



# Tech Hot Topics: 2026 Edition

## Capabilities built in 2025 define the ability to scale today

In our 2025 Hot Topics report, **we mapped the capabilities that organizations needed to explore to remain resilient** in an environment marked by technological acceleration and growing uncertainty.

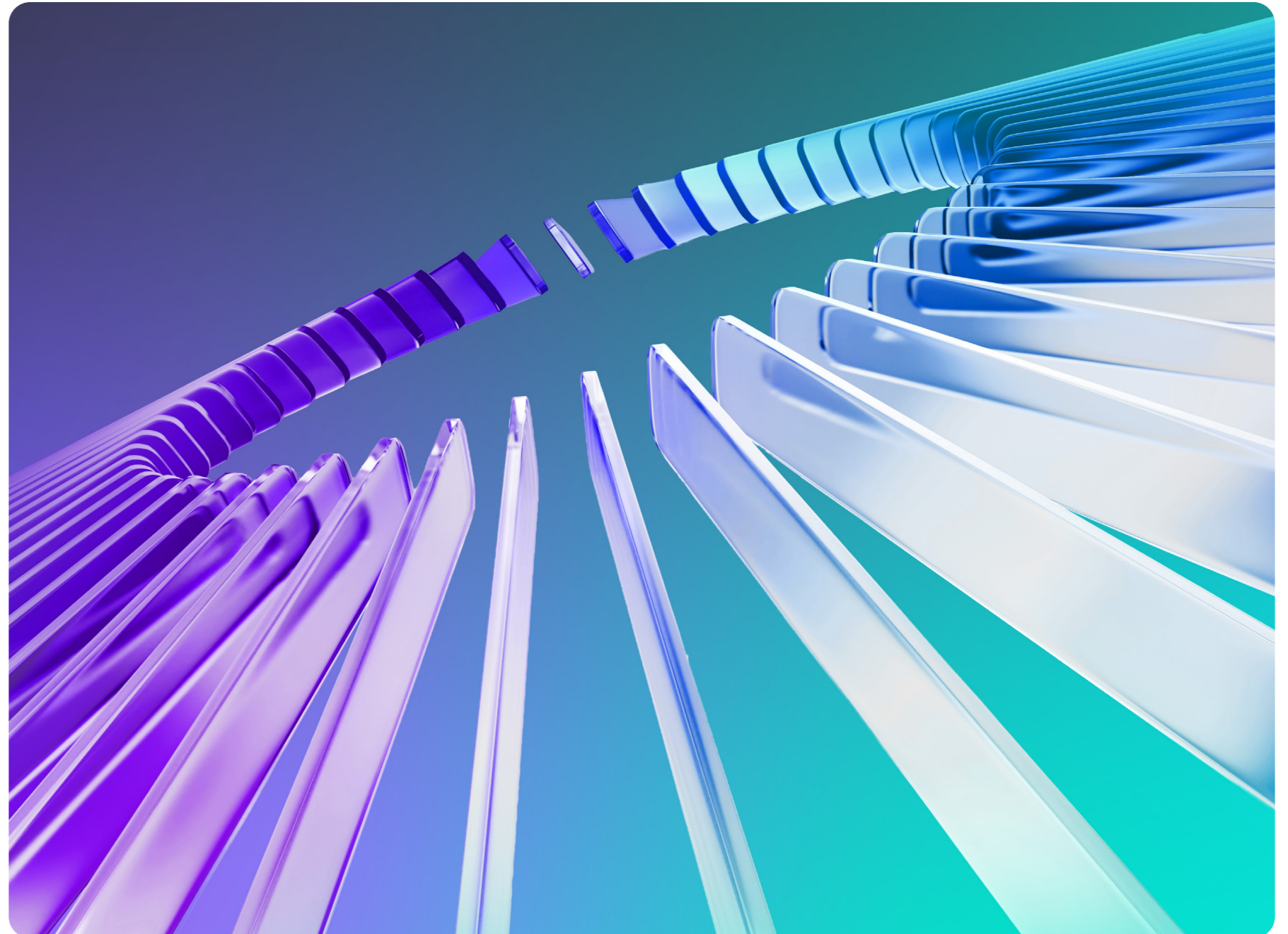
Twelve months later, **the context has changed substantially**. Not because new disruptions have necessarily emerged, but because many of the identified capabilities have moved beyond the exploratory stage to become the foundation upon which organizations are building their evolution.

This shift introduces a distinct dynamic: value no longer lies in identifying what to adopt, but in having previously **built the foundations that enable absorbing, integrating, and scaling** new capabilities without friction.










In this regard, organizations that made progress in 2025 have not only reduced their risk exposure but **have also created the conditions necessary to adapt more quickly** to an environment that demands execution, not exploration.

Those that did not now face a more structural gap, where the challenge is not incorporating new capabilities, but recovering the time not invested.

This report starts from that premise: understanding the 2026 Hot Topics not as a breaking point, but as the natural **continuation of a maturation process where prior preparation directly conditions future response capacity**.



# Softtek Tech Hot Topics: 2025 vs. 2026

	2025	2026
 <b>Artificial Intelligence</b>	<ul style="list-style-type: none"> <li>• AGI</li> <li>• AGENTIC AI</li> </ul>	<ul style="list-style-type: none"> <li>• SLMS Y EDGE AI</li> <li>• OPEN-SOURCE AI</li> </ul>
 <b>Data &amp; Software</b>	<ul style="list-style-type: none"> <li>• AUGMENTED ANALYTICS</li> <li>• REAL-TIME STREAMING</li> </ul>	<ul style="list-style-type: none"> <li>• ORCHESTRATED AGENT ECONOMIES</li> <li>• MULTIMODAL INTELLIGENCE</li> <li>• PHYSICAL AI</li> </ul>
 <b>Infrastructure</b>	<ul style="list-style-type: none"> <li>• AUGMENTED DEVELOPMENT</li> <li>• SELF-HEALING APP LAYERS</li> </ul>	<ul style="list-style-type: none"> <li>• DSLMS</li> <li>• AUTONOMOUS DIGITAL TWINS</li> </ul>
 <b>Customer Experience</b>	<ul style="list-style-type: none"> <li>• LOW-LATENCY CLOUD</li> <li>• MULTI-CLOUD STRATEGIES</li> </ul>	<ul style="list-style-type: none"> <li>• IDD ENGINEERING</li> <li>• DIGITAL PROVENANCE</li> </ul>
 <b>Design</b>	<ul style="list-style-type: none"> <li>• CONVERSACIONAL AI</li> <li>• EMOCIONAL AI</li> </ul>	<ul style="list-style-type: none"> <li>• INFERENCE ECONOMICS</li> <li>• NEXT-GEN HYBRID COMPUTING</li> </ul>
 <b>Business Innovation</b>	<ul style="list-style-type: none"> <li>• PRIVACY BY DESIGN</li> <li>• GENERATIVE INTERACTION BLUEPRINTS</li> </ul>	<ul style="list-style-type: none"> <li>• INTELLIGENT CX</li> <li>• AGENTIC COMMERCE</li> </ul>
 <b>Cybersecurity</b>	<ul style="list-style-type: none"> <li>• MICRO-INTERACTION DESIGN</li> </ul>	<ul style="list-style-type: none"> <li>• AI-AUGMENTED DESIGN SYSTEMS</li> <li>• TRUST-CENTRIC EXPERIENCE</li> </ul>
 <b>Sustainability</b>	<ul style="list-style-type: none"> <li>• DIGITAL BUSINESS TWINS</li> <li>• CORPORATE VENTURE BUILDING</li> </ul>	<ul style="list-style-type: none"> <li>• AI-NATIVE TRANSFORMATION</li> <li>• AUTONOMOUS SUPPLY NETWORKS</li> </ul>
 <b>Workforce &amp; Talent</b>	<ul style="list-style-type: none"> <li>• POST-QUANTUM CRYPTOGRAPHY</li> <li>• PETS</li> </ul>	<ul style="list-style-type: none"> <li>• PREEMPTIVE CYBERSECURITY</li> <li>• NHIS &amp; ZERO TRUST</li> </ul>
	<ul style="list-style-type: none"> <li>• ENERGY MANAGEMENT AUTOMATION</li> <li>• GREEN CODE</li> </ul>	<ul style="list-style-type: none"> <li>• GREEN AI &amp; COMPUTING</li> <li>• SUSTAINABILITY-AS-A-SERVICE</li> </ul>
		<ul style="list-style-type: none"> <li>• FULL-JOURNEY PROCESS AUTOMATION</li> <li>• AI-AUGMENTED WORKFORCE</li> </ul>

