# Smart Manufacturing: Where AI Meets Industrial Impact



The industrial sector is responsible for 1/5 of global emissions and consumes 54% of delivered energy, positioning smart manufacturing as one of the most powerful levers for climate action. A new generation of AI-powered software tools is unlocking the ability to optimize energy and material flows at scale, making climate impact and industrial productivity mutually reinforcing.



#### Thesis

At AENU, we back startups applying next-gen digital technologies – from advanced IIoT to AI agents – to cut energy and material waste in CO<sub>2</sub>-intensive industries. We look for solutions focused on process optimization, energy efficiency, or predictive maintenance, and that succeed by getting data integration, user adoption, and scalable execution right.

#### Why It Matters

Decarbonizing manufacturing is one of the biggest levers to reach climate goals:

54%

of global delivered energy use (FIA)

The Massive Energy Footprint
The industrial sector uses more delivered energy than any other end-use sector, consuming about 54% of the world's total delivered energy.

1/5th

of global carbon emissions (<u>World Bank</u>)

The High-Emissions Problem

Manufacturing and construction are responsible for one fifth of global emissions. They are key leverage points for climate impact through decarbonization and efficiency innovation.

1% vs. 4%

current efficiency progress vs. target (IEA / COP28)

The Efficiency Gap

Global energy efficiency improvements have slowed to just 1% per year, while the COP28 pledge calls for over 4% annually until 2030. A fourfold acceleration is needed to stay on track for 1.5°C.

### The Opportunity

From Data to Impact

Smart manufacturing stacks now enable the full loop: data capture → insight → action.

With measurable energy and material gains at every step.

ΑI

Agents

Autonomous systems that perceive, plan, and act, continuously optimizing energy-intensive processes like heating, cooling, or production flows for real-time CO<sub>2</sub> reduction.

AI models that learn from production data to minimize waste, reduce downtime, and optimize efficiency, cutting unnecessary energy and material use.

Holistic Data Models Unified factory models that power process optimization, scenario simulation, and chatbased interfaces, enabling smarter, faster decisions for energy and material efficiency.

Operational Machine Learning

Data Operations

Prepares and connects data across systems: the backbone for reliable, sustainability-driven AI.

## The Challenge

### The Gap

Manufacturing companies want to

implement AI

68% Have already **started** implementing at

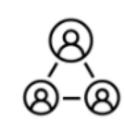
least one AI use case

Of AI-adopters have **achieved** their AI-related targets (WEF & BCG)

### The Bottlenecks

**#1 Challenge: Data Infrastructure**Scaling smart manufacturing solutions is held back most by data infrastructure

(<u>McKinsey</u>)



People & Processes = 70% of the Work
As tech advances, people, processes,
and culture remain a key barrier to
scaling AI in manufacturing. (BCG)

### What AENU Looks For

### Data-defensible solutions

with flywheels or contextual layers that make AI sticky and scalabe

### Plug-and-play architecture

Fast ROI without deep consulting or integration

### Impact Focus & Next-Gen AI

Solutions that drive measurable CO<sub>2</sub>e and energy savings via process optimization or predictive maintenance, powered by nextgen AI and digital technologies.

### Operator-first design

trusted, intuitive systems for industrial users

### Vertical focus, horizontal reach

beachhead in one industry, platform potential in others

Startups To Watch

Aris Machina







Intelecy

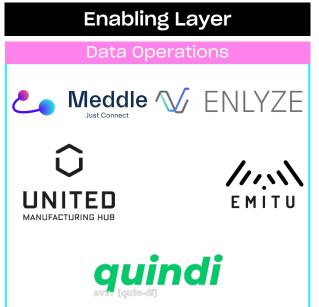




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**Operational Machine Learning** 







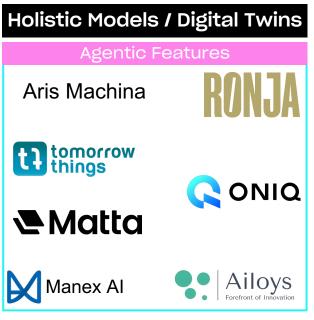
IPercept



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**OPTIMITIVE** 







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<sup>\*</sup>Includes only companies with European headquarters, at pre-seed to Series A stage, offering solutions with impact potential.