#### ElectroCap Pitch Deck



# Smart Grid Optimization through V2G System and Blockchain Transactions

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**TÉCNICO LISBOA** 

#### Team



#### Advisors and Mentors





#### Problem definition

- Challenges in maintaining load stability
- Intermittent nature of Renewable Energy Sources (RES)
- Lack of optimization for grids with mixed generation sources
- Need for decentralized energy storage solutions
- Challenges posed by the increasing adoption of electric vehicles (EVs)
- The absence of a charging management strategy: additional demand, power grid congestion, and energy import costs
- Potential for EVs to assist in balancing the grid
- Issues with traditional energy market transactions

### Solution beneficiaries

- Citizens and businesses reliant on the power grid
- Energy market participants
- Electric vehicle owners
- Energy producers
- Grid operators: Distribution System Operator (DSO) and Transmission System Operator (TSO)



### Technological solution

- Predictive Machine Learning models:
	- Energy Consumption Forecast
	- Renewable Energy Production Forecast
	- Dynamic Market Price Forecast
- Blockchain Transactions
- Machine Learning algorithms for optimizing grid flexibility and load distribution
- Real-time Adaptation to Energy Demand: more responsive and efficient energy distribution system
- Market dynamics and algorithms for profit maximization

### Competitors and previous work

#### **Competitors Synop** Synop**,**, POCITYF, **POCITYF,**EnerNoc **EnerNoc**(EnelX **(EnelX**), AMS, **), AMS,** Siemens **Siemens**

#### **Previous work**

These companies are developing solutions for optimizing energy management, integrating renewable sources, smart grid technologies, microgrid solutions and enhancing grid flexibility.

#### Solution requirements

- The prediction models should be capable of giving accurate information about the energy market to correctly calculate the user's gains
- The monetary transactions need to be fast, secure and trustworthy
- Different energy usage profiles need to be created to satisfy a wide range of users
- Limits to the energy transactions between the vehicles and the grid need to be established in order to reduce the impact on the vehicle's battery lifetime and daily life of the user





### Technical challenges

- Ensuring the quality of the data used to train the ML models, and consequently the quality of the models themselves
- Integration of the prediction models into existent infrastructure
- Volatility of the energy market and how exceptional cases might influence the final users
- Implementing a system that correctly reflects the user's needs
- Immutable smart contract security vulnerabilities
- Gas optimization for cost-efficient smart contract execution
- Legal and regulatory compliance challenges in decentralized smart contracts

### Testing and validation metrics

- Impact on the Load Profile
- Total energy exchanged with the grid
- Peak Load Reduction and Peak Load Shifting
- Load Flattening
- Total monetary value exchanged
- EV owner's monetary gains
- Scalability (how does the model respond to different numbers of connected EVs)
- Transaction security
- Prediction accuracy



#### Project Organization



### Division of labor (1)



### Division of labor (1)



### Division of labor (2)



### Division of labor (2)



#### **BLOCK DIAGRAM**



#### User Interface

#### **User Interface**

Choose the values taking into account the given parameters:



Simulate Now

Or Generate random values:

**Generate Now** 

#### Achieved Results - ML Energy Consumption Model



Actual and predicted consumption – 01/04/2023

The model can generate predictions on the amount of energy consumed in Portugal per intervals of 15 minutes for a given day of the year or a specific time period

After statistical study, the model was trained with a large database that follows a coherent trend, from 2011 to 2023, to provide accurate predictions

consumption based on historical data, weather conditions and time context

## Model Metrics and Evaluation

- Training Loss:
	- Measure of how well the model is fitting the data during training.
	- Used during the optimization process
	- It decreases as the model becomes more accurate at the task being trained
- Validation Loss:
	- Measure of how well the model is generalizing to data not seen during training
- Optimization Strategy:
	- Improving model accuracy through reinforcement of weekly and annual periodicity



Training and Validation Loss

#### Achieved Results - ML Energy Production Model

The model can generate predictions on the amount of energy generated in Portugal per intervals of 15 minutes for a given day of the year or a specific time period on multiple sources of energy generation (Solar, Wind, Natural Gas, etc)

RFS (Random Forest Regression)

Training data taken from 2023 in order to give more accurate predictions. (Using older data would not give an accurate depiction of the current energy generation in Portugal due to the great changes that occured in this sector in the last decade)

The model uses information on meteorologic data and other variables like day of the week and if it's a holiday to further tweak the results



Example of evolution of daily production of renewable energy (Winter) from 2010 to 2023



Example comparison between real data (blue) and prediction (orange) for energy generated by hydro in a specific day

### Achieved Results - ML Energy Price Model

The model can generate predictions on the wholesale enrgy market price in Portugal per intervals of 15 minutes for a given production and consumption data.