# Global volatility is no longer cyclical, organizations must design for continuous disruption

The global landscape in 2026 is being shaped by a convergence of self-reinforcing tensions, **reconfiguring the strategic and technological priorities of organizations. What emerges from this situation is the industry's direct response to an environment that fragments, hardens, and simultaneously accelerates.**

Uncertainty translates into concrete operational tensions; from disruptions, cost overruns, or loss of control to limitations in value capture, which are forcing organizations to mobilize and redesign both their operating model and their technology decisions. In response, **companies are prioritizing a new generation of capabilities, “hot topics”, designed to reduce exposure, strengthen control, and enable efficient scaling.**

## Convergence of global conflicts on multiple fronts



### Geopolitical and commercial fragmentation

Tensions, tariffs, and export controls are **reshaping supply chains, forcing entities to design for resilience**



### Regulatory pressure and compliance

Different regulatory frameworks are making the environment more stringent and complex, turning **explainability and control into operational requirements**



### Computational cost, energy, and sustainability

Scaling AI is an economic and environmental challenge. Companies are forced to **optimize and redesign models and align with ESG**



### The gap between adoption and impact

Accelerated adoption over inefficient processes is destroying ROI. **The focus shifts toward redesigning internal operations**



### Redefinition of cross-cutting relationships

Brands lose control of the entry point and teams are resized. **The advantage shifts to orchestrating human and digital talent**

# Hot Topics identify capabilities required to respond to structural pressures in 2026



## The 25 Hot Topics and the 9 categories

**Hot Topics** are **technological and organizational capabilities that companies are prioritizing** because they respond to environmental tensions. They are characterized by: **arising from real pressure** (geopolitical, regulatory, economic, operational, or social), **offering a concrete path to address it, and the cost of inaction** is already greater than the cost of integration.



Artificial Intelligence



Design



Data & Software



Business Innovation



It Infra-Structure & Cloud



Cybersecurity



Customer Experience



Sustainability



Workforce & Talent

## Key evaluation variables

The **25 main hot topics** are evaluated across the following **5 variables**. From this analysis, **an average rating is obtained, reflecting the strategic relevance and potential** impact of each trend in the industry, facilitating informed decision-making.

### Popularity

Measures the recognition and adoption of the trend in the industry and its relevance across various sectors.

**High (8-10):** Widely recognized with strong investment. Leading organizations already operate it at scale.

**Medium (5-7):** Gaining traction, but most implementations remain in pilot phase or limited to specific areas.

**Low (1-4):** Adoption concentrated in specialized sectors or in very early phases without significant operational deployment.

### Disruption

Assesses the trend's potential to transform existing systems, business models, and industries.

**High (8-10):** Causes fundamental changes and can displace current systems.

**Medium (5-7):** Represents significant innovations, but does not necessarily completely alter existing models.

**Low (1-4):** Represents incremental improvements without significantly altering sector dynamics.

### Immediacy of impact

Measures how quickly the trend generates real and measurable impact in the enterprise.

**High (8-10):** Immediate impact or within 1-3 years. The technology is already available and its integration produces short-term results.

**Medium (5-7):** Visible impact in 3-5 years, will have relevant effects in the medium term.

**Low (1-4):** Requires more than 5-8 years for mass adoption and maturity.

### Business value

Measures the potential to deliver economic benefits, competitive advantage, or tangible value creation.

**High (8-10):** Promises high return on investment and competitive advantages.

**Medium (5-7):** Has significant value, offering moderate returns with growth potential as it evolves.

**Low (1-4):** Its financial impact is uncertain or depends on external factors.

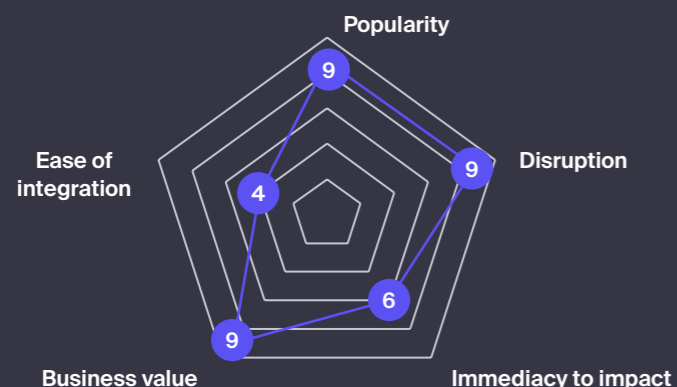
### Ease of integration

How easily an organization can integrate the trend into its current processes and structure.

**High (8-10):** Simple implementation with few technical barriers.

**Medium (5-7):** Requires some modifications to processes or infrastructure, but integration is feasible.

**Low (1-4):** Requires complex changes, specific infrastructure, or regulatory adaptation.



7,4

01

### Orchestrated Agent Economies

Deploy multi-agent systems to automate complex workflows and scale decision-making without human bottlenecks

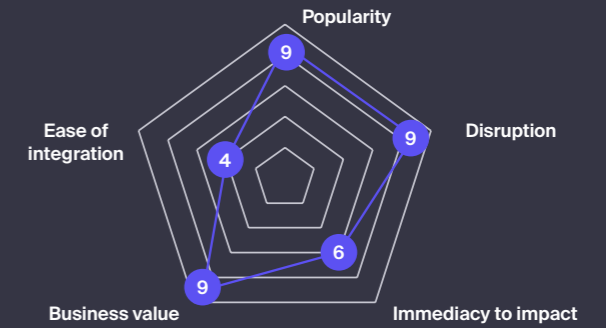
**Orchestrated Agent Economies** (multi-agent systems or MAS) represent an evolution of AI toward distributed architectures, where **multiple autonomous agents collaborate, share context, and coordinate decisions** to solve complex problems. Through direct communication or shared environments, these agents build dynamic models of objectives, memory, and action plans, integrating continuous learning to adapt and optimize outcomes.

#### Insight

The transition toward multi-agent systems marks the shift from point automation to autonomous orchestration of complex processes, enabling organizations to **scale decision-making and execution without direct human intervention**. Their ability to distribute intelligence, coordinate specialization, and dynamically **adapt makes them an enabler for environments where complexity, speed, and uncertainty surpass traditional models**, with direct applications in supply chain, financial operations, customer service, or intelligent physical systems.

