



eomer: Enterprise Forecasting at Scale

Forecasting as a Service for time-series data - without ML infrastructure

eomer.ai | February 2026

Photo: © Shutterstock

Why enterprise forecasting fails today

Costly compute infrastructure as well as lack of in-house competence block adoption

Key Obstacles

70%

Infrastructure Overhead

\$500k+

Typical Setup Cost

1 year

Time-to-Value for In-House

Concrete Challenges for Enterprises

ML Expertise Gap



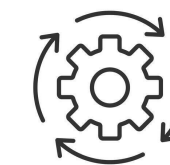
- Accurate forecasting requires specialised ML teams
- Hiring, ramp-up, and continuous research take 6-12 months

Infrastructure Investment



- GPU infrastructure requires \$50K–\$500K+ upfront investment
- Cloud and on-prem setups lead to unpredictable, escalating costs

Ongoing Model Maintenance



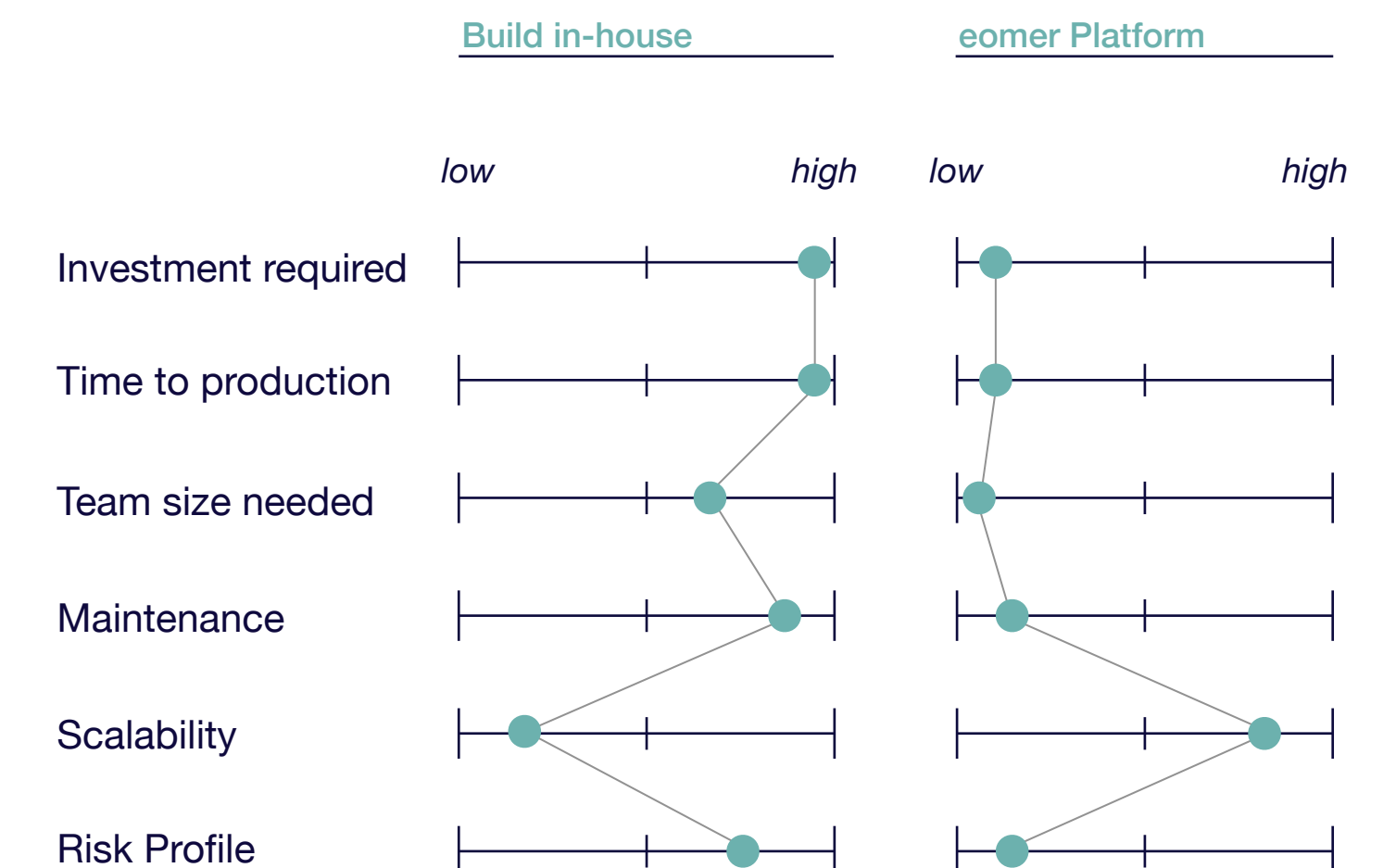
- Models degrade over time and require continuous retraining
- Overhead through monitoring, versioning, and deployment