The model was trained with the Random Forest Regressor model using a database from 2015 to 2023 in order to provide accurate predictions, This extensive database enhances the robustness of the model.

The model uses information about the production and consumption combined with energy price data to allow for futher tweaking of the results (data sources: REN, OMIE).



Actual and predicted wholesale energy price  $(E/MW) - 22/04/2023$ 

To test the model, a forecast was made for the wholesale energy market price for 04/2023, calculating the Mean Absolute Error (MAE) and Coefficient of Determination (R²) on the results obtained on each day. The average MAE value was 3.552, and the average  $R<sup>2</sup>$  value was 0.958, suggesting that the model is performing well.

### Achieved Results - ML Energy Price Model

The model predicts wholesale energy market prices. However, to make predictions more relevant, it scales them to match the average prices consumers encounter. The scaling process incorporates reference values of energy price [€/MWh] for a residential and work profile.



#### Research: The Potential of V2G Technology

- 1. Investigation into V2G Technology
	- V2G Technology Overview
	- V2G Services and Advantages
- 2. Main Actors and System Architecture
- 3. Challenges and Decarbonization
- 4. Analysis of Charging Profiles
- 5. Comprehensive Examination of V2G Impact
	- Consumption Patterns Analysis
	- Assessment of Load Management Flexibility
	- Prediction of EV Penetration Effect (2025 to 2040)
	- Understanding Grid Implications
	- Quantification of Key Performance Indicators



## Load Profiles



Workplace Charging Stations

Public Charging Stations



**COLOR** 

Scalable load profiles depending on the number of EVs in the national fleet and installed power at charging stations





# Ideal Transactions

- Adjusts charging intervals using Vehicle-to-Grid (V2G) technology.
- Uses predictive modeling and real-time data for efficient charging.
- Maintains battery charge within specified limits to extend life.
- Optimizes charging to reduce energy costs.
- Maximizes use of renewable energy sources.
- Schedules charging to balance grid demand and reduce peak loads.
- Enhances reliability and efficiency of the energy distribution network.
- Adapts to changing conditions for optimal battery management and grid stability.

# Ideal Transactions - Results



# Ideal Transactions - Results



# Ideal Transactions - Impact



Figure 3. Before load management algorithm

Figure 4. After load management algorithm

#### Achieved Results - Blockchain Transactions



Fictional virtual wallets on Ganache

#### Achieved Results - Blockchain Transactions



#### Transactions record

#### Achieved Results - Blockchain Transactions



Blockchain

#### Achieved Results - Prototype Presentation



Mockup simulation

#### Achieved Results - Prototype Presentation









Mockup Presentation

#### Achieved Results - Site & Marketing

- Developed a functional website as per project requirements
- Ensured compatibility across devices and browsers for a consistent user experience
- Successfully published the website online with proper hosting configuration for stability and performance
- Integrated a blog for updated content and incorporated visual effects for enhanced aesthetics and interactivity
- Integrated visual effects to enhance the aesthetics and interactivity of the site
- Utilized CSS, JavaScript, and frameworks to efficiently implement desired effects
- Conducted thorough testing to ensure functionality across scenarios

```
\Diamond index.html \Diamond ...
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          title>
            Smart Grid Optimization through V2G System and Blockchain Trans
          \langle/title>
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 $\mathbf{1}$ 

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f_{\text{O}}nt-size: 16px
```


### Challenges faced by the team

- Defining requirements for the prototype
- Selecting the appropriate data for model training
- Finding suitable datasets
- Complexity of implementation

#### Contribution of each team member (1)



#### Contribution of each team member (1)



#### Contribution of each team member (2)



#### Contribution of each team member (2)



#### Vídeo Demonstração



<https://drive.google.com/file/d/1lIxI9gIOx-L1yBhjn-3-uDgCeWIVOgGB/view?usp=sharing>

#### Conclusion

The innovative approach to smart grid optimization through Vehicle-to-Grid (V2G) technology and blockchain transactions addresses critical challenges in the energy sector.

By leveraging predictive machine learning models, we have developed robust solutions for energy consumption and production forecasting, dynamic market price prediction, and real-time energy distribution adaptation.

Our blockchain implementation ensures secure and efficient transactions, while our comprehensive mock-up and user interface demonstrate practical applications and user benefits.





# For more information or any questions scan the QR code below and visit our website