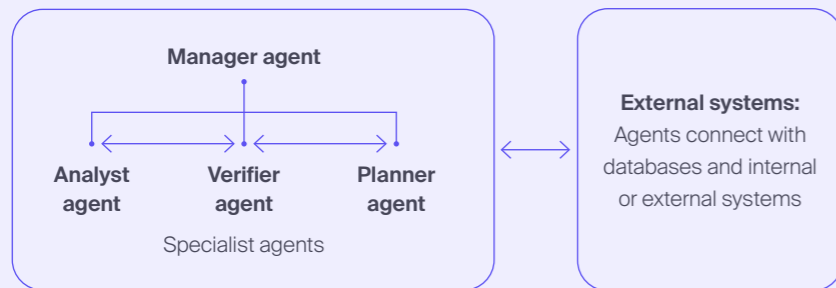
#### Trend prioritization matrix

7,4



### A manager agent coordinates specialized agents and external systems, progressing from context acquisition to distributed execution and continuous adaptation.

#### Agent ecosystem connection



- 5 **Continuous adaptation:** capturing information from the environment and building a state representation.
- 4 **Interpreting and aligning objectives:** analyzing information, adjusting objectives, and synchronizing them with the system.
- 3 **Coordinating and negotiating actions:** interacting to define action through direct coordination, negotiation, or adjustment.
- 2 **Execute in a distributed manner:** acting autonomously, working in parallel, and contributing to the global objective.
- 1 **Perceiving and contextualizing:** integrating feedback and adjusting behaviors through collective learning.

#### Use cases

##### Autonomous Supply Chain

Negotiate and plan flows in real time during disruptions, reducing dependence on centralized planning

##### Advanced Healthcare and Simulation

Model complex scenarios through interaction of multiple specialized agents

##### Intelligent Financial Operations

Monitor markets, execute decisions, and manage risks in a coordinated and continuous manner

##### Distributed Cybersecurity

Monitor, detect, and respond to threats in a coordinated manner, anticipating attacks at multiple points

##### Autonomous Customer Service

They collaborate to resolve incidents end-to-end without intervention, integrating context and multiple systems

02

### Physical AI

Deploy autonomous physical systems to operate in real environments and reduce manual intervention at scale

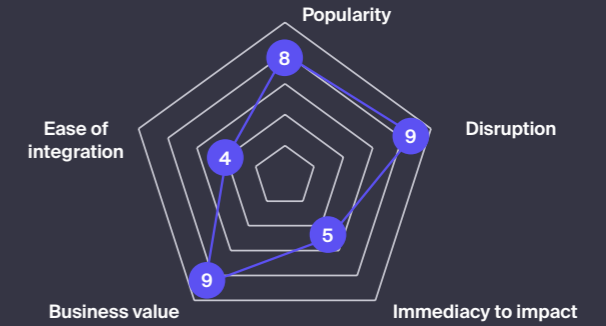
**Physical AI** represents the union of artificial intelligence with **systems capable of directly interacting with the physical environment**. Integrating advanced models (including vision and multimodality) with sensors, actuators, and control systems, these are capable of perceiving, reasoning, acting, and learning in real environments, transferring intelligence from software to physical operation.

#### Insight

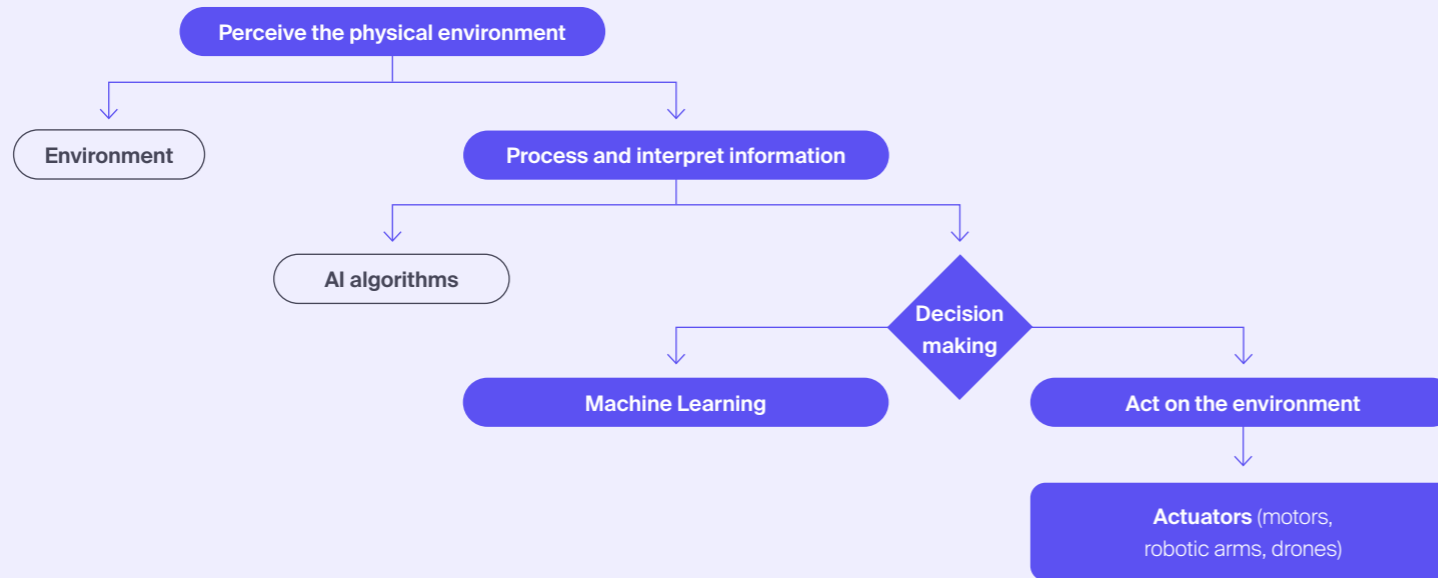
The convergence of foundational models, advanced simulation, computational capacity, and intelligent hardware is unlocking a **new generation of autonomous systems capable of operating in unstructured environments**. This marks the shift from rigid industrial automation to adaptive physical systems, capable of **generalizing tasks, learning in real time, and executing complex operations without explicit programming**, with direct impact on critical industries such as manufacturing, logistics, mobility, and healthcare.

#### Trend prioritization matrix

7,0



#### AI perceives, processes, and makes decisions to act in the physical world.



#### Cross-cutting levers for scaling

##### Focus on high-ROI use cases

Prioritize cases where tangible immediate economic impact is generated

##### Integration with existing systems

Connect physical systems with ERP, data, and operational processes

##### Real-time decision-making

Reduce latency through autonomous data-driven decisions

##### Workforce preparation

Adapt talent to human-machine collaboration and new roles

##### Distributed scaling

Replicate solutions across multiple locations while maintaining operational efficiency

##### Continuous ROI measurement

Monitor performance, costs, and generated value to optimize operations

03

### Domain-Specific Language Models (DSLMs)

Adopt domain-specific models to improve accuracy, reduce risk and enable AI in regulated workflows

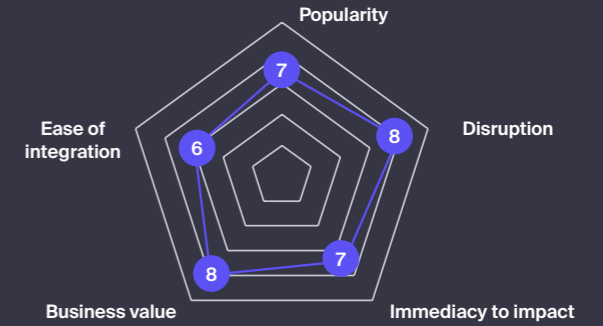
**Domain-Specific Language Models (DSLMs)** are AI models trained or fine-tuned with datasets specific to a particular industry, function, or process. Unlike generic LLMs trained on the entire internet, DSLMs **integrate knowledge directly into their parameters, enabling them to understand sector-specific terminology, regulatory context,** and domain-specific processes with greater relevance and precision through fine-tuning.