Static Resource Allocation



- Fixed infrastructure causes 40–60% idle capacity off-peak
- Manual scaling leads to missed SLAs during demand spikes

Business Consequences: In-house vs. eomer



Benefits at a glance

10x

Deployment Speed

90%

Cheaper

Low

Risk

Enterprise Forecasting, Delivered as a Service

You provide data - we deliver production-ready forecasts for your use case

Benefits of the eomer platform

Cost Effective

1



- 90% cost reduction
- Pay only for what you use
- Predictable opex models

Time to Value

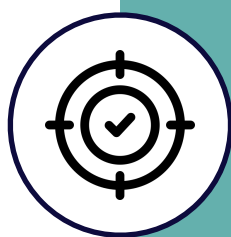
2



- Live in days
- Zero infrastructure setup
- Immediate ROI

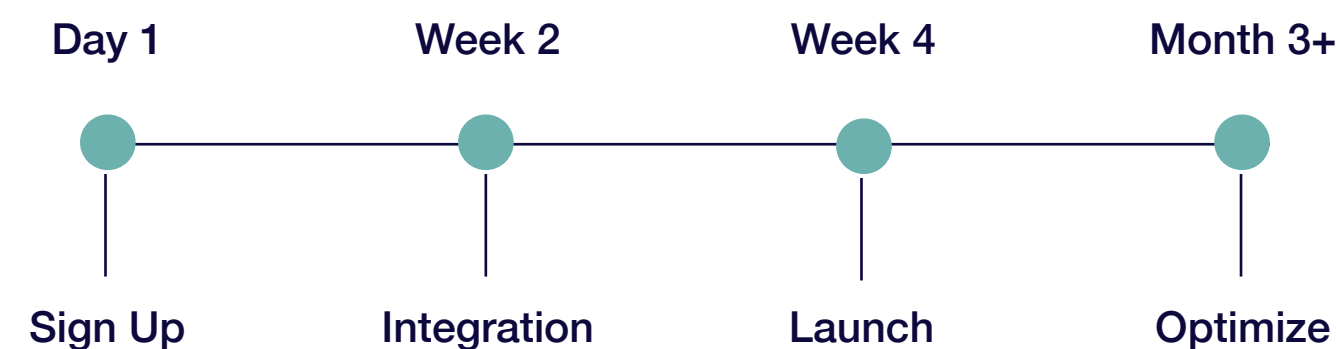
Scale & Accuracy

3



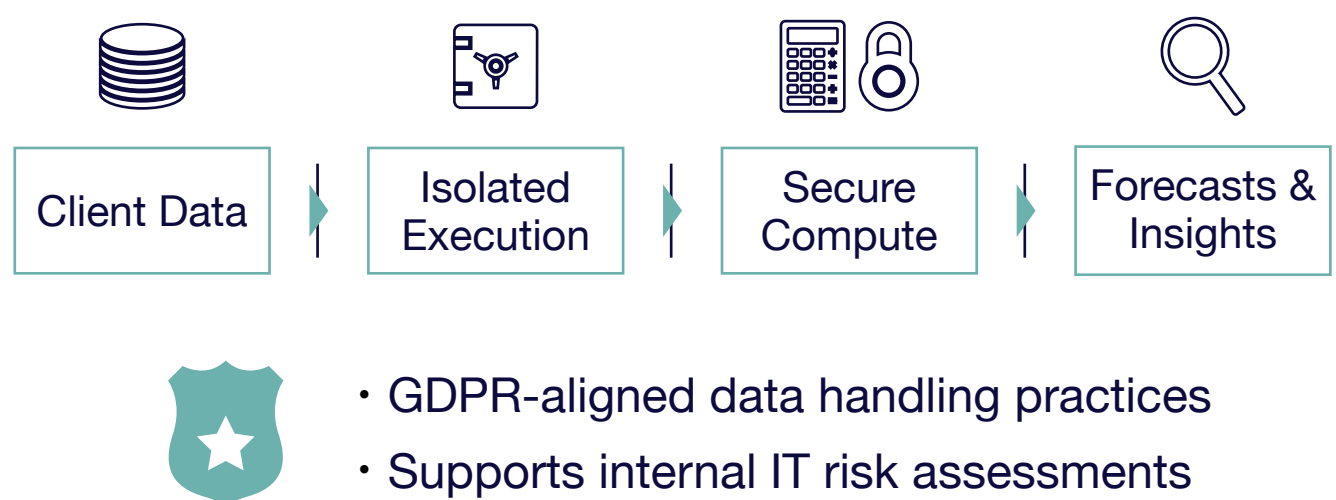
- Scale up to 100+ GPUs
- Pay-as-you-use structure
- High accuracy and reliability

Faster, Better, Cheaper



“ *Start forecasting in days, not months.
Scale effortlessly.
Focus on insights, not infrastructure.* ”

