TROIA

Efficient electricity distribution and sustainable energy management through Big Data analytics and machine learning

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ENERDAY 2023 5. 5. 2023



AGENDA

- INTRODUCTION
- BIG DATA/ANALYTICS PLATFORM
- DATA VISUALIZATION USE CASES
- CHALLENGES IN IMPLEMENTING SOLUTION
- AI/ML
- DISQUSSION & QUESTIONS



- Company founded 2010
- The core team has 20+ years of experience in the energy industry
- o Our customers in Energy sector: distribution and transmission companies





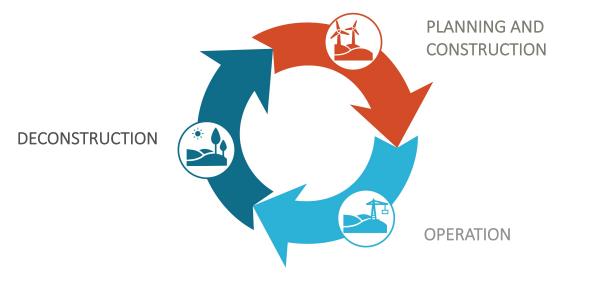
DACH-Partner: RODIAS

For more than 35 years, we have been **supporting utilities in the digitalisation** of their processes. In the **nuclear industry**, RODIAS has been setting the standard for **plant management systems** in the Germanspeaking market for several decades.

Throughout the last decade, we **entered** other industries like **mobility&transport**, **pulp&paper**, **facility management** and **automotives**.

For the **future**, we are positioning ourselves in the **hydrogen market**.

Our future-oriented solutions enrich Enterprise Asset Management systems with big data, analytics and machine learning use cases.









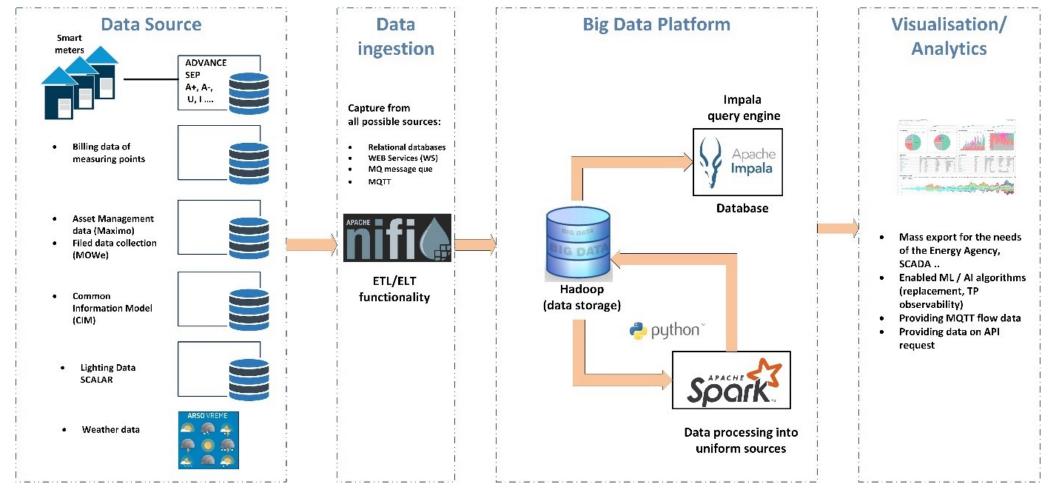
IMPLEMENTING BIG DATA SOLUTION/PLATFORM

- Digitalisation:
 - Data-driven organisations
 - Generating huge amounts of data (IoT devices/smart meters)
 - Siloed data
- The volume and quantity of data makes it impossible to manage data in the traditional way (storage/processing)
 - Implementing big data solutions
- Network management, optimisation and development

Accurate and timely data and information is the cornerstone of optimising energy use and provides many untapped opportunities to reduce energy use, thereby lowering our carbon footprint and making our operations even more sustainable.



