ELISE action Webinar Series

Data-driven methodology for electricity characterisation of districts

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European Location Interoperability Solutions for e-Government

Enabling Digital Government through Geospatial and Location Intelligence



Commission

What is ELISE?

A BIT OF HISTORY ...

0 2004

IDABC: Interoperable Delivery of European eGovernment Services

2010 ISA: Interoperability solutions for public administrations

• Actions: EULF ARE3NA

2016

ISA²: Interoperability Solutions for European Public Administrations, Businesses and Citizens

ELISE

DIGITAL: Digital Europe Programme

ELISE builds upon the outcomes of the former ISA actions EULF and ARE3NA. It is the only action of the ISA² Programme, aiming to improve Digital Government through Location Interoperability.

WHAT?

ELISE stands for European Location Interoperability Solutions for e-Government. It is one of the more than 50 actions in the European Interoperability Programme ISA2

•••• WHAT FOR?

To support Digital Government Transformation by making the best use of location data and technologies in an interoperable manner

✓ FOR WHOM?

For all: citizens, businesses and public administrations

Location-enabled Digital Government Transformation









5 Years

SOME ACHIEVEMENTS

- Complemented the EIF and NIFO with an extensive location interoperability framework and state of play assessments
- Helped put the INSPIRE Directive into practice with tools for data providers and a strong focus on use cases
- Built an extensive community of European and international stakeholders
- Raised awareness on new approaches to location-enabled digital transformation

- Helped to assess the role of SDIs in evolving business models, e.g. data ecosystems, digital platforms.
- Assessed new policies (e.g. GDPR, European Data Strategy) and technologies (e.g. Artificial Intelligence, Blockchain, API...)
- Promoted and facilitated better links on location data between public and private actors
- Provided guidance on improving spatial awareness and analytical skills for best use of data

3 EIF Toolbox solutions

Active engagement of

ISA² Member States





Our speakers









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The views expressed are purely those of the authors and may not in any circumstances be regarded as stating an official position of the European Commission.



What we will cover today

1. Introduction to ELISE Energy & Location Applications

2. State of the art data-driven methodologies and technologies for electricity characterisation of districts

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3. How these methodologies and technologies may support the business of start-up companies?

4. How these methodologies and technologies may support urban planning and energy efficiency policies of a Regional Government?

5. Urban energy simulation supporting the Local Digital Twins toolbox

6. Key messages, challenges and future outlook

Q&A

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Introduction to ELISE Energy & Location Applications





ELISE has developed cross-border pilots and applications to test location data interoperability principles in the following sectors:

APPLICATIONS

MARINE

Supporting Member States in the management of Marine Strategy Framework Directive (MSFD) related spatial information



TRANSPORT

Developing and sharing best practices for the implementation of the Intelligent Transport Systems (ITS) Directive



ENERGY EFFICIENCY

Supporting public administrations, businesses and citizens engaged in energy policies' cycle



https://www.youtube.com/ watch?v=Ftgy8uU9y2A

CULTURAL HERITAGE





Location data interoperability principles and methodologies applied in the Energy Efficiency sector (1/3)



- To leverage location data at building level as an enabling factor to scale-up a set of methodologies to assess energy efficiency from local to district/city level and beyond.
- To use location-based data to support different types of stakeholders engaged in energy efficiency policies' cycle



Location data interoperability principles and methodologies applied in the Energy Efficiency sector (2/3)



- Buildings are responsible for the 40% of final energy consumption
- Over 75% of building stock is older than 25 years
- Averaged final energy consumption data: 185 kWh/m² for residential buildings and 280 kWh/m² for non-residential buildings
- Extensive renovation of buildings could cut 36% of their energy consumption by 2030



Location data interoperability principles and methodologies applied in the Energy Efficiency sector – <u>use cases</u> (3/3)



- Generalisation at EU level of a digital platform for public lighting being implemented in Italy in 8.000+ Municipalities
- Energy Performance Certificates (EPC) of Buildings harmonisation
- Harmonisation of SECAP (Sustainable Energy and Climate Action Plans), to support smart communities made by 100+ municipalities of the same Province, CoM signatories
- Harmonisation of energy simulations to assess the energy heat demand of buildings
- Assessment of energy performance of buildings using energy consumption data from smart meters.
- Role of geospatial information in a regional energy strategy



State of the art data-driven methodologies and technologies for electricity characterisation of districts







Data requirements

- Case study source
- High frequency electricity consumption



- Building' area
- Building' uses
- Weather data

Sede Electrónica del Catastro





Optional data requirements

- Building' year of construction
- Building' volume
- Socioeconomic indicators
 - Amount of annual incomes and sources
 - Number of people per household
 - o Gini index
 - Population age

Case study source

Sede Electrónica del Catastro

Instituto Nacional de Estadística



Geographical levels



Case study: Postal code, census tract, building level





Geographical levels

- Electricity consumption

 Postal code level
 - By economic sector and tariff
- Weather data
 - Estimated at certain lat/lon
- Cadaster
 - **o Building level**
- Socioeconomic data
 - Census tract level