#### Insight

DSLMs represent the evolution from generalist AI to profitable vertical specialization. They offer surgical precision, integrated compliance, and economic efficiency in sectors where the margin of error is zero. Furthermore, they are a key architectural component of multi-agent systems, understood as agents powered by specialized models that execute more precise autonomous decisions in regulated contexts, enabling the transition from point automation to complex orchestration.

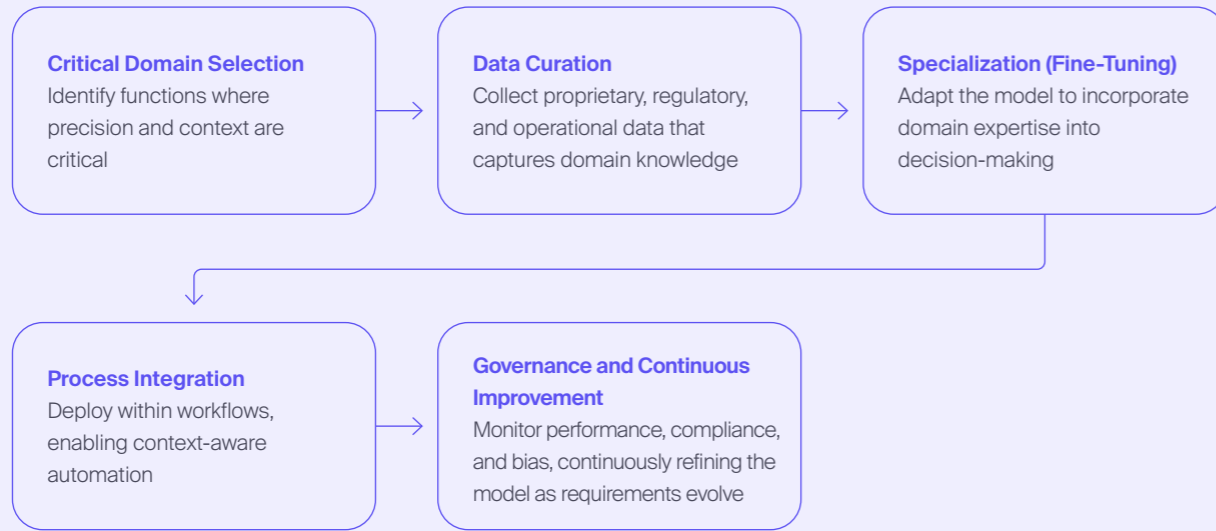
#### Trend prioritization matrix

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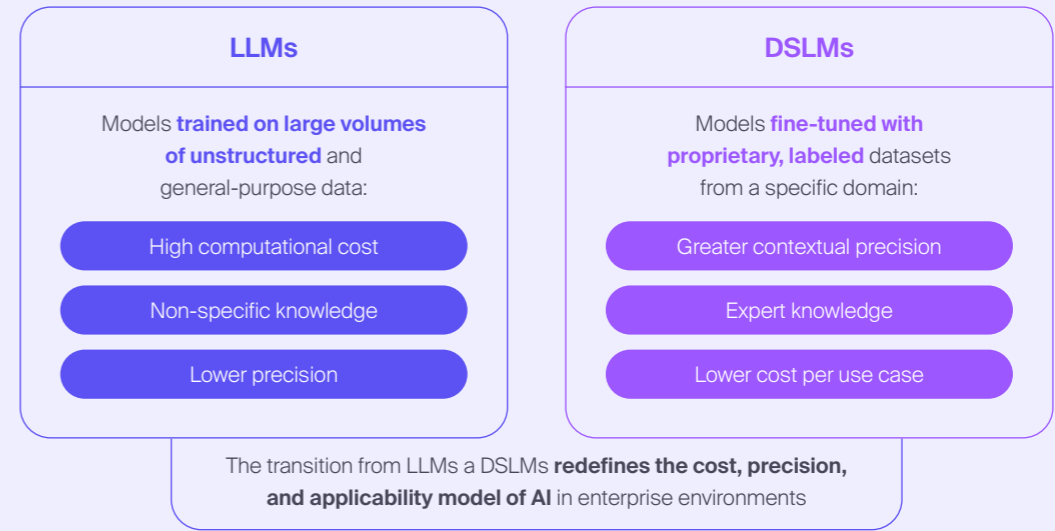


### From Generic Model to Expert Model

#### How DSLMs operate in enterprise



#### Fine-Tuning with domain data



04

## Unified multimodal intelligence

Integrate multimodal AI to improve decision accuracy and unlock richer, context-aware automation

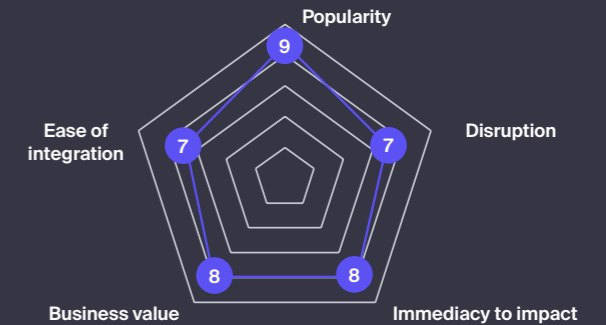
**Unified multimodal intelligence** refers to **models capable of processing, integrating, and generating information from multiple data types** (text, image, audio, or video) **within the same architecture**. Unlike unimodal LLMs, these systems capture relationships between different information streams to generate deeper contextual understanding and more precise decisions.

### Insight

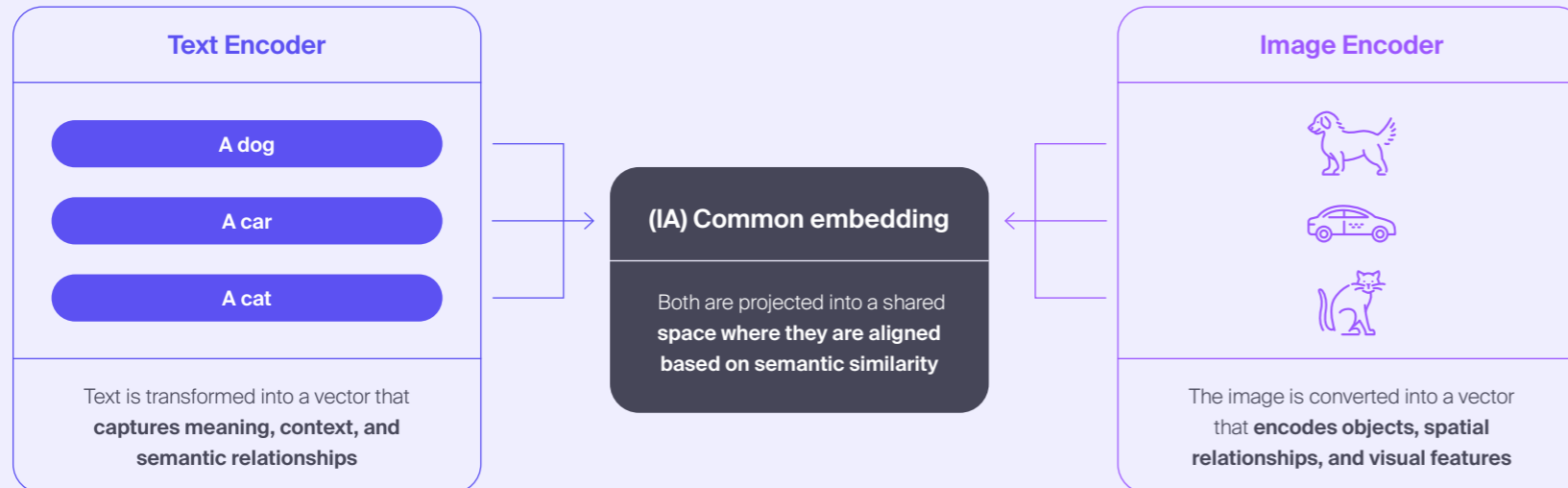
Multimodal AI transforms artificial intelligence from systems that interpret isolated inputs to **systems that comprehensively understand complex contexts**. This enables more precise **decisions, greater robustness against incomplete data, and more natural human interaction**. The benefit lies in the fusion of modalities, where heterogeneous **data is integrated into shared representation spaces through approaches like early, mid, or late fusion**, enabling diverse information to be related and contextual reasoning.

