



Advanced machine learning to transform utilities & system operators.

Lower Costs, Higher Reliability—All in Real Time.

Mohak Mangal, Stanford GSB

Dhruv Suri, Stanford Doerr School of Sustainability



Meet the team



Mohak Mangal

MA , MBA



THE WORLD BANK



J-PAL
ABDUL LATIF JAMEEL POVERTY ACT



Experienced Economist and
Statistician



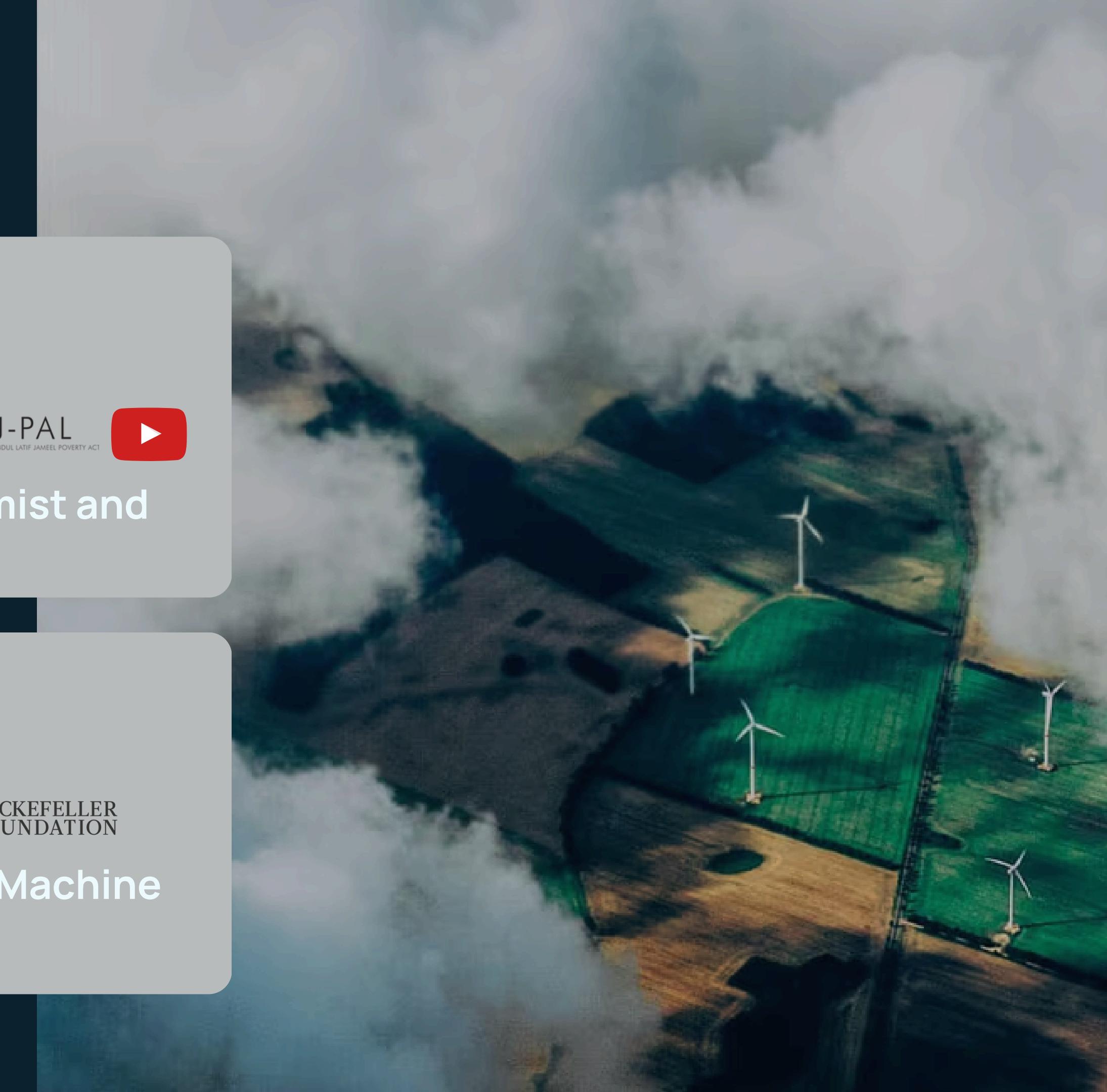
Dhruv Suri

PhD Candidate



The
ROCKEFELLER
FOUNDATION

Power Systems and Machine
Learning Engineer



Our Expert Advisors



Ines Azevedo



Professor, Stanford University

Leads a \$20 million DoE-funded consortium for energy reliability



Adam Brandt



Professor, Stanford University

Optimization pioneer, coupled-gas and electricity systems planning



Lincoln Bleveans



Sustainability Utilities & Infrastructure

Ex-utility executive; 30+ years experience in the energy sector

Utilities and system operators lose billions due to outdated forecasting and procurement methods



Senior executive of one of the
largest utilities in India
(Rs. 15,000 cr+ annual revenue)

“We live and die by our
ability to plan for the future
... If we get it wrong, the
cost implications for
customers and for our
business are huge.”

These costs have an economy-wide impact affecting businesses and consumers



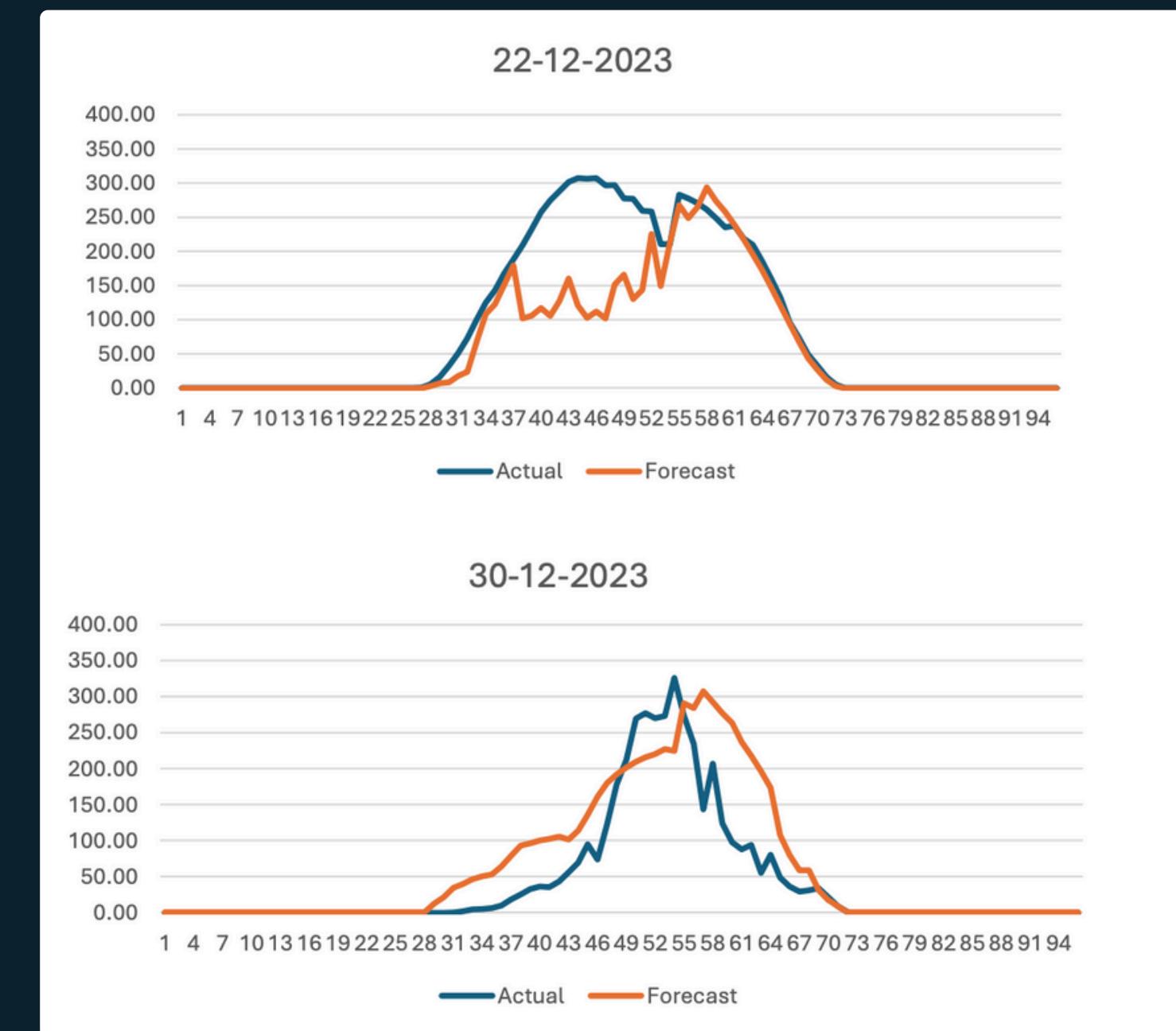
Source: PRS

We spoke to executives of 40+ utilities and system operators to understand their pain points