BIG DATA/ANALYTICS PLATFORM – ATHENA ANALYTICS





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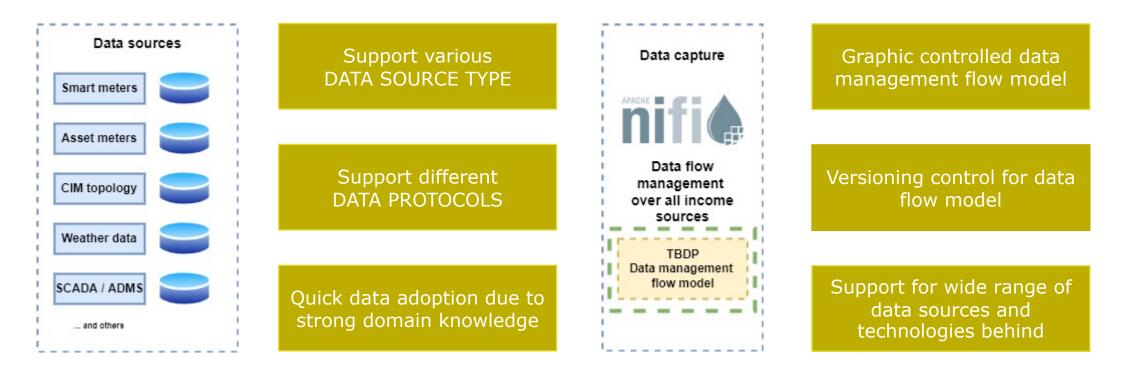
- **Open Source** Solutions (Cloudera)
- Central data platform
- It provides:
 - Security
 - Robustness
 - High mass data processing capability
 - Scalability:
 - Data volume
 - Scope of data sources
- Business Analytics (Data Analytics, Visualisations, Dashboards, Reports)
 - Power BI, Tableau, Grafana, Superset, QGis, Neo4j
 - Ad Hoc Queries (Hue)
- Advanced Analytics (AI/ML/Data mining)
- Data services
 - on-demand access to data by external stakeholders (REST API)

Capture data from different sources:

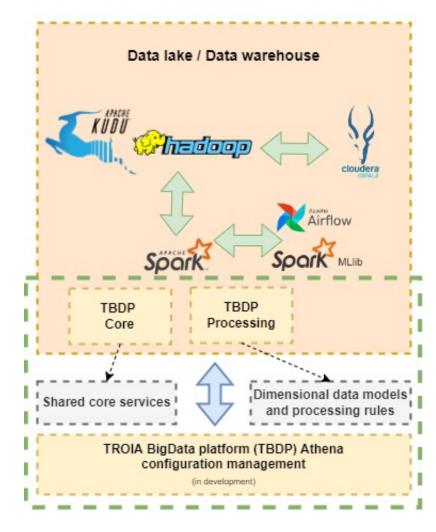
- Flow data (sensors, iOT, smart meters)
- Relational databases
- Online sources
- Data processing/preparation/ETL/ELT
 - Cleaning and editing
 - Standardisation, segmentation, aggregation
 - Deduplication, imputation
 - Transformation, substitution, conversion
- Implementation options:
 - On premise/private cloud
 - Public Cloud/Hybrid Cloud



BIG DATA/ANALYTICS PLATFORM – ATHENA ANALYTICS



CLOUDERA



ATHENA ANALYTICS Core

- Shared utility functionalities
- Core data libraries
- Supports many SCD types
- Resilient by design
- Highly configurable

ATHENA ANALYTICS Processing

- Business logic
- Dimensional data models (Kimball DWH approach)
- Processing rules
- Custom environmental variables

ATHENA ANALYTICS configurations (in development)

- Intuitive web interfaces for configuration management
- Graphical view of data model relations
- Other platform configurations
- Deployments