Your data in safe hands



Industry Applications

Supply Chain & Logistics



- Demand forecasting across network
- Transportation capacity planning

- Reduced logistics costs
- Improved on-time delivery

Retail & E-Commerce



- Product demand forecasting
- Dynamic pricing recommendations

- Reduced stockout and overstock
- Improved customer satisfaction

Applications Key Benefits

Foundation models: Proven accuracy across real-world use cases

Benchmarked, deployed, and verified in multiple use cases

Corporación Favorita: Ecuadorian grocery retailer¹⁾

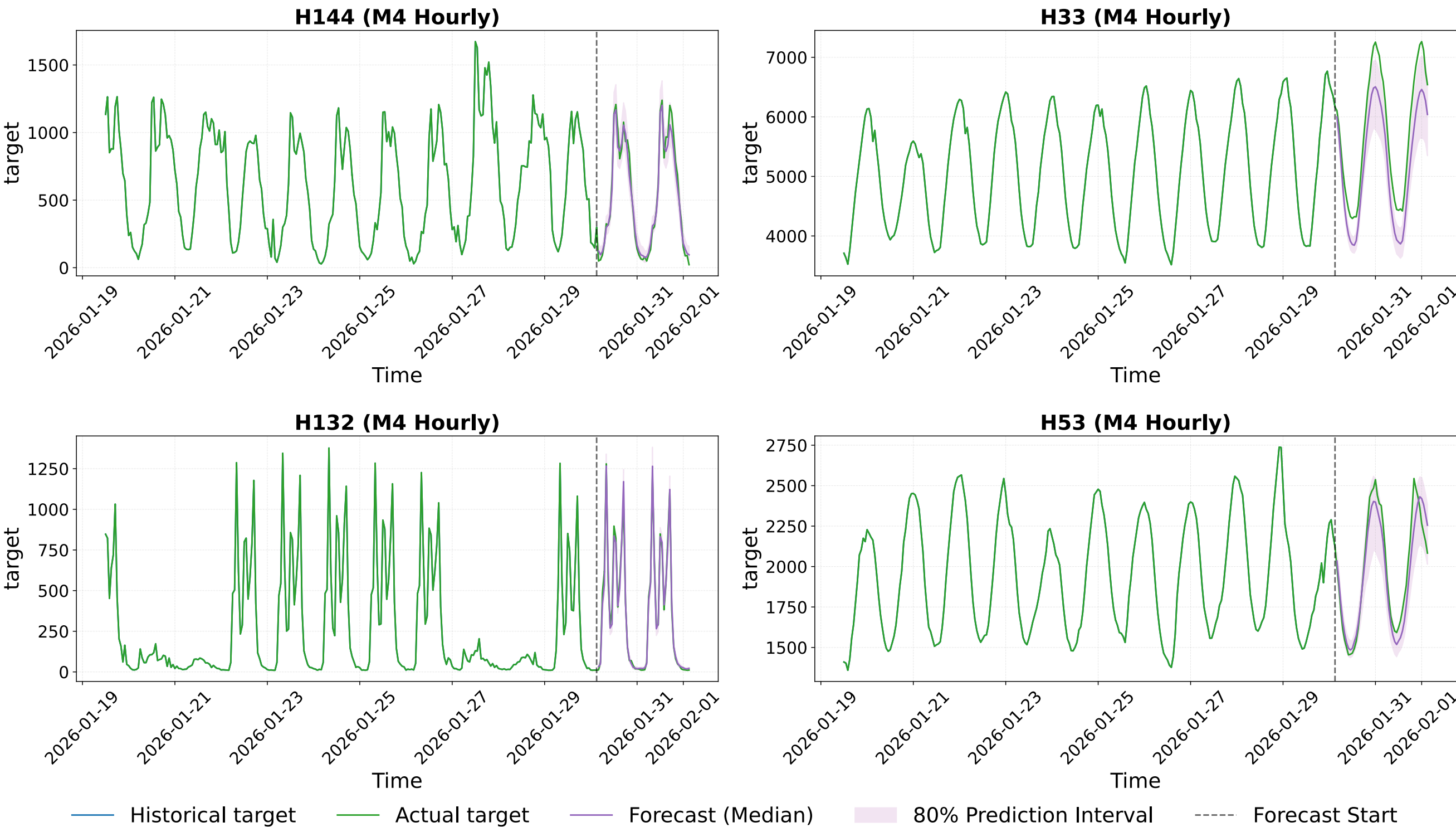
Top 3%
accuracy of all contestants

353,000
data points processed

~10 min
fine-tuning on personal laptop (CPU)

<10 sec
inference time

Great accuracy in the M4 Time Series benchmark dataset: Widely used in academia & data science



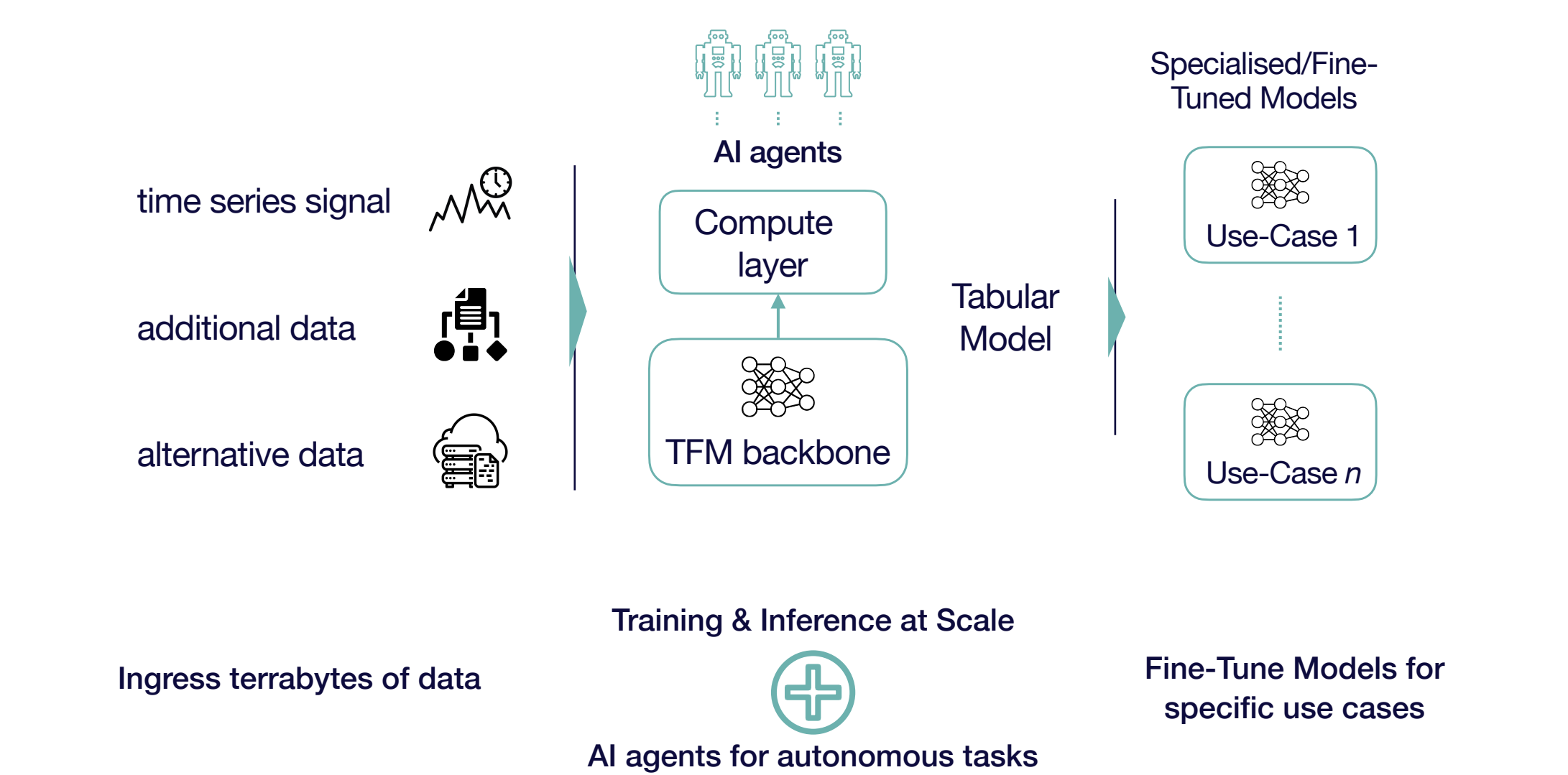
Note 1) International Time-Series Forecasting Competition ([Link](#))