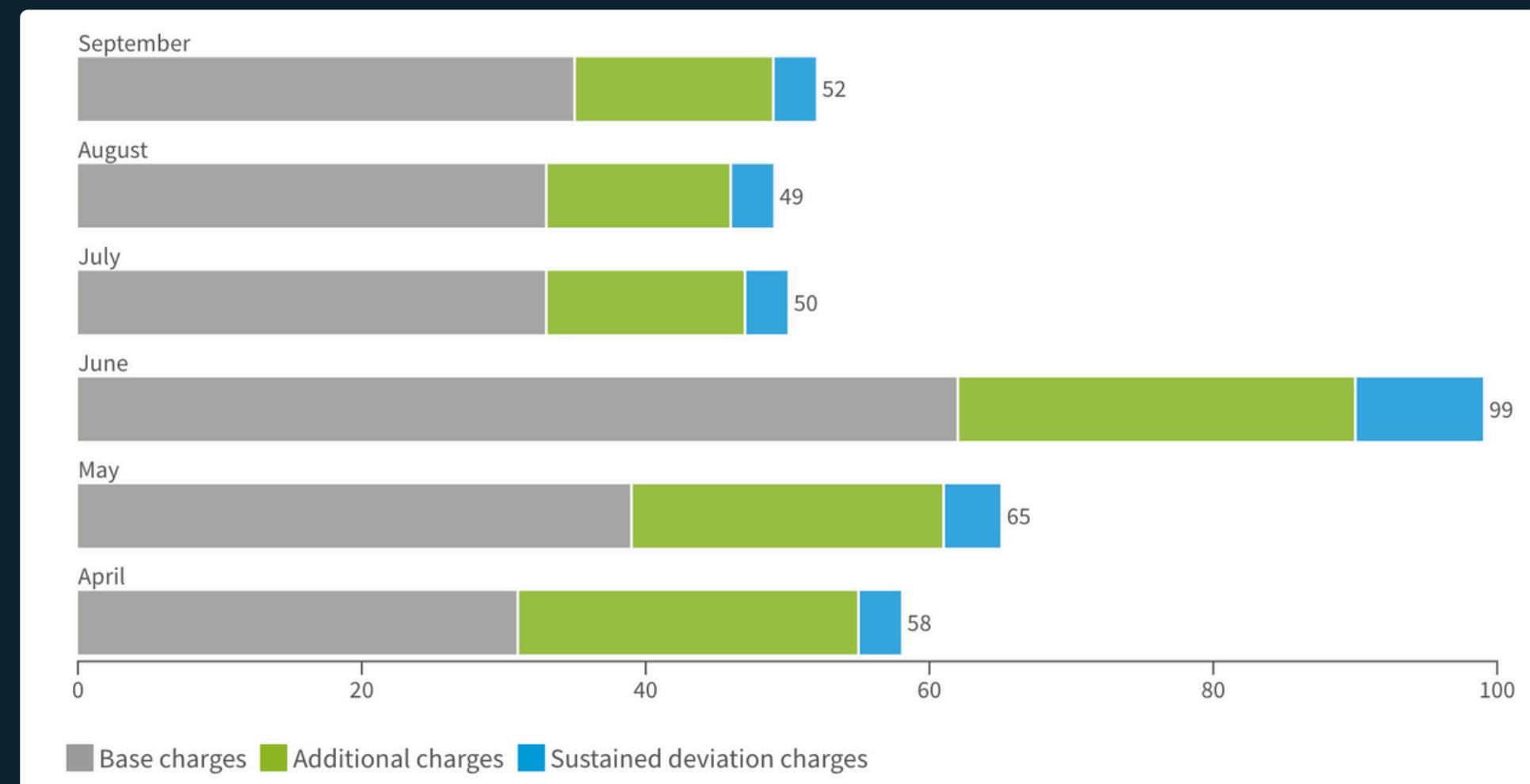
Current forecasting stack relies on deterministic models that don't reflect grid fluctuations

In November-December 2023, a 320 MW solar park in Bikaner produced far lower than its typical output in one day due to a sudden foggy and cloudy day.



This leads to frequent over-purchasing or under-purchasing of power, increasing losses

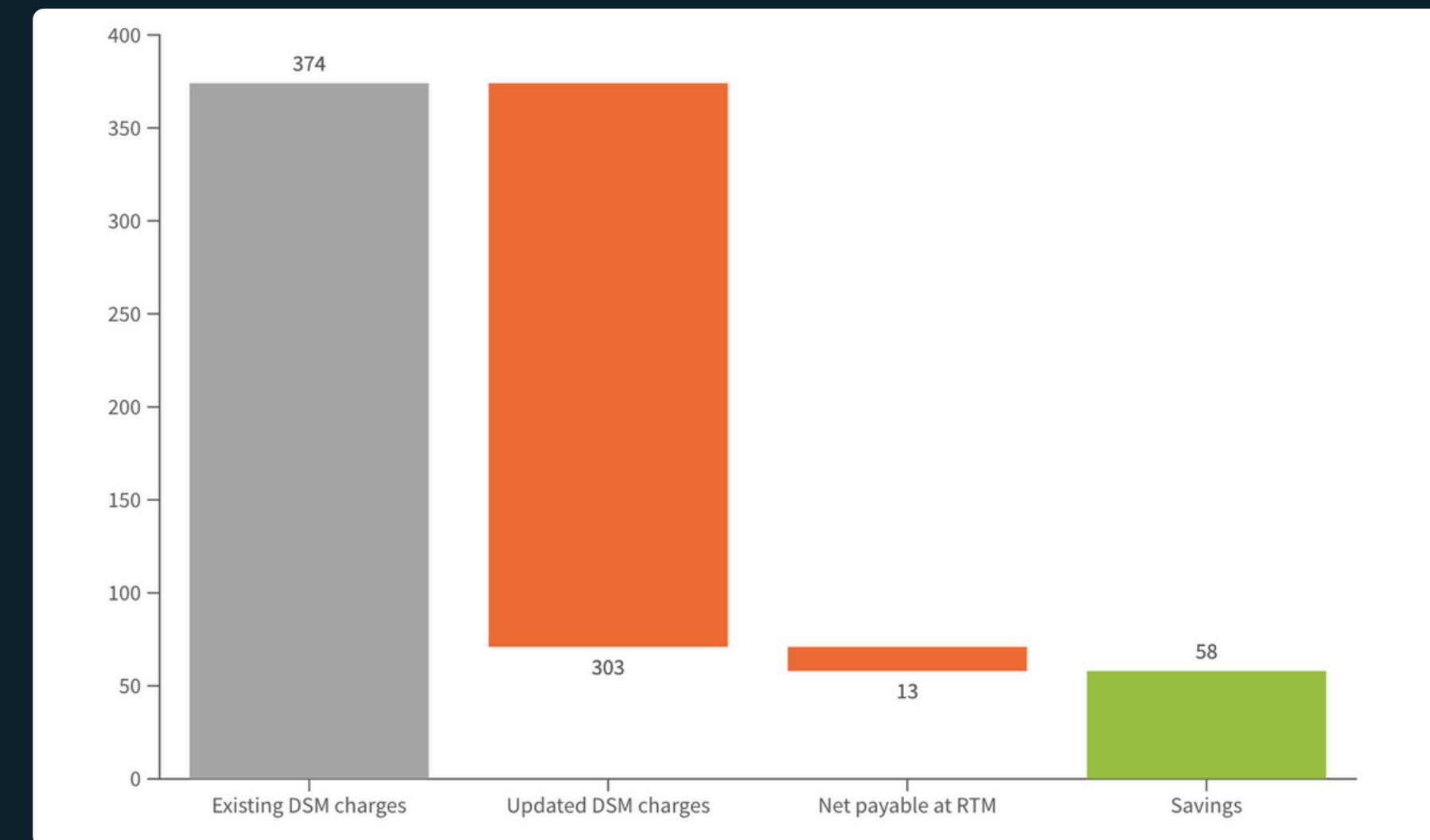
UP incurred **INR 374 crores** as DSM penalties during H1 FY23.