TROL Data visualization / Data export services Enerday 2023 5.5.202

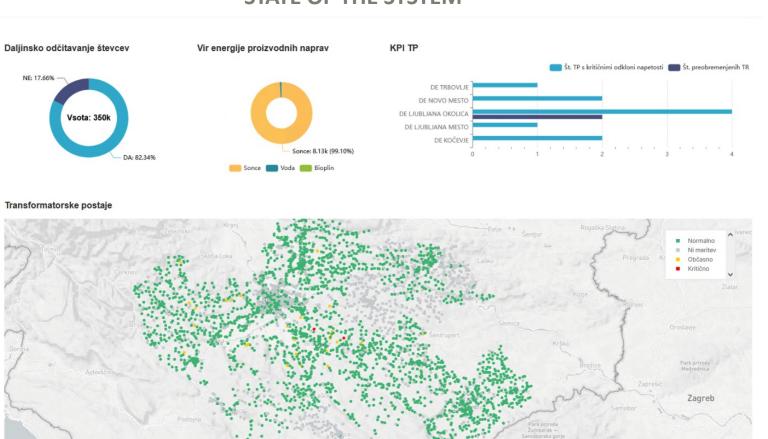








© Mapbox © OpenStreetMap Improve this map



STATE OF THE SYSTEM

mapbox



TRANSFORMER STATION LOAD

TP Naziv TP Tip						
	TP S_projekt. (kVA)	Leto izgradnje	SN-izvod 😄	RTP/RP 🗇	Nadzorništvo 🝵	Seznam MM 😄
2000608 VRH PRI LESKOVCU 20/0.4 G-389 JAMBORSH	KA ŽELEZNA 50	1984	J28 DV 20KV POLICA	RTP 110/20 KV GROSUPLJE	GROSUPLJE 1	Odpri

Specifikacije transformatorjev (podatki o nameščenem sredstvu iz MX)											(1) :
Šifra tr. 😄	Tip tr. 🗇	U₁/U₂ (kV) ≑	S _n (kVA) =	I₂a (A) ≎	u _k (%) 😄	Izg. krat. stika (W) 🔅	Izg. praz. teka (W) 😄	Hladilni medij 🌣	Leto izdelave 🔅	Obratovalno MM 🗇	ID števca 🔅
3784148	TR. 4HT 50/20-0.4	20/0,4	50	80.83	4.19	1075	188	OLJE-S	1984	8038950	70630275

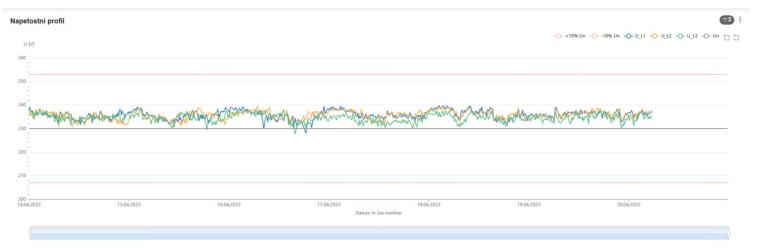
Obremenitev TP / transformatorja



Q (p.u.)

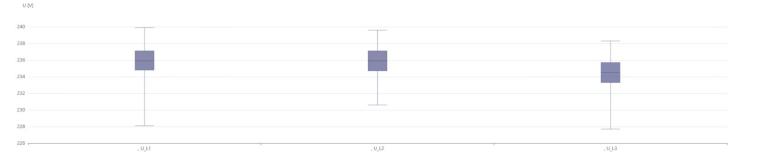


MEASURING SITE VOLTAGE TIME SERIES









Statistični podatki za izbrano časovno obdobje									₹2 :
U_L1 max 🗘	U_L2 max ¢	U_L3 max 👳	U_L1 min =	U_L2 min 🕆	U_L3 min 🗧	U_L1 povp. 🗘	U_L2 povp.	U_L3 povp. 🗇	
239.9	239.6	238.3	228.1	230.6	227.7	235.9	235.8	234.5	