### Trend prioritization matrix

7,6



## Text and images are transformed into vectors and aligned within a shared semantic space



### Transformational impact cases

#### Healthcare

Combination of medical imaging, records, genomic data, and context to improve diagnosis, personalization, and efficiency

#### Transportation

Fusion of camera, LiDAR, radar, and map data to interpret complex environments and execute safe decisions in dynamic conditions

#### Retail & E-Commerce

Integration of visual search, in-store behavior, and transactional data to optimize discovery, recommendation and conversion

#### BFSI

Biometric and behavioral authentication, assistants capable of interpreting text, voice, and documents to improve CX and reduce fraud

#### Smart cities

Integration of camera data, IoT and external conditions to optimize traffic, infrastructure maintenance, and operational response

05

## Autonomous Digital Twin Systems

Use autonomous digital twins to simulate, predict and optimize operations in real time

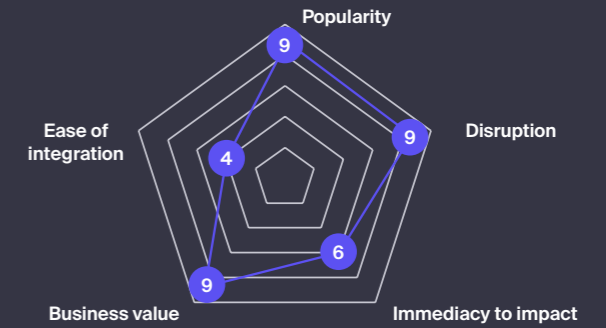
A digital twin is a virtual replica of a physical object, system, or process that uses real-time data to accurately reflect the behavior, performance, and conditions of its real counterpart. **Autonomous Digital Twin Systems integrate AI agents** that not only replicate and monitor, but simulate future scenarios, predict failures before they occur, and optimize parameters autonomously without direct human intervention (**closed-loop digital twins**).

### Insight

The evolution toward autonomous digital twins **transforms simulation into active operational capability**, where systems not only predict scenarios but **recommend or execute actions in real time**. This allows organizations to move from reactive management to continuous data-driven optimization where the twin analyzes, executes predictive simulations, makes decisions, and transmits adjustments to systems, **reducing risk, accelerating decision-making, and improving efficiency at scale**.

### Trend prioritization matrix

6,0



### Autonomous Simulation and Action Loop: the digital twin replicates, analyzes, and continuously adjusts the physical system using real-time feedback.

#### Simulation Engine

Integrates physical models and real-time data to replicate asset behavior and maintain continuous alignment between the physical and digital worlds



#### AI Analytics Engine

AI models formulate decisions, optimize parameters, and coordinate actions based on operational objectives and risk levels



#### Autonomous Decision Layer

Apply machine learning on historical and operational data to detect anomalies, generate diagnostics, and simulate future scenarios



#### Digital Twin Behavior

Translates decisions into actions on the physical system and generates a loop continuous learning loop through feedback in real time



### Use cases by industry



#### Energy & Power

Facilitates monitoring and optimization of distributed networks and assets, anticipating failures and adjusting consumption



#### Supply Chain & Logistics

It simulates end-to-end scenarios and adjusts logistics flows in real time in response to changes in demand, costs, or external constraints



#### Healthcare

Models patient or treatment evolution to optimize clinical decisions and reduce risk in high-uncertainty environments

06

## Intent-Driven Development (IDD)

Shift from coding to intent definition to accelerate development and reduce delivery cycles

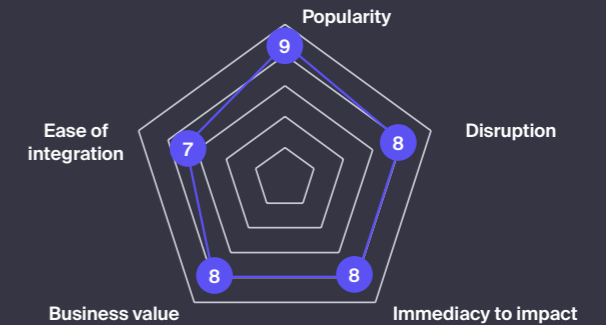
**Intent-Driven Development (IDD)** is a new methodology where humans define what should exist and why it matters, while autonomous agent systems determine how and when it gets built. Unlike previous methodologies where humans execute and processes coordinate human work, IDD **delegates full implementation to AI agents**: planning, coding, testing, iteration on failures, and deployment.

### Insight

The model redefines software development by **shifting value from implementation capability to specification quality**. In an environment where code can be generated automatically, **the differentiating factor shifts from development speed to the clarity with which the problem, objectives, and constraints are defined**. This transforms both how software is built and team structure, where the focus moves from technical execution to product thinking, continuous validation, and the ability to translate intent into executable systems.

### Trend prioritization matrix

8,0



### The flow translates objectives into hypotheses, autonomous execution, and continuous validation, while managing key risks

#### How IDD operates

- 1 Intent Mapping:**  
Objectives, hypotheses, constraints, and metrics.
- 2 Specify as hypotheses:**  
Testable hypotheses that enable validation of whether the system delivers value.
- 3 Autonomous execution:**  
Agents generate, test, and deploy code without direct intervention.
- 4 Measure and Validate in Use:**  
Monitoring and comparison with initial hypotheses.
- 5 Social Skills:**  
Intent adjustment based on results.

#### Key challenges or barriers

- 1 Limited Trust in AI Outputs:**  
The inconsistent quality of generated code reduces trust.
- 2 Vulnerability Risks:**  
Generated code may incorporate security weaknesses.
- 3 Development Lifecycle Instability:**  
Continuous code generation increases changes and instability.
- 4 Skills Gap:**  
The developer's role expands and evolves toward agent supervision and orchestration.
- 5 Input quality dependency:**  
La precisión de la intención se convierte en el principal factor de calidad

### Relevant IDD use cases



#### Product development

Small teams define intent and agents execute full development, enabling product iteration in days instead of weeks



#### Modernization at scale

Multiagents refactor legacy applications and migrate architectures, reducing transformation timelines



#### Compliance-by-design

Agents develop complex systems while humans validate security and compliance, with controls from specification



#### Citizen Development

AI-powered platforms enable non-technical users to build complex applications and workflows from intent

07

## Digital Provenance

Implement traceability systems to ensure data authenticity and enable trust in AI-generated content

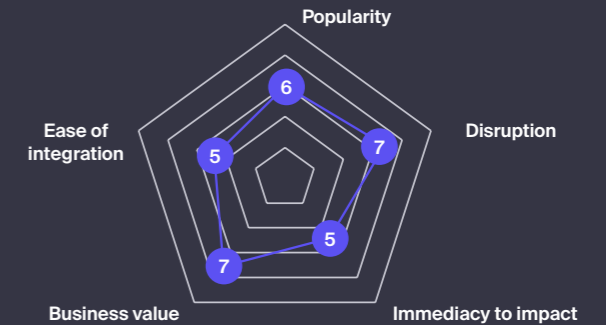
**Digital Provenance** is the **verifiable record of the origin, transformations, and chain of custody of a digital asset throughout its entire lifecycle**. Unlike traditional metadata or data lineage, it not only describes how information flows but provides an auditable history that identifies what data was generated, how it was modified, and who was responsible at each stage, ensuring authenticity, integrity, and accountability.