This leads to frequent over-purchasing or under-purchasing of power, increasing losses

UP incurred **INR 374 crores** as DSM penalties during H1 FY23.

By increasing its forecast accuracy to 95%, UP could have saved **~Rs. 60 crores**.



Source: CEEW

Several states are facing challenges due to inaccurate forecasting

Delhi Penalizes Three DISCOMs for Defaulting on Renewable Purchase Obligations

The commission asks distribution companies to pay the penalty within a month

October 10, 2019 / Anjana Parikh / Other, Renewable Energy,



Andhra Pradesh: ‘High demand, low availability leading to load-shedding’

There is a shortage of 50 million units of power, says official

Published – April 09, 2022 09:35 pm IST – VIJAYAWADA

STAFF REPORTER



READ LATER

PRINT

Telangana government rejects proposal to hike power tariff of consumers

By Srinivasa Rao Apparasu X , Hyderabad

Jan 30, 2025 08:06 AM IST

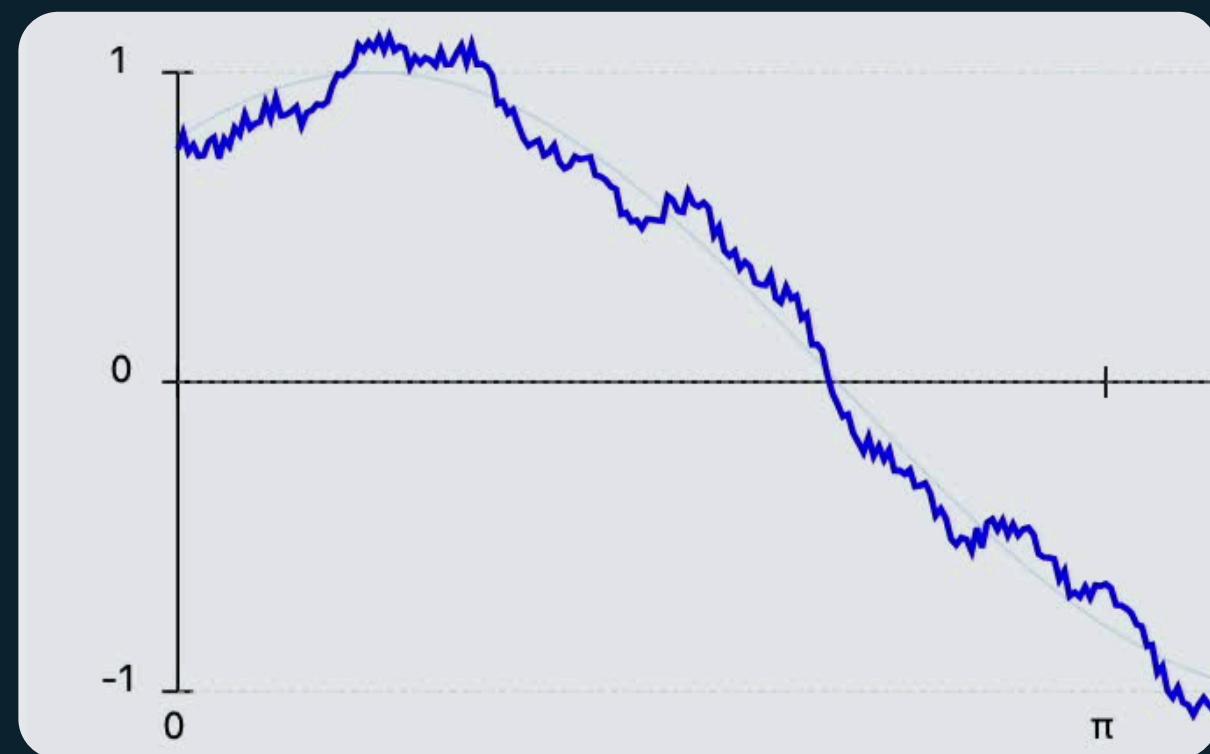


The Telangana government is unlikely to increase the power tariff for consumers in the state for the upcoming financial year 2025-26

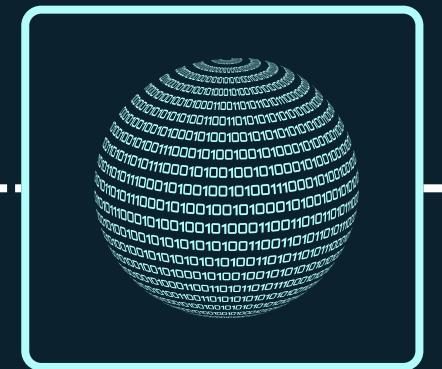


We cut the losses by enabling real-time optimization that keeps costs down and reliability up