MEASURING SITE LOAD AND POWER CONSUMPTION

TP 🗇	MM 🗇	Vrsta MM	Napetostni nivo 🗇	Vrsta odjema 🖨	Vezalna shema 🗉	Tovarniška št. 👙	Proizvajalec 👙	Tip 😄	Daljinsko odčitavan
TP0454-ŠOLA JARŠE	3102790	OBRAČUNSKO	Nizka napetost	Gospodinjstvo	PS.1A	77260743	ISKRAEMECO	AM550-TD1 PLC	DA
stični podatki za izbra									

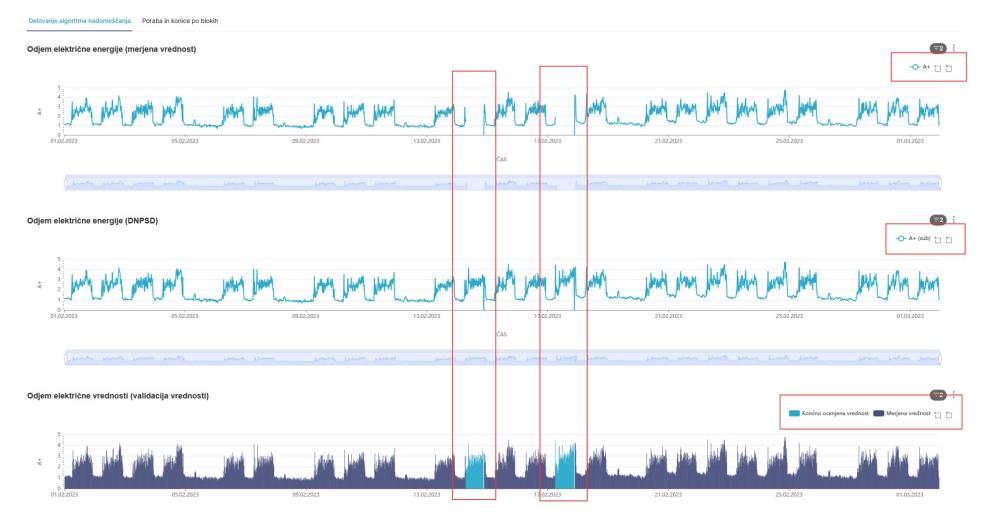
P+_max [kW] =	P+_povp. [kW] 😄	A+ [kWh] 🗘	Pmax [kW] =	Ppovp. [kW] 🕆	A- [kWh] ≑
8.7	1.1	165.6	0	0.0	0.0

Dnevna stanja naprave Prejeta in oddana električna energija Analiza delovne in jalove moči Dinamičen tarifni sistem





IMPUTATION OF MISSING METER DATA





Datum in čas meritve

Dynamic tariff system – change of electrial billing

TP 💠	MM ÷	Vrsta MM 👙	Napetostni nivo 👙	Vrsta odjema 👙	Vezalna shema 👙	Tovarniška št. 🍵	Proizvajalec 👙	Tip 💠	Daljinsko odč	itavanje 👙
P0454-ŠOLA JARŠE	3102790	OBRAČUNSKO	Nizka napetost	Gospodinjstvo	PS.1A	77260743	ISKRAEMECO	AM550-TD1 PLC	DA	
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[kW] =		P+_povp. [kW] 😄		A+ [kWh] ≑	Pmax [kW] ⇔		Ppovp. [kW] =		A- [kWh] ≎	
		1.1		165.6	0		0.0		0.0	
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t profila odjema gle	de na dinamič	en tarifni sistem								
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	ed	63.5		44.5		34.8		26.1	•4 •	3 •2 •5
	ed	63.5		44.5		34.8		26.1	•4 •	3 •2 •5
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Grouped OStacker				44.5	Apr 2023	34.8		26.1		2:
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djema energije glec				44.5	Apr 2023	34.8		26.1		1
djema energije glec				44.5	Apr 2023	34.8		26.1		₹2

