

EU-CIP Project type B Resilience Intelligence Tool

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<u>B</u>usiness <u>I</u>ntelligence (BI) for <u>C</u>ritical Infrastructure (CI) <u>R</u>esilience <u>A</u>nalysis

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General introduction



<u>B</u>usiness <u>I</u>ntelligence (BI) for <u>C</u>ritical Infrastructure (CI) <u>R</u>esilience <u>A</u>nalysis

(BICRA)

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Solution Description

- In order to maintain and enhance a high level of resilience of modern critical infrastructures in Europe, it is necessary to manage and keep transparent huge amounts of varied data (measurement, monitoring, management systems, etc.) and maintain information interconnection integrity.
- A tool was developed with a goal of integrating data and information and providing transparent and smart decision making. The solution is a Business Intelligence (BI) tool, using Artificial Intelligence (AI) focused on the resilience of Critical Infrastructure (CI).





Solution Innovation

The main problems addressed in the area of CIP/CIR are the three best identified ones in the industry so far:

- a) "data/information flood" solved by integrating intelligently large amount of different types/sources of data information also in combination with data-lake/data-mart and digital twin technologies, on the leading side, and the legacy systems, on the lagging side;
- b) "lack of transparency" solved by providing interactive, BI-based intuitive interfaces;
- c) "missing the early warnings and signals" solved by smart focusing on most important subsets of resilience data/information.





- We use the general Microsoft solutions (Microsoft Power BI and the Microsoft SQL server) as the development basis, but the final solution is fully customized to the end-user needs in terms of data and information sources addressed and used (documents, measurement records, analysis results...) from a whole spectrum of tools.
- The technology allows to perform the what-if analysis interactively and with practically no IT-training, all the knowledge prerequisites are related to the domain knowledge (e.g. risk assessment of the CI assets).

Sector Relevance

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Power generating facilities

Raw material processing plants

Renewable energy stations





How do we assess resilience in BICRA

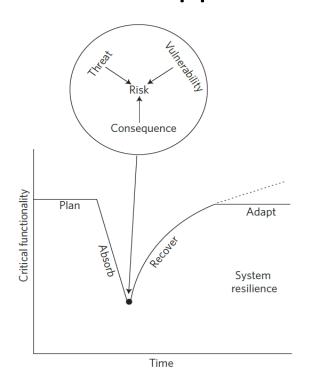


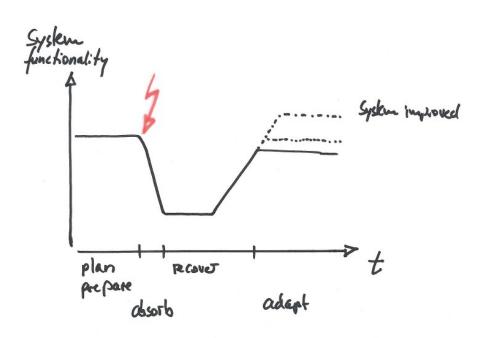
<u>Business Intelligence</u> (BI) for <u>Critical Infrastructure</u> (CI) <u>Resilience Analysis</u>

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Resilience: what happens when risk happens?



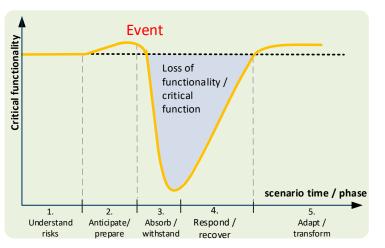




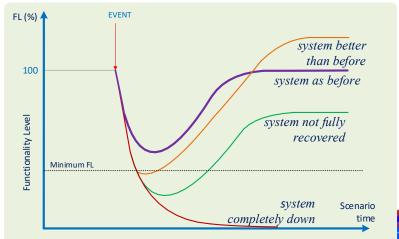
Resilience:

- understand and anticipate risks

 including new/emerging risks –
 threatening the critical
 functionality of the
 infrastructure,
- prepare for anticipated or unexpected disruptive events, optimally
- 3. absorb/withstand their impacts,
- **4. respond and recover** from them, and
- 5. adapt/transform the infrastructure or its operation based on lessons learned, thus reducing the critical infrastructure fragility.