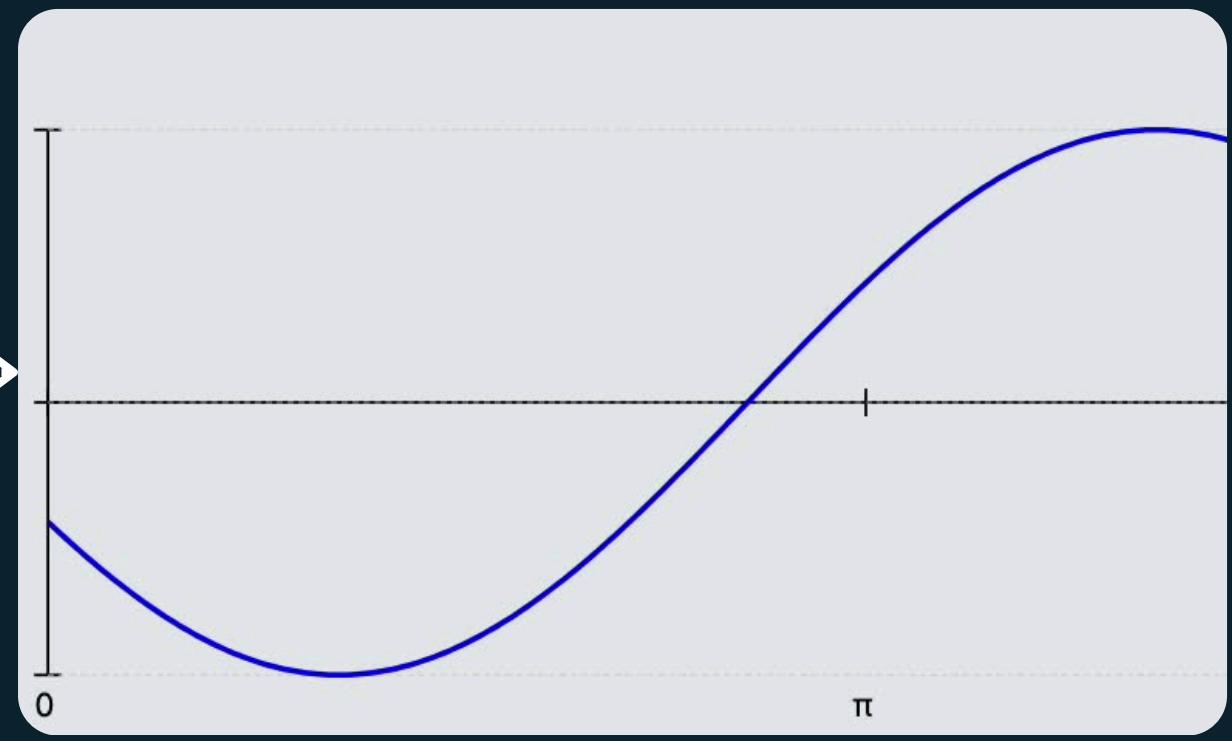
Unpredictable Grid



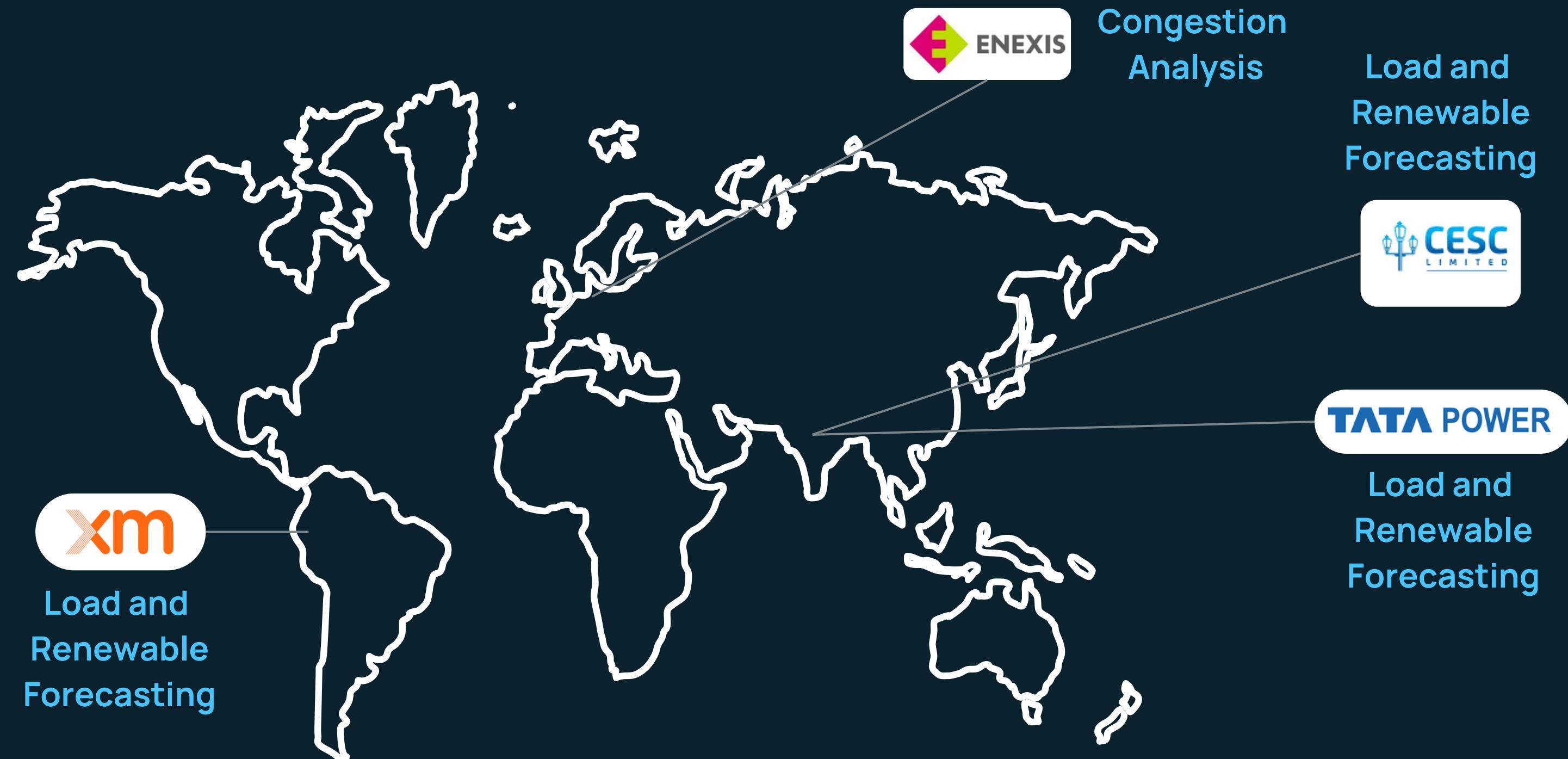
Machine
Learning



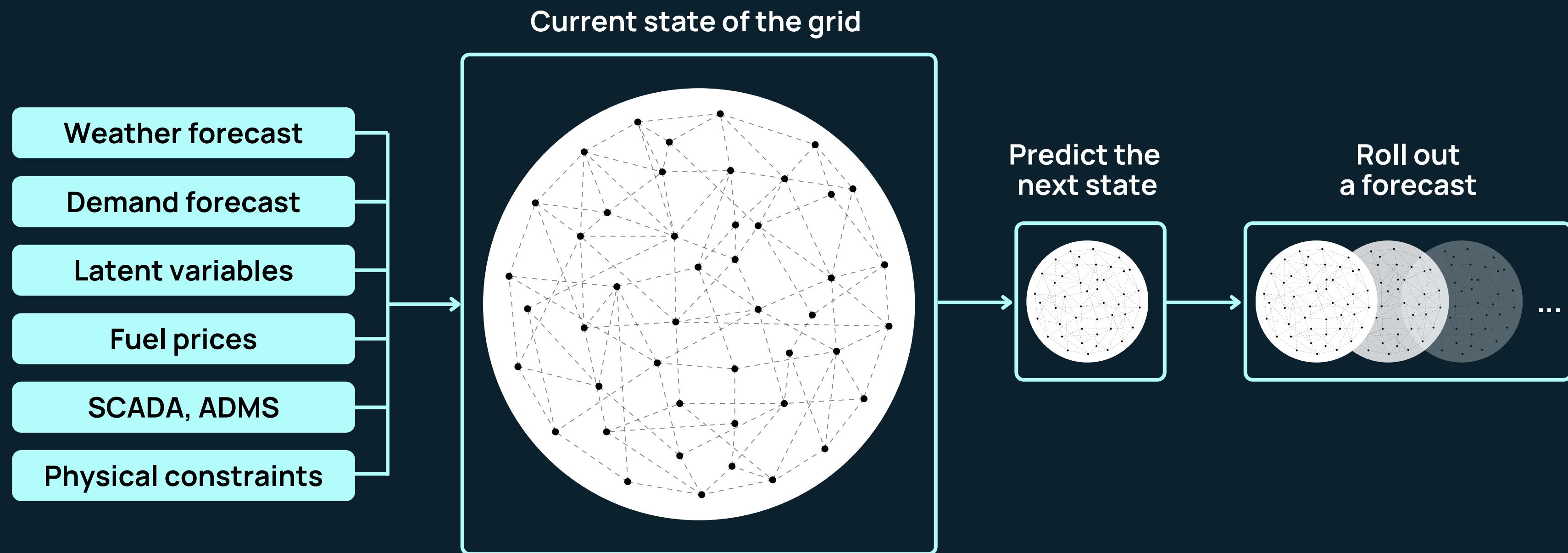
Predictable Grid