### Insight

As the massive generation of synthetic content and the proliferation of deepfakes erode trust in digital environments, organizations are evolving from reactive **verification models to approaches based on origin certification**, where authenticity and traceability are integrated from the moment of creation. This shift positions Digital Provenance as **a key capability for ensuring information integrity, enabling AI system governance, and ensuring compliance in a context** where data veracity becomes a critical requirement.

### Trend prioritization matrix

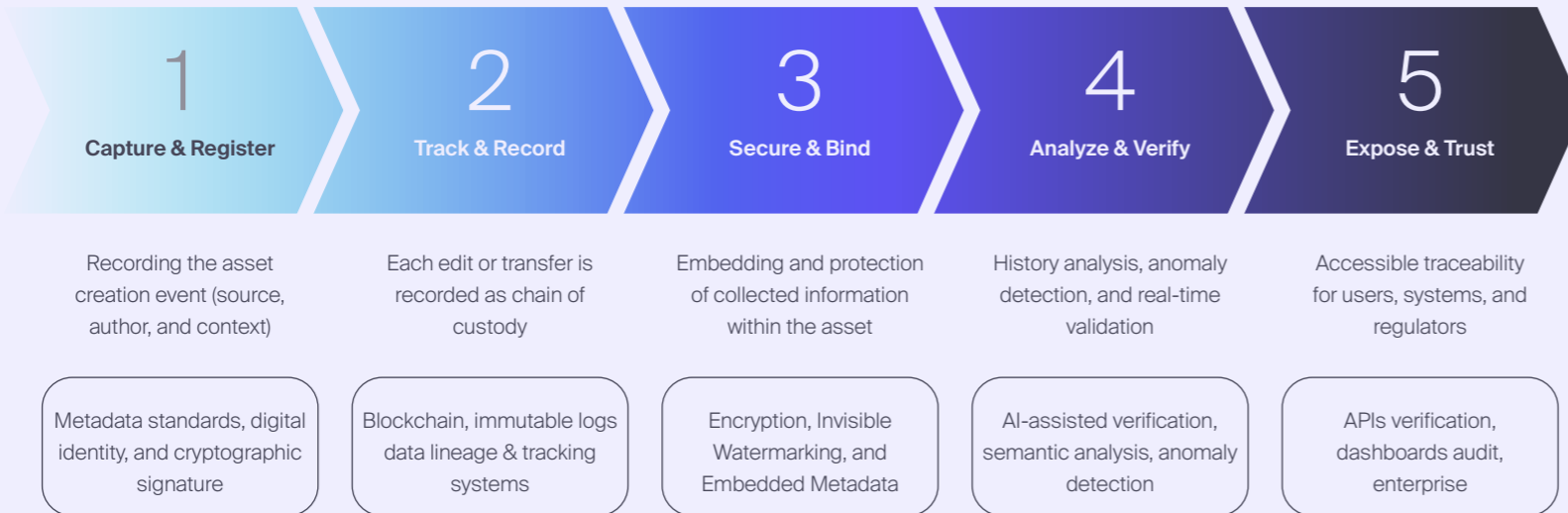
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## Digital Asset Traceability and Trust

### Operational flow

How traceability is ensured



### Key relevancies

Key benefits



08

## AI Inference Optimization

Optimize inference efficiency to reduce AI costs and enable scalable production deployment

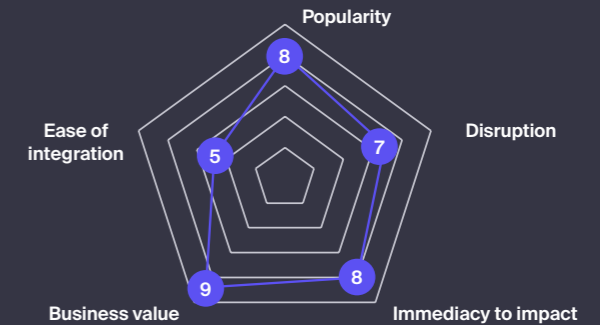
**AI Inference Optimization** refers to the **set of techniques and architectures designed to improve the performance, latency, and cost of AI models** in production. This approach integrates optimization at the infrastructure and model level, enabling organizations to scale intelligence services efficiently, sustainably, and adapted to each use case.

### Insight

While AI adoption at scale exposes the economic unsustainability of current architectures, organizations are evolving from approaches focused on adding compute capacity to **models oriented toward maximizing inference efficiency**. Rather than investing in more hardware, the focus shifts toward using existing infrastructure more efficiently through optimization at the model, runtime, and orchestration level. This shift **reduces cost per request, improves latency, and enables more responsive applications**, turning AI from an expensive burden into a scalable and sustainable production service.

### Trend prioritization matrix

7,4



## Optimize inference across the application, model, and execution layers to reduce cost and latency

### Inference optimization

#### Ai applications layer

Applications such as agents, copilots, or real-time systems **require continuous inference**, significantly increasing computational demand

#### Optimization layer

**Model efficiency**  
The model itself is optimized to make it lighter, faster, and more efficient.

**Execution efficiency**  
The way the model is executed is optimized to maximize performance and resource utilization.

#### Execution layer

**Inference is deployed across hybrid environments** (cloud, data center, or edge), **combining different hardware** architectures to balance cost, latency, and control.

### Key barriers and challenges

1

#### Technical complexity and talent shortage

Effective optimization requires deep knowledge, which limits adoption due to the shortage of profiles capable of operating these environments

2

#### High model dependency

Techniques must be adjusted to each model and context, hindering industrialization and requiring specific iterations

3

#### Performance vs. Precision trade-off

Reducing costs through aggressive optimization can degrade output quality, especially in sensitive sectors

4

#### Integration with existing systems

The transition toward runtimes and optimized architectures requires adapting or rebuilding legacy applications, increasing implementation effort



## Next-gen hybrid computing

Adopt hybrid compute models to balance cost, latency and control across workloads

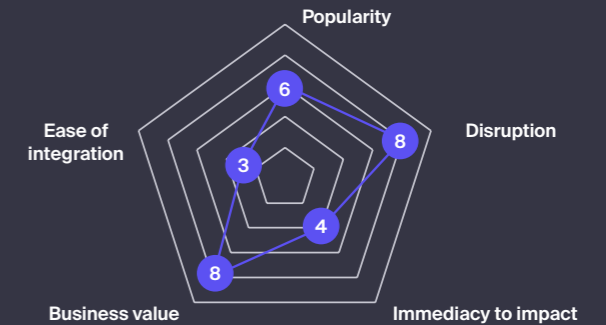
**Next-gen hybrid computing** combines different environments along with multiple types of specialized hardware, such as CPUs, GPUs, and AI accelerators, to execute complex workloads more efficiently. Unlike traditional models centered on a single infrastructure, **this approach distributes each task to the most suitable environment based on cost, latency, data control,** or performance requirements.