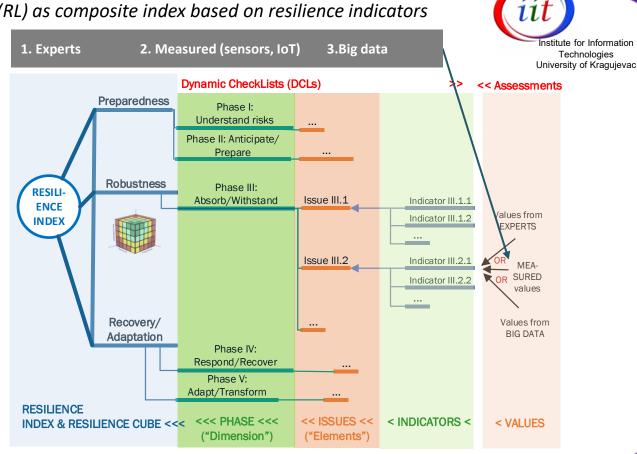
Practical ways of assessing resilience – e.g. indicators

Resilience Level (RL) as composite index based on resilience indicators

Hierarchical organisation of indicators in a dynamic checklist (DCL)

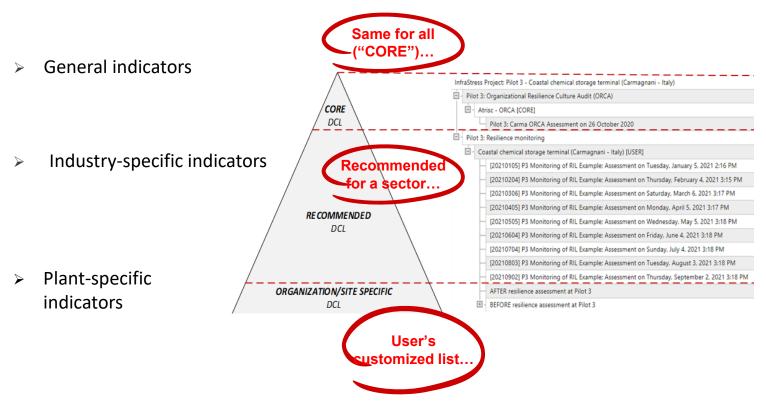
- Bottom-up process (right to left!)
- Indicator values aggregated to a single summarized index value:

Resilience Level: 0-5



How does a dynamic checklist of indicators look like?











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(BICRA)



Mostly used in BICRA

1. FCM:

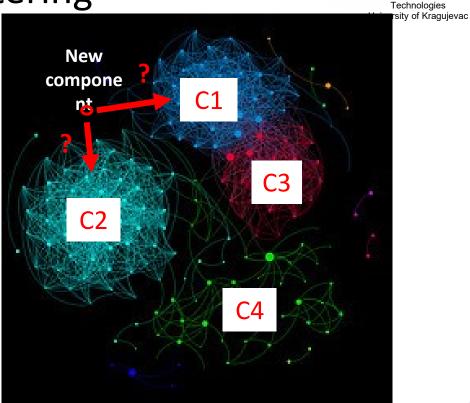
Fuzzy C-mean clustering - Continuous and dynamic data mining over the data already acquired and results already achieved

2. NN/ML (neural networks & machine learning): Prediction based on neural networks

FCM:

Fuzzy C-mean clustering

- Practically it means that the components are dynamically clustered (e.g. "all superheater outlet headers") according to their characteristics and associated risks
- Based on their characteristics, the new, yet to be analysed components (e.g. a new superheater), can than be preassigned to the corresponding clusters and a preliminary, "automatic" risk assessment made (e.g. "medium risk")



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Application of FCM

- 1134 valve data from projects
- 7 characteristics of valves

| Cluster Centers | | | | | | | | | | |
|--------------------|-----------------------|---------------------|-------|-----------------------|--------------------------|-------|--|--|--|--|
| Design Pressure | Design Temperature | Nominal Diameter | Cr% | Operating Pressure | Operating Temperature | Risk | | | | |
| 25850,401 | 312,616 | 89,979 | 0,749 | 18819,224 | 274,327 | 0,076 | | | | |
| 11377,388 | 516,103 | 101,974 | 1,477 | 9563,869 | 498,621 | 0,125 | | | | |
| 13862,614 | 267,932 | 96,588 | 0,878 | 10368,124 | 235,766 | 0,055 | | | | |
| 5732,171 | 349,437 | 85,160 | 0,415 | 3058,279 | 296,933 | 0,038 | | | | |