We are working with the best utilities and system operators

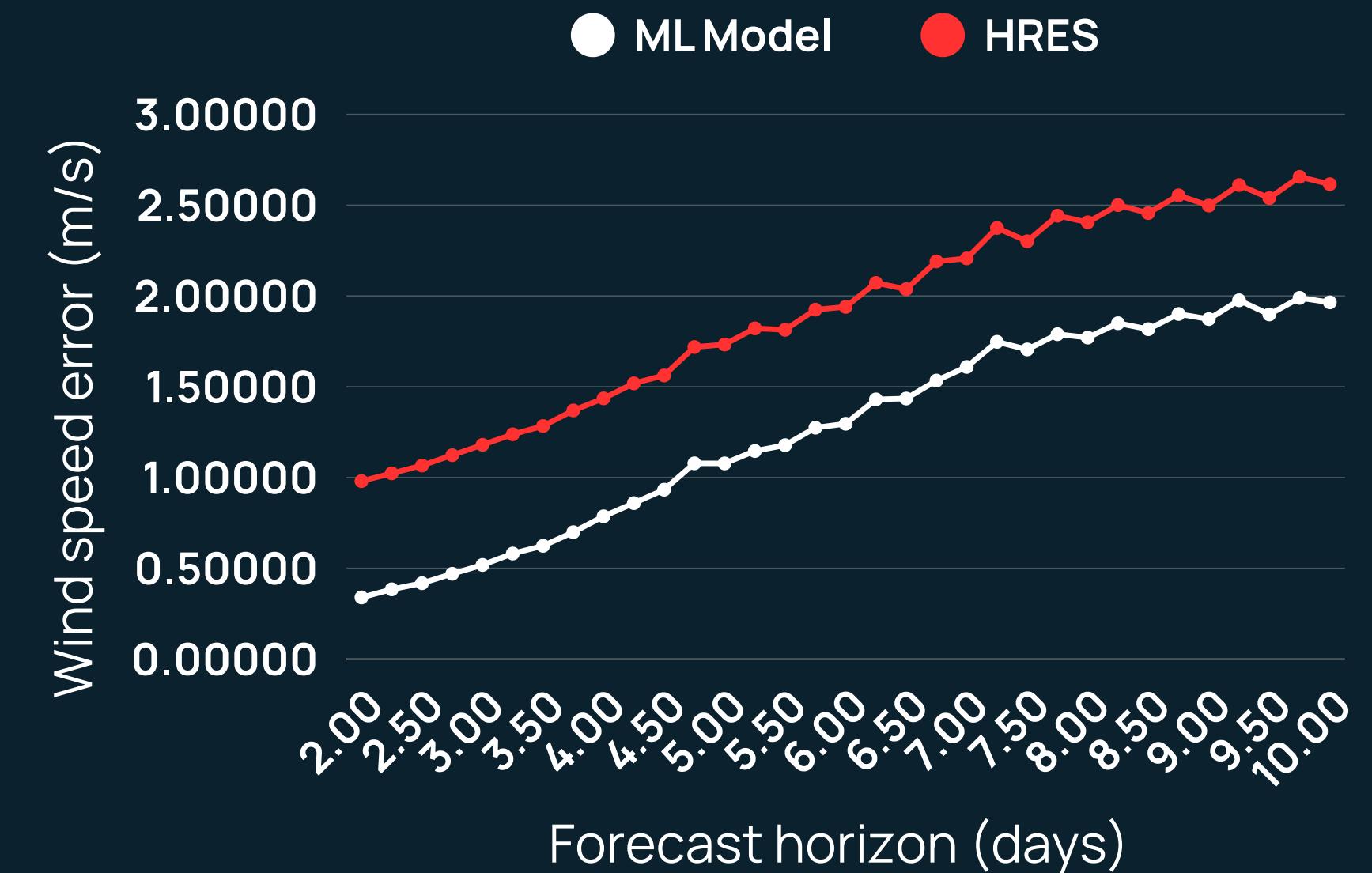
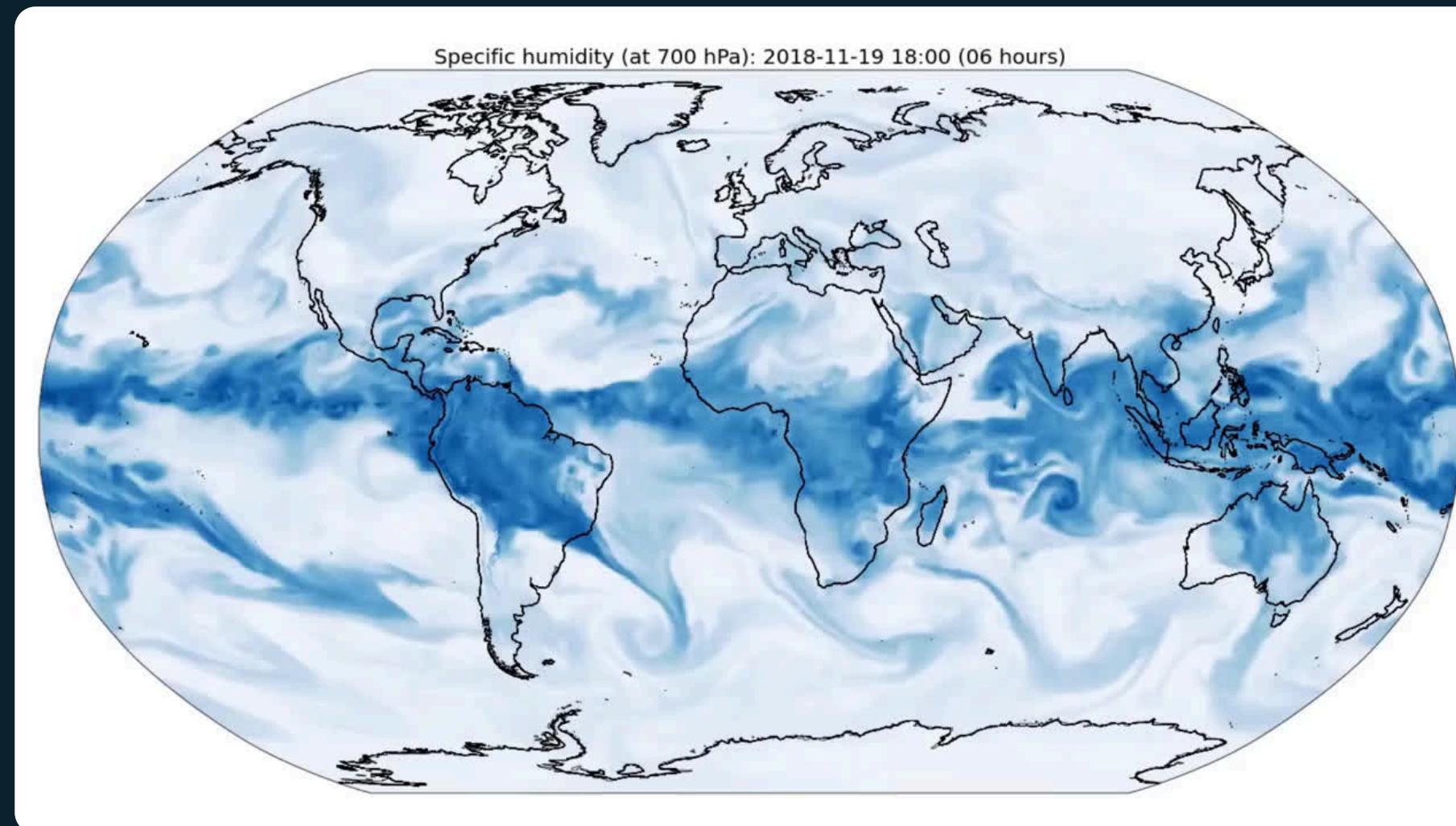


We combine deep learning with utility data to give you the most accurate forecasts

