

Energy Shield AI

Electricity Theft & Anomaly Detection System

Project Overview

Energy Shield AI is an end-to-end, production-oriented system designed to **detect electricity anomalies and potential theft** using smart-meter data.

It combines **large-scale data engineering**, **unsupervised machine learning**, and **human-interpretable explainability** into a deployable solution.

The system is inspired by real-world utility challenges where **ground-truth theft labels are scarce**, making traditional supervised learning unreliable.

Dataset Description

- **Source:** London Smart Meter Dataset (Kaggle)
- **Scale:**
 - ~3.47 million records
 - 71 engineered features
- **Granularity:**
 - Daily household consumption derived from **48 half-hour readings (hh_0 - hh_47)**

Key Feature Categories

1. Consumption Statistics

- `daily_sum`, `daily_mean`, `daily_std`
- `energy_min`, `energy_max`, `energy_median`
- `peak_intensity`

2. Behavioral Indicators

- `flatness_index`
- `peer_ratio`

3. Environmental Context

- `temperatureMin`, `temperatureMax`, `temp_gap`

4. Socio-Economic Signal

- `Acorn_grouped`, `Acorn_encoded`

These features enable **behavioral profiling**, not just raw consumption analysis.

Modeling Strategy

Why Unsupervised Learning?

- Electricity theft is **rare and weakly labeled**
- Behavior changes over time
- Patterns differ by household type

➡ The problem is best framed as **anomaly detection**

Model Selection: Isolation Forest

Isolation Forest was chosen as the core model because:

- Designed specifically for anomaly detection
- Scales to **millions of rows**
- Works well with engineered tabular features
- Requires no labeled theft data
- Widely used in industry for fraud & energy analytics

Training Highlights

- Input: 15 carefully selected behavioral features
 - Scaling: StandardScaler applied and persisted
 - Output:
 - -1 → Anomalous (suspicious behavior)
 - +1 → Normal
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Feature Engineering (Core Signals)

Feature	Meaning
Peer Ratio	Household usage vs similar socio-economic group
Flatness Index	Unnaturally constant consumption (tampering signal)
Peak Intensity	Strength of daily consumption peaks
Temp Gap	Sensitivity to environmental change

Most detection power comes from **feature design**, not model complexity.

Backend Architecture (FastAPI)

The inference layer is built using **FastAPI**, enabling:

- Low-latency predictions
- Structured input validation (Pydantic)
- Production-grade deployment

API Flow

1. Receive household summary metrics
 2. Reconstruct engineered features
 3. Apply saved scaler
 4. Run Isolation Forest inference
 5. Return:
 - `status` (Normal / Suspicious)
 - `risk_score` (0–100)
 - `peer_ratio`
 - Explainable summary
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Frontend Interface (Streamlit)

A clean, professional **Streamlit dashboard** provides:

UI Sections

1. Input Simulation Panel

- Energy usage
- Variability
- Temperature
- Socio-economic group

2. Final Verdict

- Clear Normal / Suspicious status
- Risk score gauge

3. Key Risk Indicators

- Peer comparison
- Weather sensitivity
- Consumption consistency

4. Explainable AI (XAI) Table

- Human-readable reasoning
- Feature-level interpretation

The UI is designed for **utility analysts**, not ML engineers.

Explainability (XAI)

Rather than black-box scores, the system explains:

- *Why* a household was flagged
- *Which behaviors* triggered concern
- *How* it compares to similar households

This builds **trust**, which is essential in governance and billing decisions.

Real-World Applicability

Energy Shield AI can be used for:

- Electricity theft detection
- Meter tampering alerts
- Abnormal usage monitoring
- Revenue loss prevention
- Governance & compliance audits

It mirrors how **real power utilities** deploy AI systems.

Final Outcome

Energy Shield AI demonstrates:

- Scalable ML on real-world data
- Correct use of unsupervised learning
- Strong feature engineering
- Production-ready API
- Interpretable, analyst-friendly UI

This is not just a model — it is a **complete, deployable AI system**.

Future Enhancements

- Household history timeline
 - Batch CSV inspections
 - SHAP-based explanations
 - Cloud & Docker deployment
 - Continuous learning pipeline
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Energy Shield AI — Smart Meter Governance System