Enterprise-grade Time Series Forecasting Service at Scale | Singapore & Munich | Reach us at: <https://eomer.ai>, contact@eomer.ai

Faster insights. Lower cost. Better decisions.

Up to 10x faster deployment and ~90% cost reduction versus in-house solutions

Solution Architecture



Fast training and inference pipelines

1

- Rapid fine-tuning with minimal data and compute
- Low-latency predictions suitable for real-time systems
- Unified pipeline without custom feature engineering

Exceptional prediction quality over multiple horizons

2

- State-of-the-art accuracy across short- & long-term horizons
- Robust generalisation from large-scale pre-training
- Consistent performance across diverse domains & regimes

Efficient scaling infrastructure for all data sizes

3

- Seamless scaling from small datasets to enterprise-scale data
- Compute-efficient architecture enabling cost-effective deployment
- Distributed-ready design for cloud and multi-node environments

Trusted Technology

<60 sec

cluster spin-up time

100+

GPU, TPU & CPU nodes

No

Manual intervention

Fast implementation backed by global presence & expertise

We serve our clients across the globe enabling them to generate value from day 1 with eomer platform

Strong coverage across EMEA & APAC served by our offices in Munich (GER) and Singapore (SGP)



Thomas Kopfmüller

 Munich

- Portfolio Lead EMEA at Imubit with >7 YOE across AI implementation in process industries
- MSc Chemical Engineering from TU Munich with research stay at MIT and MSc Business Administration & Economics







Lukas Voss

 Singapore

- Machine Learning (AI) experience from QRT & BlackRock, Fraunhofer and business strategy from Roland Berger
- Research stays at NUS and NTU Taiwan on Reinforcement Learning and High-Performance Computing (HPC)

From setup to launch in a few days yields fast ROI

	Step	Timeline
1	 Setup <ul style="list-style-type: none">• Create account & get API key• Configure security/authentication settings• Review documentation	Day 1
2	 Integration <ul style="list-style-type: none">• Connect data sources• Install SDKs, and configure data pipelines• Map data scheme	Days 2-5
3	 Test <ul style="list-style-type: none">• Run test predictions• Validate accuracy metrics & load testing• Performance benchmarking & fine-tuning	Days 6-10
4	 Launch <ul style="list-style-type: none">• Deployment to production• Enable monitoring & alerts• Configure scaling policies	Days 11-14

You can focus on insights - we handle all the rest.

- Forecasting as a Service
- Enterprise-grade scale & reliability
- No ML infrastructure or ML experience required

Let's discuss your business use case.

contact@eomer.ai

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