### Insight

AI is advancing toward increasingly complex and expensive models in terms of compute, energy, and operation, making traditional architectures based on a single environment or hardware type no longer viable. In this context, **balancing variables such as cost, energy availability, latency, data sovereignty, and compute capacity becomes a critical challenge,** driving the adoption of hybrid computing environments where **different infrastructures and technologies combine to execute each workload in the most cost-effective way.**

### Trend prioritization matrix

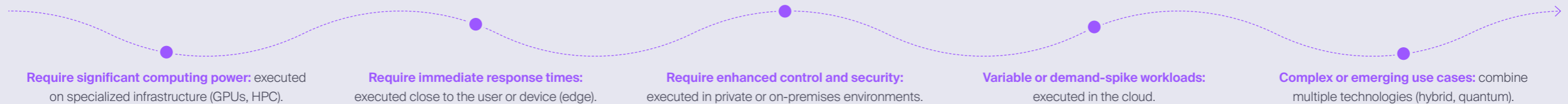
5,8



## Assign each workload to the most appropriate environment based on cost, latency, control, and performance



## Workload-to-compute logic by problem type



10

## Geopatiation

Localize data and workloads to reduce geopolitical risk and ensure regulatory compliance

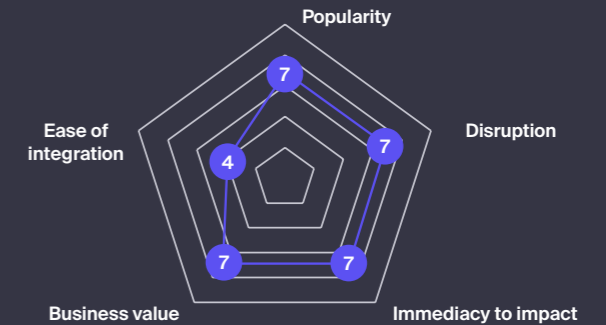
**Geopatiation** refers to the **relocation of workloads from global public clouds to local, regional, or sovereign infrastructures**, with the aim of maintaining control over data, systems, and their jurisdiction. This approach includes both migration to sovereign cloud regions and repatriation to on-premise infrastructure environments, as well as architecture redesign to limit cross-border data movement.

### Insight

The **growing geopolitical fragmentation and regulatory tightening are transforming cloud** infrastructure into a strategic risk decision, where operating in external jurisdictions implies loss of control over data, compliance, and response capacity. Therefore, organizations are prioritizing sovereignty and resilience over global cloud efficiency, **adopting models where critical workloads remain under local or regional control**. This shift redefines the cloud model of the last decade: it is no longer about centralizing for efficiency, but distributing to ensure control and operational continuity.

### Trend prioritization matrix

6,4



## Geopatiation vs. Cloud repatriation comparison

Aspect	Cloud repatriation	Geopatiation
<b>Key question</b>	Should we stay in the public cloud?	Where should this workload legally operate?
<b>Main trigger</b>	<ul style="list-style-type: none"> <li>• Cost and Performance</li> <li>• Architecture</li> <li>• Operational Control</li> </ul>	<ul style="list-style-type: none"> <li>• Geopolitical Instability</li> <li>• Jurisdictional Challenges</li> <li>• Exposure to Sanctions</li> </ul>
<b>Cloud usage after the change</b>	Often reduces or abandons public cloud usage	Can continue using the cloud, but within national or regional boundaries
<b>Scope of the change</b>	Redefinition of infrastructure strategy	Mitigation of jurisdictional and territorial risks
<b>Typical outcome</b>	Migration to private cloud or on-premise	Regional cloud, domestic infrastructure, or controlled environments
<b>Relationship</b>	Broad category	Specialized subset focused on location risk

### Why is it important now?



#### New legislation

Require storing or processing data within specific borders



#### New laws or agreements

Limit the use of services of a global or international nature



#### Cybersecurity

Requirements mandate keeping sensitive data at the local level



#### Trade conflicts

Restrict the use of certain providers in specific regions

11

## Autonomous Customer Experience

Automate end-to-end customer interactions to reduce resolution time and improve service scalability

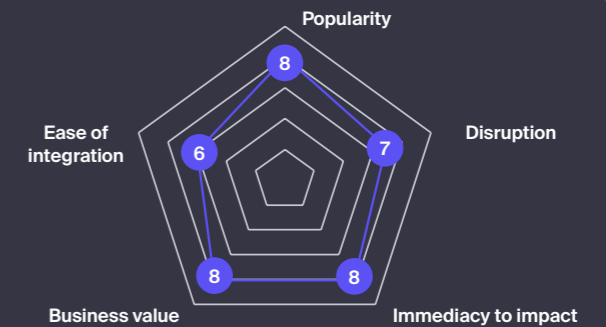
**Autonomous Customer Experience** refers to interaction models where AI-based systems autonomously manage the complete customer relationship cycle, **from understanding the need to its resolution, without direct human intervention.** This approach integrates natural language capabilities, decision-making, and continuous learning to manage complex interactions across multiple channels.

### Insight

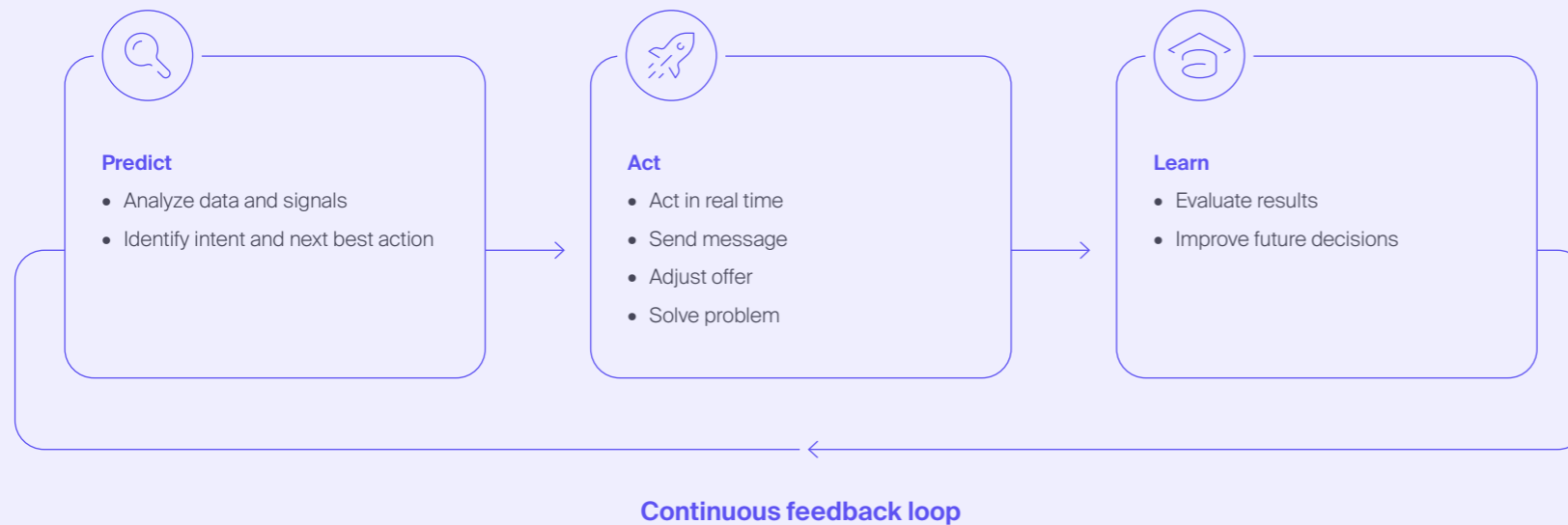
The adoption of Autonomous Customer Experience opens the door to transforming the customer relationship toward **more agile, personalized, and frictionless models, where interactions are resolved immediately and continuously** without depending on human availability. This approach enables organizations to significantly improve operational efficiency, reduce resolution times, and deliver more consistent experiences at scale, **while freeing teams to focus on higher-value interactions.**