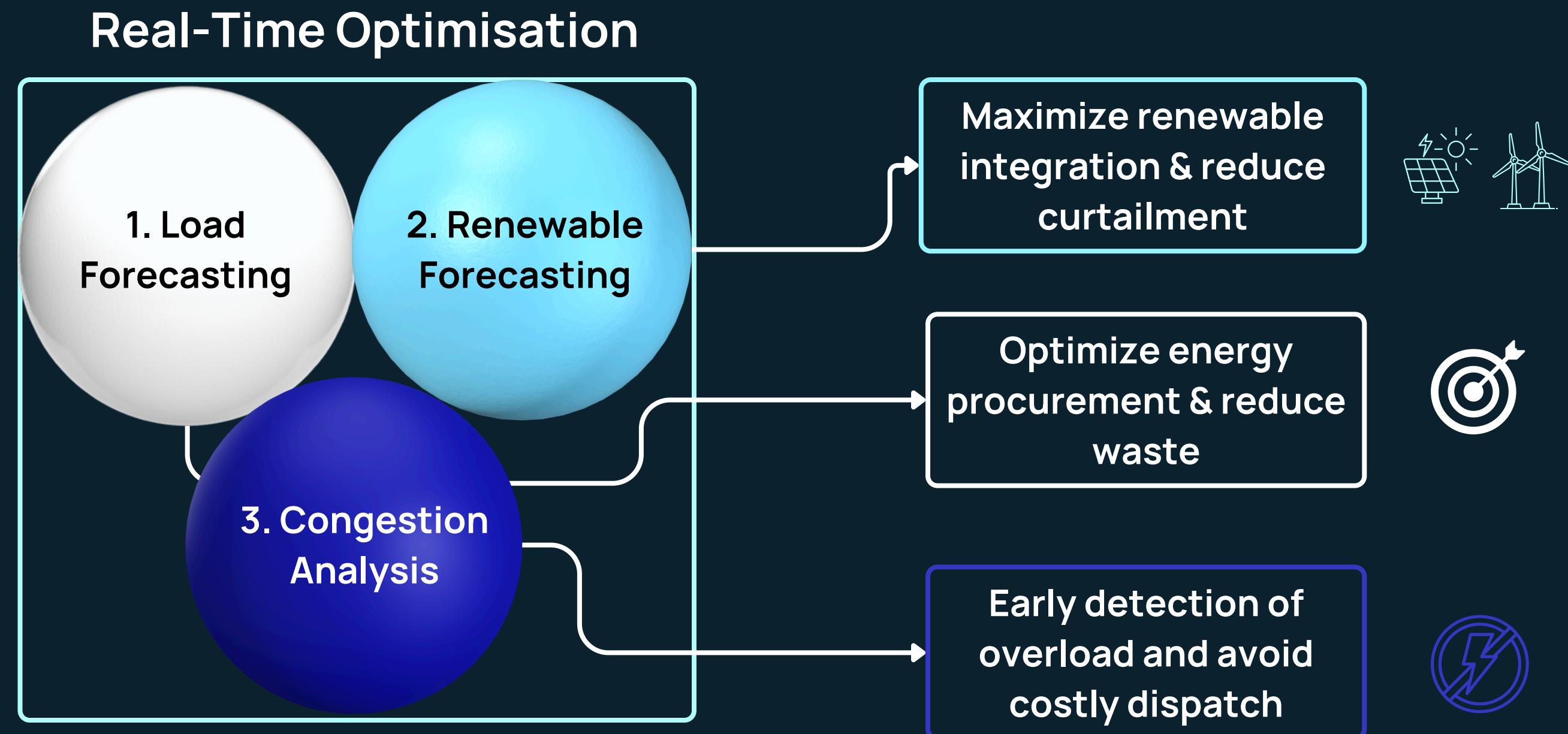


Our ML models beat the State of the Art

In a partnership with Chile's system operator, we built a machine-learning weather prediction model in 90 days that outperformed ECMWF HRES predictions.