Methodology





Methodology steps

- 1. Reshape all data to postal code level
- 2. For each postal code:
 - 1. Data cleaning of the electricity consumption data
 - 2. Inferring usage patterns with clustering and classi
 - 3. Regression model
 - Output: Consumption/built area
 - Inputs: Calendar, usage patterns and weather features
 - Disaggregate the electricity consumption to baseload, heating, cooling and holidays terms
- 3. Web application interface to visualise the results



Data cleaning

Valid consumption range per contract, tariff and economic sector:

 $0 \le Q_t^e \le P_t^e \Delta t$

- Data padding: **detect gaps**
- **Z-score** using a rolling window of two weeks $\frac{x \bar{x}}{\sigma(x)} \ge 4$
- Creation of an all-contracts synthetic tariff and economic sector



Inferring usage patterns

- Based on relative daily load curves during:
 - March to May + September to November (No much weather dependence)
 - Avoid COVID-19 lockdown periods and holidays

Using Spectral clustering

- Alternatively, a Gaussian Mixture Model could be used
- Kernel: Polynomial of degree=2
- Best K: Gap in the eigenvalue spectrum
- Classification of non-clustered days using Multinomial logistic regression model and calendar features



Inferring usage patterns







Modelling

Penalised linear regression model

 L1-regularisation

$$\begin{aligned} Q_t^e &= (B_t \times s_t) + (H_t \times dh_t) + (C_t \times dh_t) + \varepsilon_t \\ & \text{baseload} \quad \text{heating} \quad \text{cooling} \end{aligned}$$

$$B_{t} = \omega_{b} + S_{N_{d}}(p_{t}^{d}) + S_{N_{w}}(p_{t}^{w})$$

$$S_{N_{d}}(p_{t}^{d}) = \sum_{n=1}^{N_{d}} \omega_{b,d,n,cos} \cos(2\pi n p_{t}^{d}) + \omega_{b,d,n,sin} \sin(2\pi n p_{t}^{d}) \qquad p_{t}^{d} = \frac{dh_{t}}{24}$$

$$S_{N_{w}}(p_{t}^{w}) = \sum_{n=1}^{N_{w}} \omega_{b,w,n,cos} \cos(2\pi n p_{t}^{w}) + \omega_{b,w,n,sin} \sin(2\pi n p_{t}^{w}) \qquad p_{t}^{w} = \frac{wh_{t}}{168}$$



Modelling

$$H_t = \omega_{h,lp}^+ T_t^{h,lp} + \omega_h^+ T_t^h + \omega_{ah}^+ A_t^h$$
$$C_t = \omega_{c,lp}^+ T_t^{c,lp} + \omega_c^+ T_t^c + \omega_{ac}^+ A_t^c$$

$$\begin{split} T^{h,lp}_t &= (T^{bal,c}_{dh_t} - T^{o,lp}_t) d_{s_t} & T^{c,lp}_t = (T^{o,lp}_t - T^{o,lp}_t) d_{s_t} \\ T^h_t &= (T^{bal,h}_{dh_t} - T^o_t) d_{s_t} & T^c_t = (T^o_t - T^o_t) d_{s_t} \\ A^h_t &= W^s_t T^h_t d_{s_t} & A^c_t = \\ T^{o,lp}_t &= \begin{cases} \alpha T^o_t & \text{if } t = 0, \\ \alpha T^o_t + (1 - \alpha) T^{o,lp}_{t-1} & \text{if } t > 0. \end{cases} & \alpha = 1 - e^{-t_{samplin}} \\ d_{s_t} &= \begin{cases} 1 & \text{if weather dependence in } s_t, \\ 0 & \text{if no weather dependence in } s_t. \end{cases} \end{split}$$

Example of low-pass filter



$$T_t^{c,lp} = (T_t^{o,lp} - T_{dh_t}^{bal,c})d_{s_t}$$
$$T_t^c = (T_t^o - T_{dh_t}^{bal,c})d_{s_t}$$
$$A_t^c = W_t^s T_t^c d_{s_t}$$
$$\alpha = 1 - e^{-t_{sampling}/(2\pi\tau/24)}$$



Modelling

Training period:

75% training25% validationDays randomly distributed



Known terms and time series: Unknown fixed terms: $\tau(*)$, N_d and Q^e , s, p^d , p^w , dh, wh, T^o , W^s and N_w . $t_{sampling}$.

Unknown terms for each usage pattern: ω_b , $d_s(*)$, $\omega_{b,d,n,sin}$, $\omega_{b,d,n,cos}$, $\omega_{b,w,n,sin}$ and $\omega_{b,w,n,cos}$. Unknown terms for each day part: $\omega_{h,lp}^+$, ω_h^+ , ω_{ah}^+ , $\omega_{c,lp}^+$, ω_c^+ , ω_{ac}^+ , $T_{dh}^{bal,h}(*)$ and $T_{dh}^{bal,c}(*)$.

