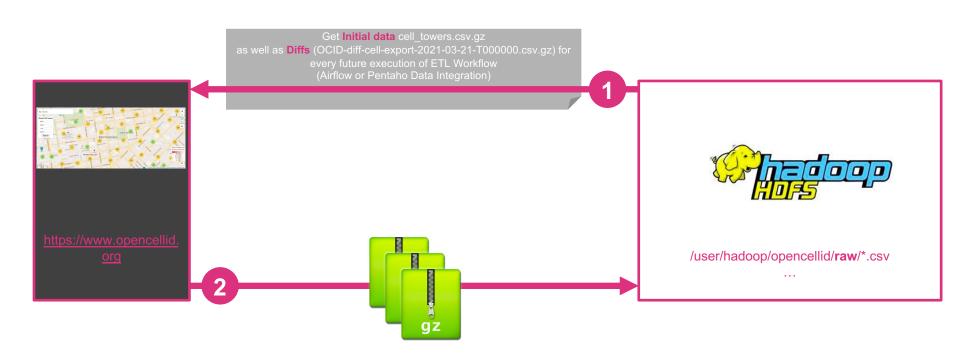
Workflow:

- Gather data from OpenCellID.com
- Save raw data (CSV files) to HDFS (partitioned by radio, e.g. GSM, UMTS, LTE...)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Export address data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple **HTML Frontend** which is able to:
 - read from end-user database
 - process user input (Latitude, Longitude...)
 - checks against OpenCellID data in enduser database
 - Display result (GSM, LTE and UMTS coverage)
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically





Dataflow: 1. Get Cell Data



Dataflow: 2. Raw To Final Transfer



/user/hadoop/opencellid/**raw**/*.csv





- move data from raw to final directory
- merge full dump and diffs
- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/opencellid/**final**/*



Dataflow: 3. Enhance Data And Save Results





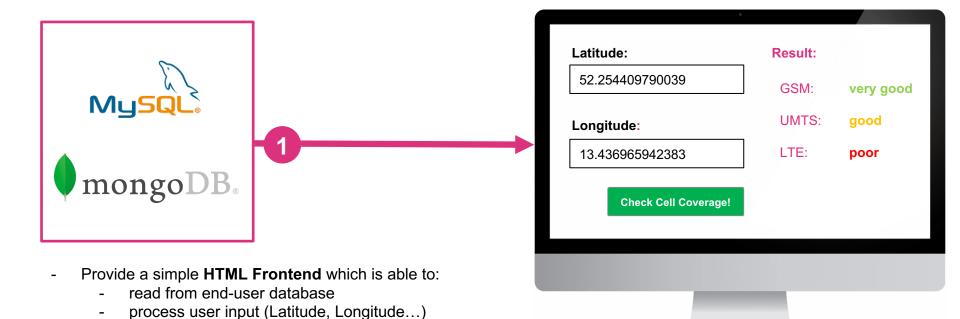
- enhance data (e.g. for later querying)
- use Hive, Spark or PySpark
- save everything to a enduser database (e.g. MySQL, MongoDB)







Dataflow: 4. Provide Simple Web Interface





user database

checks against OpenCellID data in end-

Display result (GSM, LTE and UMTS coverage)



OpenAddresses.io provides regulary exports of worldwide adresses (we will focus on US south/west/midwest/northeast for now):

https://batch.openaddresses.io/data

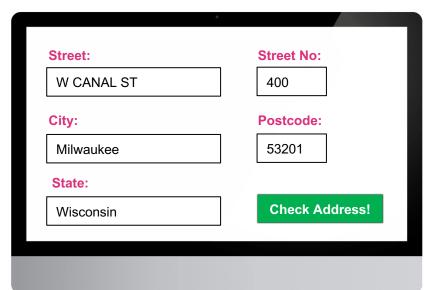


```
{"type": "Feature", "properties": { "hash": "394d6a8e3e6cecbf", "number": "7705", "stree
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tcode":"53219","id":""},"qeometry":{"type":"Point","coordinates":[-88.0088621,43
.00258451}}
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ode":"53226","id":""},"qeometry":{"type":"Point","coordinates":[-88.0551894,43.0
{"type":"Feature", "properties": { "hash": "fbf0248cdd1623ad", "number": "12137", "stre
et":"W BURLEIGH ST", "unit":"2", "city": "Wauwatosa", "district":"", "region":"", "pos
tcode":"53222","id":""},"geometry":{"type":"Point","coordinates":[-88.0649444,43
{"type": "Feature", "properties": { "hash": "6c5867d0d98b7e9a", "number": "11515", "stre
et":"W CLEVELAND AVE","unit":"B231","city":"West Allis","district":"","region":"
","postcode":"53227","id":""},"geometry":{"type":"Point","coordinates":[-88.0560
418,42.99465291}}
[\ldots]
```

We want to make use of this data to validate adresses entered on a website, to check whether they are real or not.