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3	3	4	4	5
4	3	4	4	5
5	3	4	4	5
6	2	3	3	4
7	1	2	2	3
8	1	2	2	3
9	1	2	2	3
10	1	2	2	3
11	1	2	2	3
12	1	2	2	3
13	1	2	2	3
14	2	3	3	4
15	2	3	3	4
16	1	2	2	3
17	1	2	2	3
18	1	2	2	3
19	1	2	2	3
20	2	3	3	4
21	2	3	3	4
22	3	4	4	5
23	3	4	4	5



CHALLENGES

- Data quality
- Data governance/security
 - SSO (AD/Kerberos integration)
 - Apache Ranger (security, data access)
 - Apache Atlas (Data catalogue, lineage)
 - IBM QRADAR (SIEM, Auditing)
- Advanced Analytics Department
- In Future:
 - Data processing in (near) real-time (nRT)
 - Streaming analytics and real-time intelligence
 - Kafka Streaming, Spark Streaming, Apache Flink



Data quality management

- Data needs to be properly verified before being used for BI/data-driven decision-making
- Ensuring data quality has proved to be a major challenge
- Unreliable PLC communication
- How do we technically ensure data quality:
 - Manual verification (domain knowledge required)
 - Finding outliers
 - Replacing missing data imputation
 - Cleaning duplicate data deduplication
 - Validation of data type
 - Some data quality issues have been identified in data visualisation
- In progress: Automatic Data Quality monitoring/Data Quality KPIs

Data quality and trust in data are the foundation of a data-driven organisation!

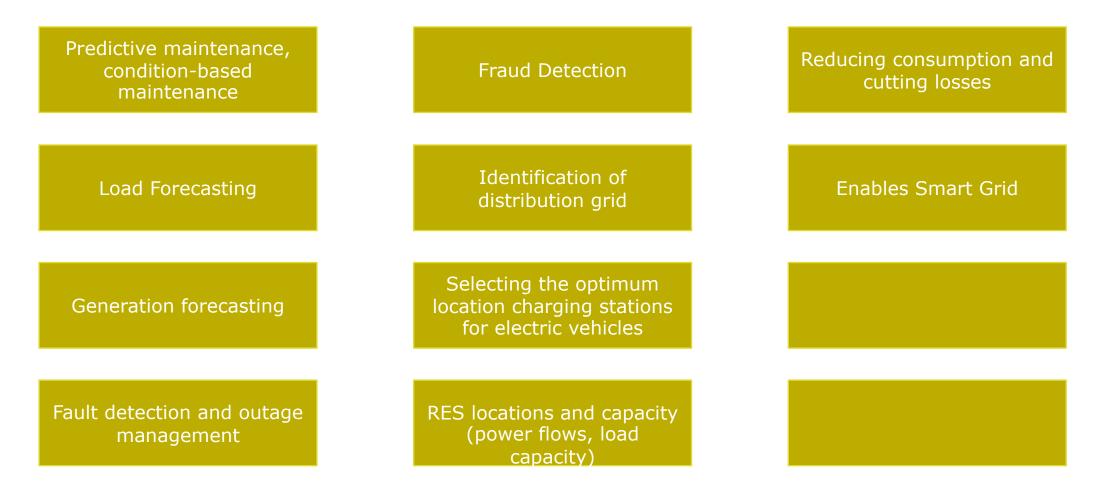
Trajnostni razvoj in podatkovna analitika 5. 5. 2023

AI/Machine learning

- Spark Mllib
 - Java, Scala, Python, R
- Dedicated compute cluster
- ML Algorithms: classification, regression, clustering, collaborative filtering ...
- Featurization: feature extraction, transformation, dimensionality reduction, and selection
- Pipelines: tools for constructing, evaluating, and tuning ML Pipelines
- Persistence: saving and load algorithms, models, and Pipelines
- MLOps
 - integrating machine learning models into the software delivery lifecycle
 - end-to-end development, deployment, and operation of machine learning models



AI/Machine learning use cases





Trajnostni razvoj in podatkovna analitika 13. 10. 2022





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WEARE

OPTIMIZING THE

OPTIMIZING THE FUTURE FOR