(*) Estimated using a genetic algorithm optimizer



Disaggregation and consumption KPIs



Clustering and classification of daily load curves



Aggregated electricity disaggregation (%)

Baseload Heating Cooling Holidays Covid Lockdown

20

Cluster

— 01

— 02

— 03

— 04

— 05

- 06



Disaggregation and consumption KPIs







Web application / KPIs on a map

- All KPIs can be represented
- Filtering capabilities





Web application / KPIs on detail

- **Dissagregation details** at each postal code, economic sector and tariff level.
- Aggregated results to understand tendencies in energy consumption





Web application / Benchmarking

- Weather-normalised comparison of the consumption KPIs between two different postal codes
- **Time-varying** benchmarking





Web application / KPIs correlation

• Infer correlations between diferent

KPIs, either consumption, socioeconomic and cadaster

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Conclusions

- Inference about buildings and their occupants in terms of energy consumption
 - User behaviour patterns
 - Useful to detect self-consumption potentiallities and occupancy patterns of buildings
 - Weather dependence
 - Useful for assessing the Energy Performance of Buildings
 - Baseload quantification
 - Useful for Energy Efficiency estimation of non-weather dependent consumptions



Conclusions

- Webapp functionalities:
 - KPIs in multiple types representations
 - KPIs filtering and correlation
 - Provides an interesting sandbox for testing visualisations



How these methodologies and technologies may support the business of start-up companies?



• Added value:

- Emerging business models addressing the concept of "energy communities" may benefit from these data driven methodologies
- District heating/cooling planning may need these tools to characterize the RES generation and the energy demand
- Public administrations need these methods to take decisions in energy transition plans and in urban planning
- Engineering and consultancy companies may require these data driven methods to substantiate their energy services



Challenges to market monetization:

- More effort on digitalization of thermal energy data sources is required (e.g. gas meters)
- The geographical scale of available data is too high (zip code)
- There is a need for standard procedures to downscale data from the zip code to the district level
- The market for added-value services based on geo-referenced data is not very clear
- The energy services companies are showing interest but actually there is no direct link with their business models



How these methodologies and technologies may support urban planning and energy efficiency policies of a Regional Government?



First case study: **Department of vice-presidency, territory and digital policies of the Generalitat de Catalunya**

- Main objectives:
- To develop and implement a georeferenced web environment able to gather, harmonize, visualize and provide KPIs for several use cases related to industrial parks, housing developments and residential districs
- 2. To validate the KPIs in the coastal area of Catalonia (Including Barcelona city)







First case study: **Department of vice-presidency,territory and digital policies of the Generalitat de Catalunya**

- Outcomes:
- A web environment based on a QGIS backend and a map viewer, to graphically show the KPIs and the data analysis result
- 2. A set of statical learning tools designed to support the administration in generating land planning scenarios





First case study: **Department of vice-presidency,territory and digital policies of the Generalitat de Catalunya**





First case study: Department of vice-presidency,territory and digital policies of the Generalitat de Catalunya





Second case study: H2O2O-ePLANET: European Public Local Authorities' Network for driving the Energy Transition

• **Objectives**:

- To establish a transparent information sharing framework among public authorities
- To deploy multi-level working groups of related to enable a collaborative approach in development and updating of Energy Transition plans
- To enable digitalization of the energy transition plans





Second case study: H2020-ePLANET: European Public Local Authorities' Network for driving the Energy Transition

• Pilot sites:

Region	Municipalities	Municipalities with SEAC
Girona	221	199
Crete	24	21
Zin	307	276





Urban energy simulation supporting the Local Digital Twins toolbox



SOURCES

- Utilities
- Consumers
- Databases
- Simulation models

PURPOSE TODAY

- Billing, network control
- Cost control
- Building design
- Urban development

GEOLOCALISATION

- Intelligent networks
- Producers farms
- Decision making
- CO₂ footprint



Georeferenced energy information



Energy data challenges

- Data held by platform economy companies is of high interest
- Appropriate **governance** > multiple data sources and background
- **Organizational, legal and cultural challenges** (forthcoming Data Act)
- Stakeholders priorities
- Data quality



City/neighbourhood modelling





national building stock models



Companies' challenges

- There is a need for **more clarity on the legislative framework**
- Lack of interoperability between different datasets
- Companies look for **scalable data sharing means** compatible with their business plans

Citizens' challenges

- Energy **bill**, Energy **poverty**
- Sustainability



Next steps



DEP



Next steps

- B2G data sharing
 - Partnerships and ecosystems
 - EU Data Spaces
 - EU regulatory framework: Data Act, AI security
- EU Data Spaces for energy/Smart Communities
 - Energy transition
 - Data market ecosystem development
 - Carbon tax
 - Consumer farms...





Key messages, challenges and future outlook



- Downscale challenges: is it a matter of granularity of statistical and socio-economic data or of privacy issues related to energy consumption data or both?
- Future outlook: re-use of the methodology presented to support the implementation of the energy efficiency related actions of the Recovery and Resilience National Plans





Thank you



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