Nominal

Diameter

89,979

101,974

96,588

85,160

Nominal

Diameter

70

300

Clusters

Design

Temperature

312,616

516,103

267,932

349,437

Examples

Design Temperature

300

400

Design

Pressure

25850,401

11377,388

13862,614

5732,171

Design

Pressure

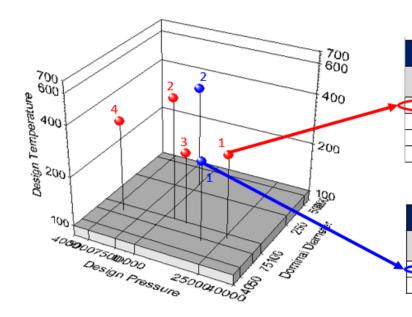
20000

10000

Clustering (Fuzzy C-Means)

- Possible valve examples are positioned.
 - 4 clusters

 (regarding
 Design
 Pressure,
 Design
 Temperature
 and Nominal
 Diameter
 characteristics)
 - 2 examples



Examples

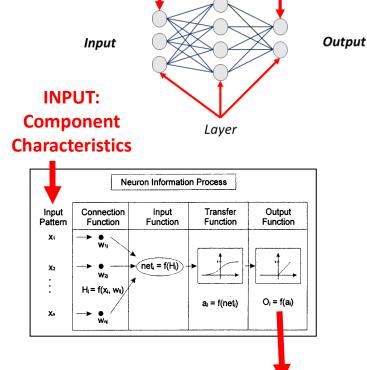
Clusters

NN:

Prediction based on neural networks connection

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- The of NN is imitating the structure and workings of the human brain through mathematical models. Three basic qualities of the human brain are taken as foundations for NN models:
 - knowledge (distributed over many neurons within the brain)
 - neurons (can communicate locally with one another)
 - brain (can adapt)
- There are basically three types of NN models:
 - Multilayer perceptron



OUTPUT: Component Risk



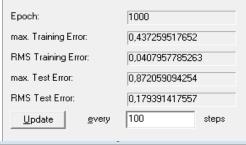
Neural Network (Multilayer Perceptron)

- for training process 168 data
- for testing process 8 data
- as <u>input</u> parameters 7 characteristics of valves
- as output parameter "risk" characteristic are used

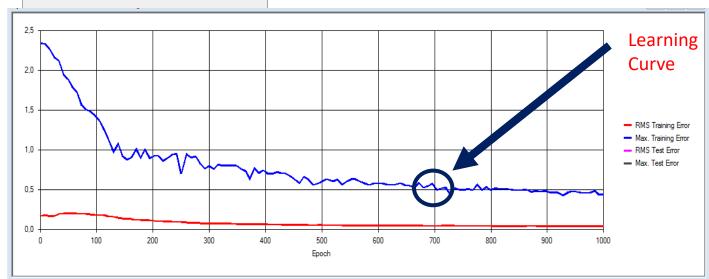
| | DesignPressure [] | DesignTemperature [] | NominalDiameter | Cr% | OperatingPressure [] | OperatingTemperature [] | B_Risk | * |
|----|----------------------|-------------------------|-----------------|-------|-------------------------|----------------------------|--------|---|
| 1 | 25.000,000 | 280,000 | 300,000 | 0,300 | 19.250,000 | 250,000 | 0,000 | |
| 2 | 33.130,000 | 280,000 | 300,000 | 0,300 | 19.250,000 | 250,000 | 0,000 | |
| 3 | 20.500,000 | 367,000 | 32,000 | 0,300 | 19.250,000 | 250,000 | 0,000 | |
| 4 | 33.130,000 | 367,000 | 150,000 | 0,300 | 18.500,000 | 280,000 | 0,000 | |
| 5 | 20.070,000 | 405,000 | 65,000 | 0,300 | 18.500,000 | 250,000 | 0,000 | |
| 6 | 20.070,000 | 405,000 | 80,000 | 0,300 | 18.500,000 | 250,000 | 0,000 | |
| 7 | 20.070,000 | 405,000 | 65,000 | 0,300 | 18.500,000 | 250,000 | 0,000 | |
| 8 | 20.070,000 | 405,000 | 80,000 | 0,300 | 18.500,000 | 250,000 | 0,000 | |
| 9 | 20.070,000 | 405,000 | 80,000 | 0,300 | 18.500,000 | 250,000 | 0,000 | |
| 10 | 18.700,000 | 545,000 | 250,000 | 2,250 | 17.500,000 | 530,000 | 0,000 | |



Neural Network (Perceptron)



Prediction of risk by NN





BI (Business Intelligence) in BICRA



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Why Power BI?