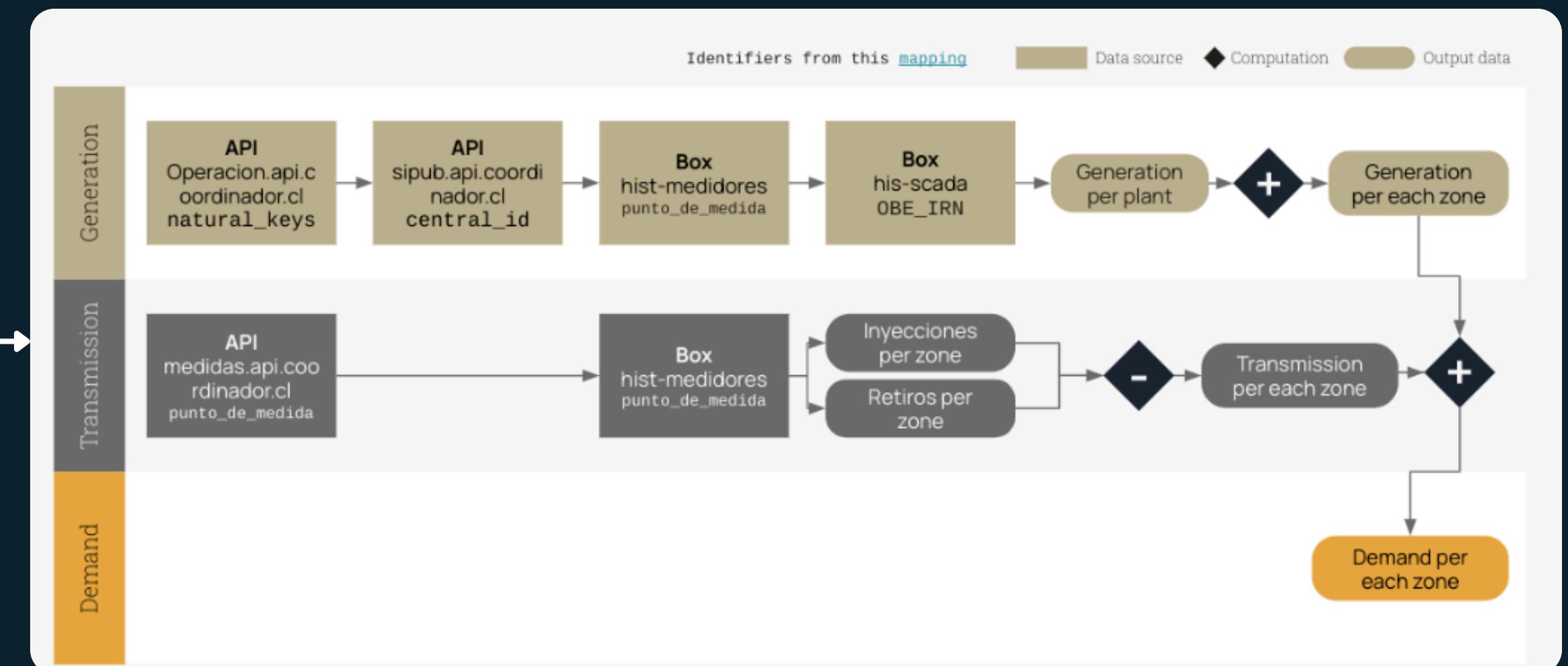


We are laser-focused on three core use cases

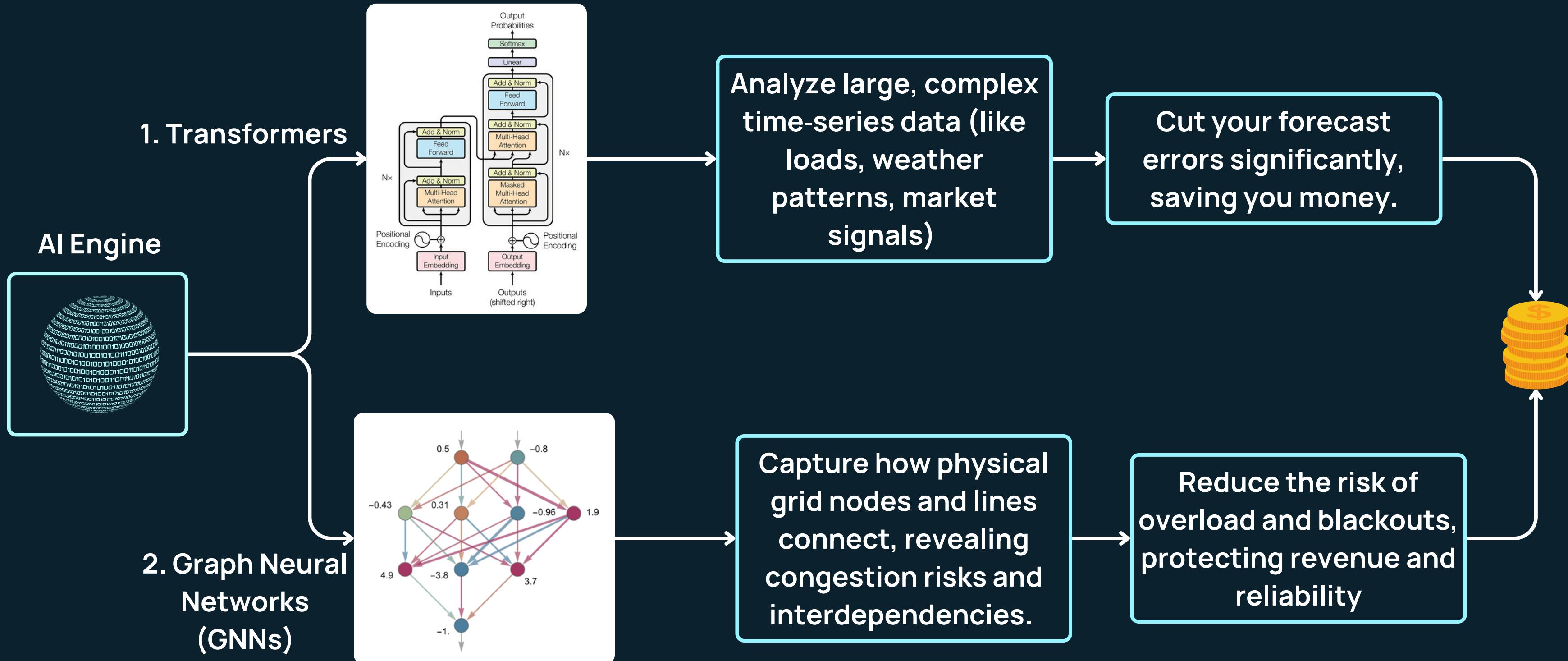


We ensure seamless integration with your existing systems

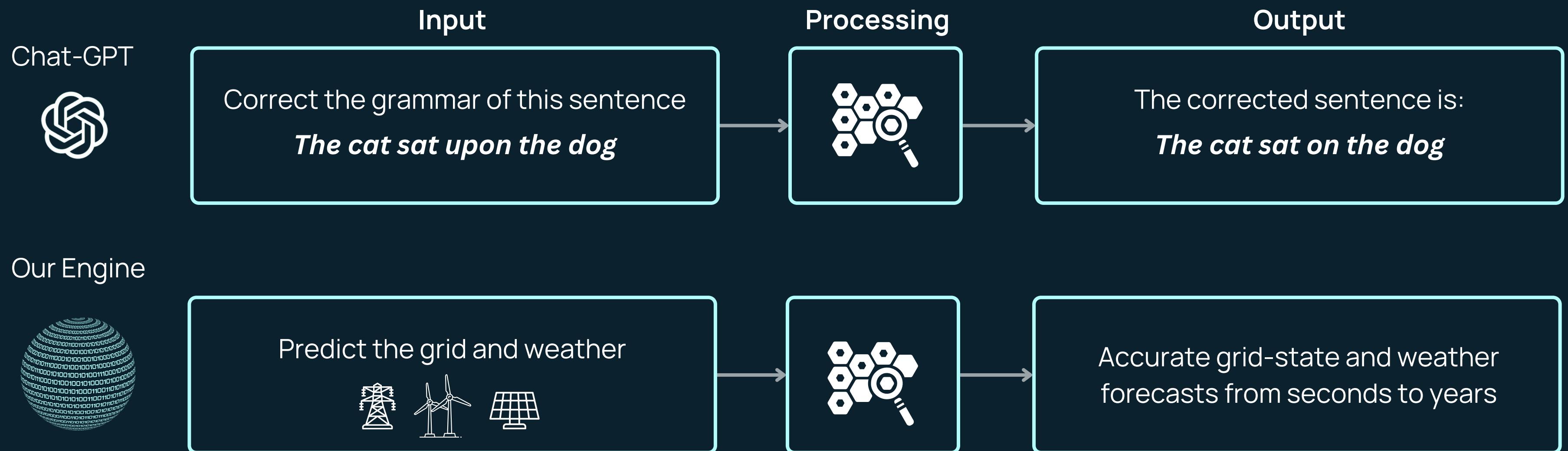
For the Chilean system operator, we designed the system architecture to ensure seamless integration.



Our AI engine uses two key approaches

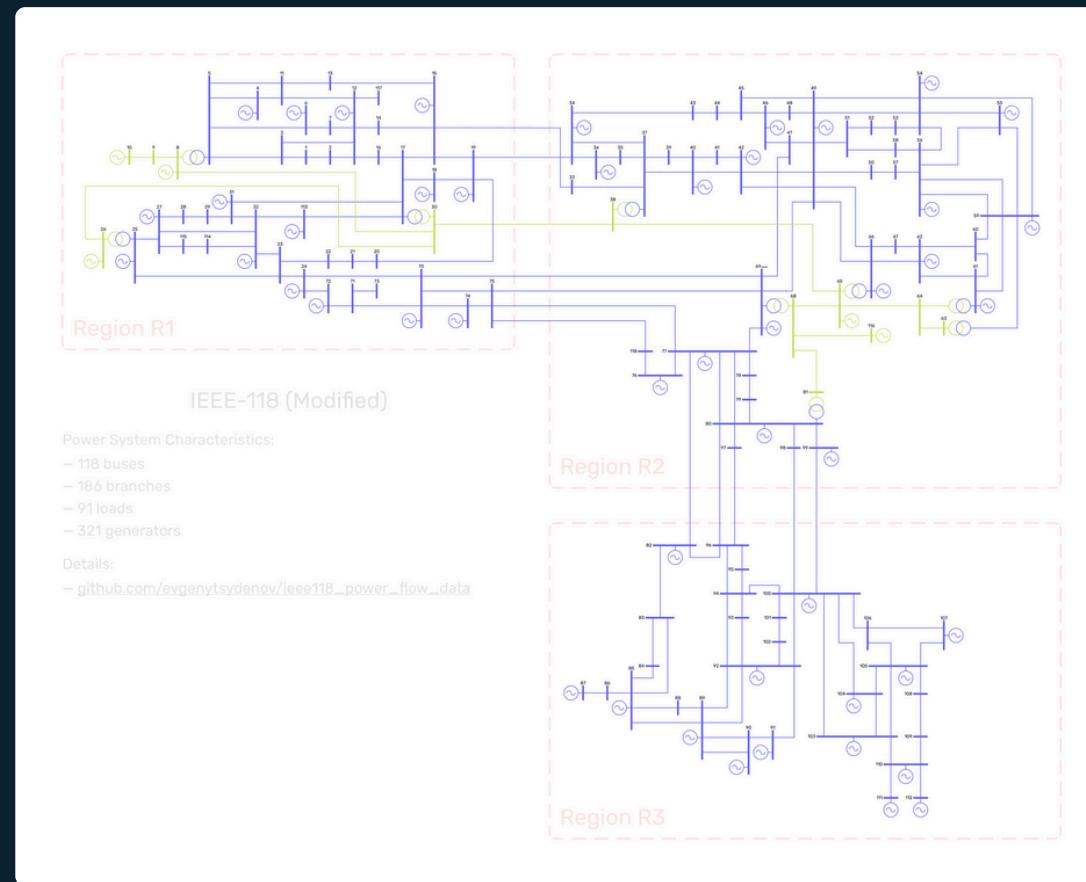


Using transformer models like in Chat-GPT, we analyze time-series data to deliver highly accurate forecasts