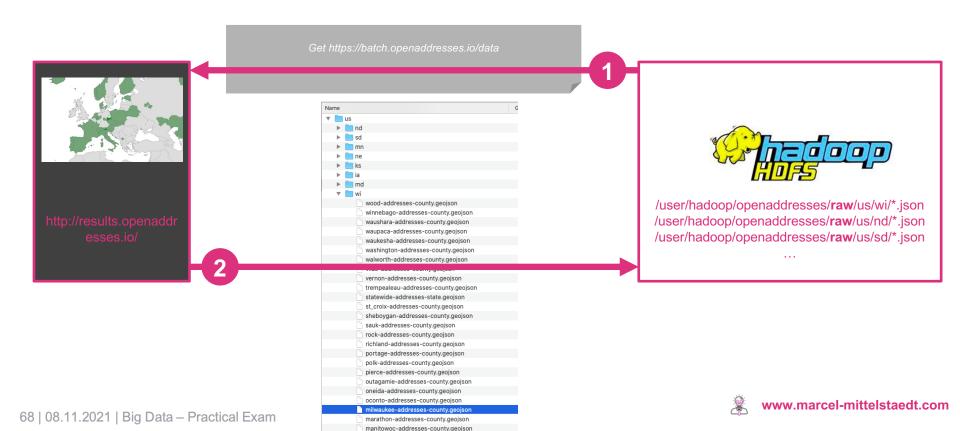
Workflow:

- Gather data from OpenAddresses.io
- Save raw data (JSON files) to HDFS (partitioned by state
- shortcut, e.g. wi, nd, sd...)
- Optimize, reduce and clean raw data and save it to final directory on HDFS
- Export address data to end-user database (e.g. MySQL, MongoDB...)
- Provide a simple **HTML Frontend** which is able to:
 - read from end-user database
 - process user input (Street, City, Postcode...)
 - validate user input against OpenAddress data in enduser database
 - Display result (real or non real address)
- The whole data workflow must be implemented within an ETL workflow tool (e.g. Pentaho Data Integration or Airflow) and run automatically





Dataflow: 1. Get Address Data



Dataflow: 2. Raw To Final Transfer



/user/hadoop/openaddresses/**raw**/us/wi/*.json /user/hadoop/openaddresses/**raw**/us/nd/*.json /user/hadoop/openaddresses/**raw**/us/sd/*.json







- move data from raw to final directory
- Convert/Explode data structure
- optimize and reduce data structure for later query purposes if necessary
- remove duplicates if necessary
- ...



/user/hadoop/openaddresses/final/us/wi/*.parquet /user/hadoop/openaddresses/final/us/nd/*.parquet /user/hadoop/openaddresses/final/us/sd/*.parquet

. . .



Dataflow: 3. Enhance Data And Save Results



/user/hadoop/openaddresses/final/us/wi/*.parquet /user/hadoop/openaddresses/final/us/nd/*.parquet /user/hadoop/openaddresses/final/us/sd/*.parquet









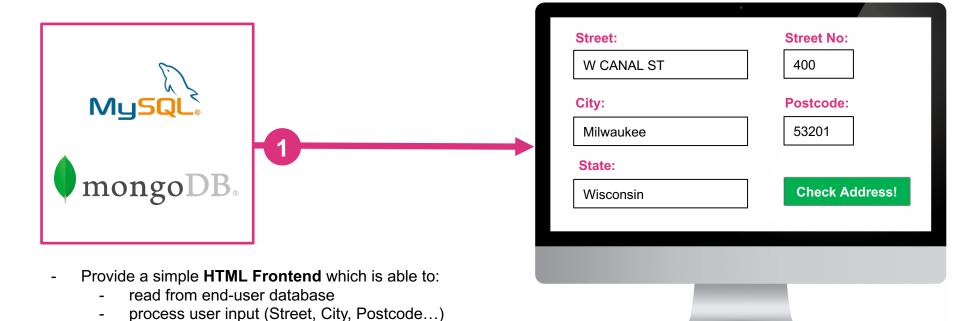
- enhance data (e.g. add missing entries of street no's)
- use Hive, Spark or PySpark
- save everything to a enduser database (e.g. MySQL, MongoDB)







Dataflow: 4. Provide Simple Web Interface



user database

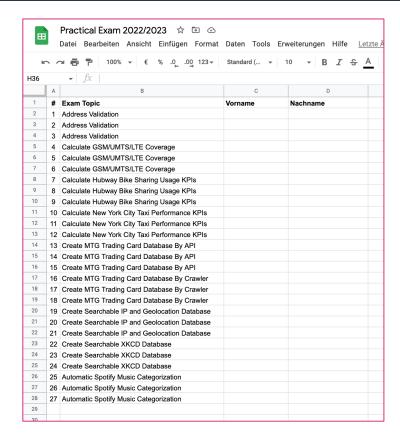
validate user input against OpenAddress data in end-

Display result (real or non real address)

Good Luck and have fun!



Please choose an exam topic!



tinyurl.com/2ft4wmf8





Stop Your VM Instances

gcloud compute instances stop big-data