- Fast development and deployment: Power BI provides vast amount of built-in visualization and functionality, eliminating the need to develop these from scratch. This significantly speed up deployment time.
- Data connectivity: Power BI supports connections to numerous data sources, as well as real-time streaming of data and monitoring.
- **Ease of use:** Allows end-users to conduct their analytics without relying heavily on IT, promoting more agile decision-making processes.



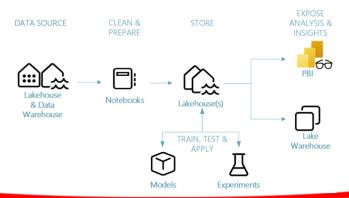
Why Power BI?

- Integration with Microsoft Ecosystem: As part of the Microsoft ecosystem, BICRA integrates seamlessly with other Microsoft products and other Power BI dashboards/reports in use by critical infrastructures
- Advanced analytics: Leverage AI capabilities and integrate machine learning models to perform advanced analytics directly within Power BI, as well as have capabilities to run R and Python scripts



Power BI integration into Microsoft Fabrics

 Recently Power BI has been integrated into Microsoft Fabrics data platform. This integration significantly expand capabilities of Power BI, allowing seamless move from data ingestion to transformation and visualization within one environment. This allows us to, beside usage of our own AI models, leverage AI-powered features available in Microsoft Fabric to enrich the insights generated through Power BI reports and dashboards.





Power BI integration into Microsoft Fabrics

- We use Power BI Desktop to create report/dashboards that can then be shared and used by end-users via Power BI Server.
- Typical workflow within Power BI Desktop includes following steps:
- 1. Connect to data, including multiple data sources.
- 2. Shape the data with queries that build insightful, compelling data models.
- 3. Use the data models to create visualizations and reports.
- 4. Share created report files for others to leverage, build upon, and share.







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Power plants (PP) and heating plants (HP) in Serbia



- PP TENT A
- PP TENT B
- HP Novi Beograd
- PP Kolubara A
- PP Kolubara B
- PP Kolubara C
- HP Novi Sad
- 7 smaller plants incl. industry (as 1 location)









<u>B</u>usiness <u>I</u>ntelligence (BI) for <u>C</u>ritical Infrastructure (CI) <u>R</u>esilience <u>A</u>nalysis (**BICRA**)

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Key features or unique selling points of the product/service



- The general BI tools are available around (e.g., the Oracle and Microsoft ones) but not the dedicated ones. We use the general Microsoft solutions (Microsoft Power BI and the Microsoft SQL server) as the development basis, but the final solution is fully customized to the end-user needs in terms of data and information sources addressed and used (documents, measurement records, analysis results...) from a whole spectrum of tools.
- The technology allows to perform the what-if analysis interactively and with practically no IT-training, all the knowledge prerequisites are related to the domain knowledge (e.g. risk assessment of the CI assets).



Future of the product

- Preliminary market search done for Serbia and the Balkans region
- Collaboration with companies in Serbia and Hungary established
- Contacts and collaboration with companies is expected to be established with help of EU-CIP project
- Pilot project costs/duration: 50-70 k€ / 6-12 months



Future of the project - Marketing

- Social Media Marketing: Establish a strong presence on relevant social media platforms; Create a content calendar and engage with the audience regularly.
- Email Marketing: Build and maintain an email list for targeted communication; Implement email campaigns for product updates, promotions, and customer engagement.
- Partnerships and Collaborations: Explore opportunities for partnerships with other businesses or organizations; Collaborate on joint promotions or events.
- Customer Feedback and Testimonials: Encourage and showcase customer reviews and testimonials to build trust.



Final conclusion

BICRA contributes to EU-CIP by

- 1. Identifying gaps: solutions for managing big data
- 2. Defining needs: power plants
- 3. Analyzing market: Serbia, Western Balkans
- Providing scalability and flexibility: Generic tool adapted for highly customized/tailored solutions
- Engaging stakeholders: In the region not represented in EU-CIP so far
- 6. Continuous improvements: Each solution is an improvement, also because the basic tool (PowerBITM) improves (now FabricTM)



Thank you for your attention! Any questions?