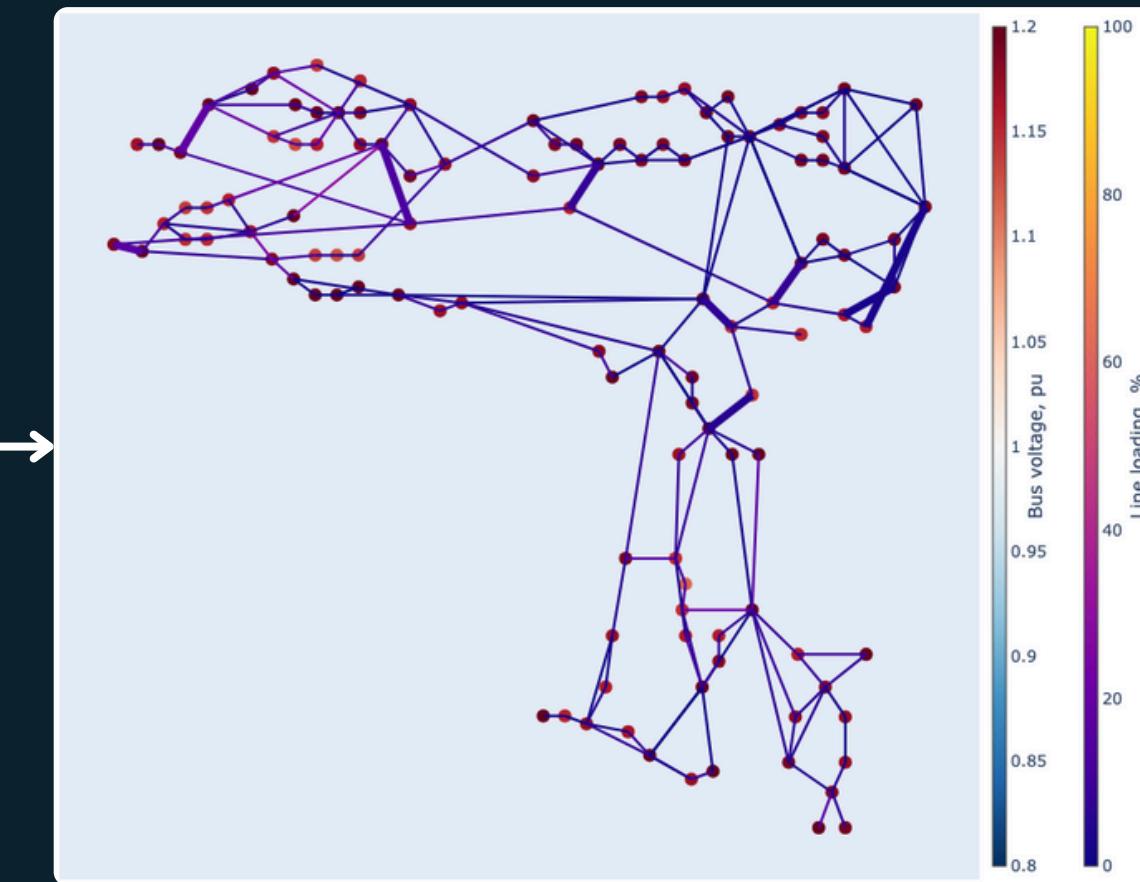


Our GNN architecture gives operators a real-time map of congestion risks and optimal dispatch decisions

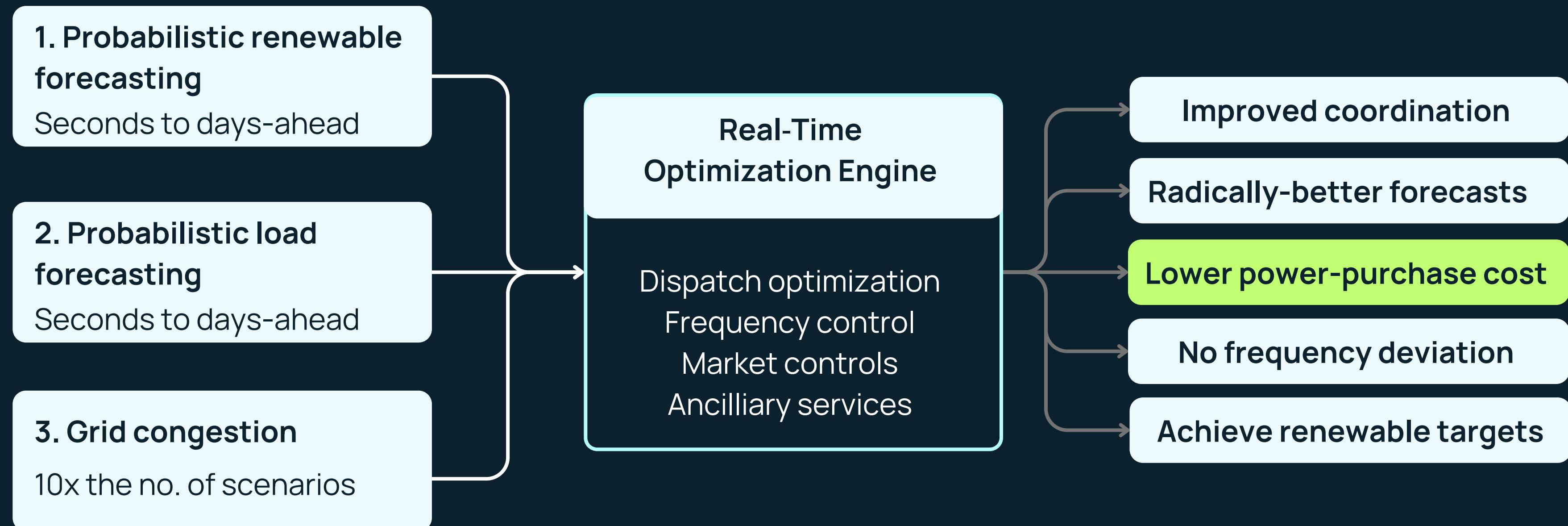
We model the grid as a graph:
Nodes = substations & transformers
Edges = power lines



Our engine learns from real-time SCADA data and historical records to predict overload risks 24 hours ahead.



Our real-time optimisation engine reduces your power procurement costs



You can read our detailed technical paper for more details

Matrix: AI-Powered Grid Operations for Electric Utilities

Dhruv Suri¹ and Mohak Mangal²
¹Department of Energy Science & Engineering, Stanford University, ²Graduate School of Business, Stanford University

The power grid is undergoing an unprecedented transformation driven by decentralization, renewable energy integration, and increasing system uncertainty. Traditional deterministic grid management approaches struggle to handle the stochastic nature of modern power systems, leading to forecasting errors, congestion mismanagement, and operational inefficiencies. To address these challenges, we present an AI-powered grid operations platform that leverages probabilistic forecasting, reinforcement learning-based congestion management, and agentic AI for autonomous decision-making.

Our framework enables real-time grid orchestration by integrating multi-source data from SCADA, AMI, PMUs, and high-resolution weather models, allowing for sub-second decision-making and adaptive grid control. The system is designed as a modular, scalable solution, enabling utilities to deploy AI-driven optimization at any level—from localized feeders to entire transmission networks. This approach enhances grid reliability, reduces operational costs, and maximizes renewable energy utilization by shifting from reactive, rule-based operations to predictive, self-learning AI systems. Success is measured through forecast accuracy improvements, uncertainty reduction, and economic efficiency gains over existing legacy models. This platform represents a transformational shift in grid operations, moving beyond static contingency planning toward a real-time, self-optimizing, and resilient power system.

Contents

1 Executive Summary	2
2 Industry Challenges & Need for AI-Driven Grid Operations	3
2.1 Overview	3
2.2 Limitations of Conventional Approaches	4
2.3 The Need for a Probabilistic AI-Based Approach	5
3 Technical Architecture	6
3.1 Multi-Timescale Probabilistic Forecasting	7
3.2 Dynamic Network Congestion Management	9
3.3 Weather-Integrated Supply Forecasting	11
3.4 DER and Grid Orchestration Engine	12
3.5 Agentic AI for Autonomous Grid Operations	13
3.6 Data Sources and System Integration	15
4 Pilot Proposal: Scalable Deployment and Demonstration	17

Please click here



Interested in a
collaboration?
Get in touch.

Mohak Mangal
mohakm@stanford.edu

Dhruv Suri
surid@stanford.edu