### Trend prioritization matrix

7,4



### The flow anticipates customer needs, acts in real time, and learns from every interaction



### Some application examples

- Retail**  
Dynamic personalization based on real-time behavior
- E-Commerce**  
Proactive service and early incident resolution
- SaaS**  
Early identification of churn and proactive retention
- Financial services**  
AI to detect patterns and trigger personalized communication
- Travel**  
Incident management and real-time communication

12

## Agentic Commerce

Adapt to AI-driven discovery to capture demand and remain visible in agent-mediated journeys

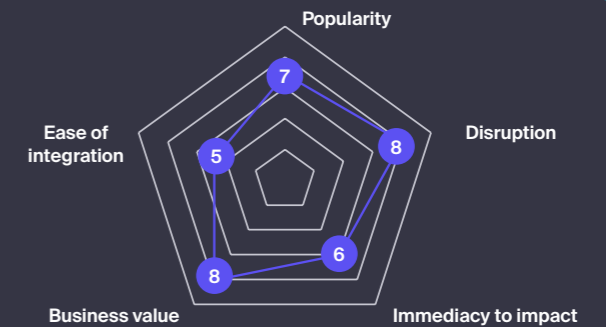
**Agentic Commerce** refers to a model in which AI agents act as intermediaries between the user and businesses, taking charge of interpreting intent, discovering options, and executing purchase decisions on their behalf. The discovery and transaction process shifts from being centered on human navigation to being structured around systems that consume information, compare alternatives, and make decisions in an automated manner.

### Insight

The emergence of agents as intermediaries in the purchase process is redefining how demand is generated and captured. This shift **enables capturing higher-intent traffic, reducing decision friction, and automating the purchase process**, but it also introduces a new competitive dynamic where visibility depends on the ability to be selected within automated decision environments. Companies must evolve from models oriented toward attracting and converting users to **architectures designed to integrate into agent ecosystems**, where the purchase decision occurs outside their own channels.

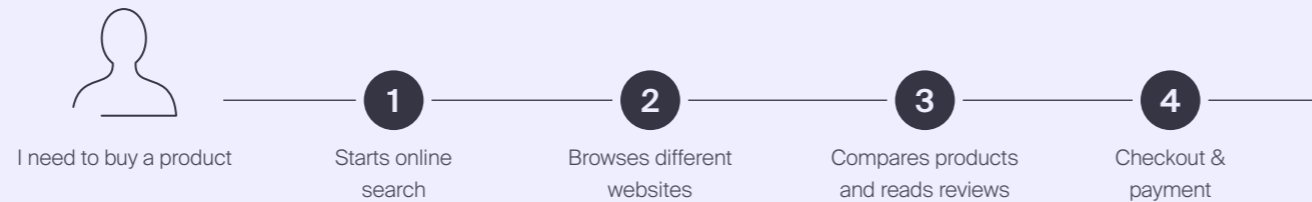
### Trend prioritization matrix

6,8



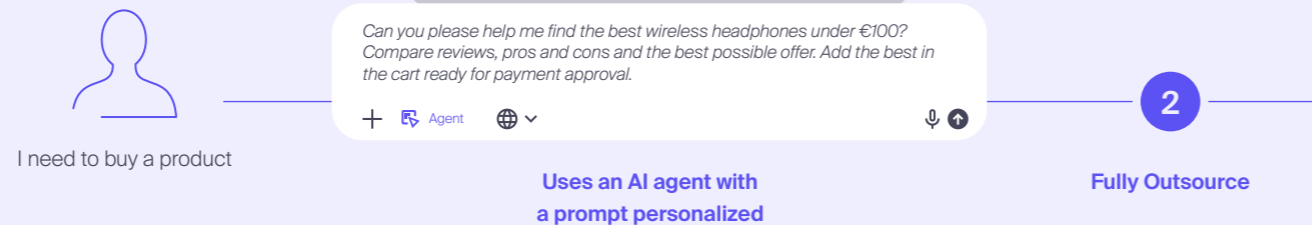
## From Traditional E-Commerce to Agentic Commerce

### Buyer journey E-commerce

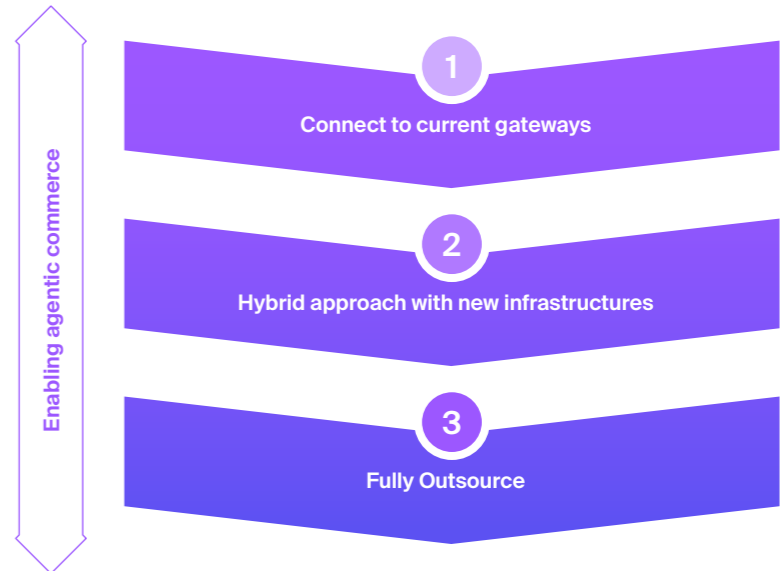


**AI agent performs an exhaustive search**, detailed product comparisons, and finds the best possible offer in second

### Buyer journey Agentic AI



## Agent Discoverability & Readiness



13

## Scalable Design Intelligence

Use generative design systems to scale experience creation without increasing team size

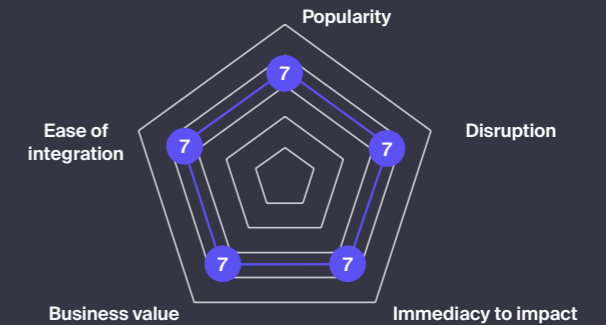
**Scalable Design Intelligence** is the evolution of design systems toward intelligent platforms that integrate generative capabilities and governance rules to design, adapt, and scale digital experiences in an automated manner. These systems enable generating interfaces, components, and complete flows from business inputs or intent, maintaining brand coherence, accessibility, and quality, without depending on the manual capacity of design teams.

### Insight

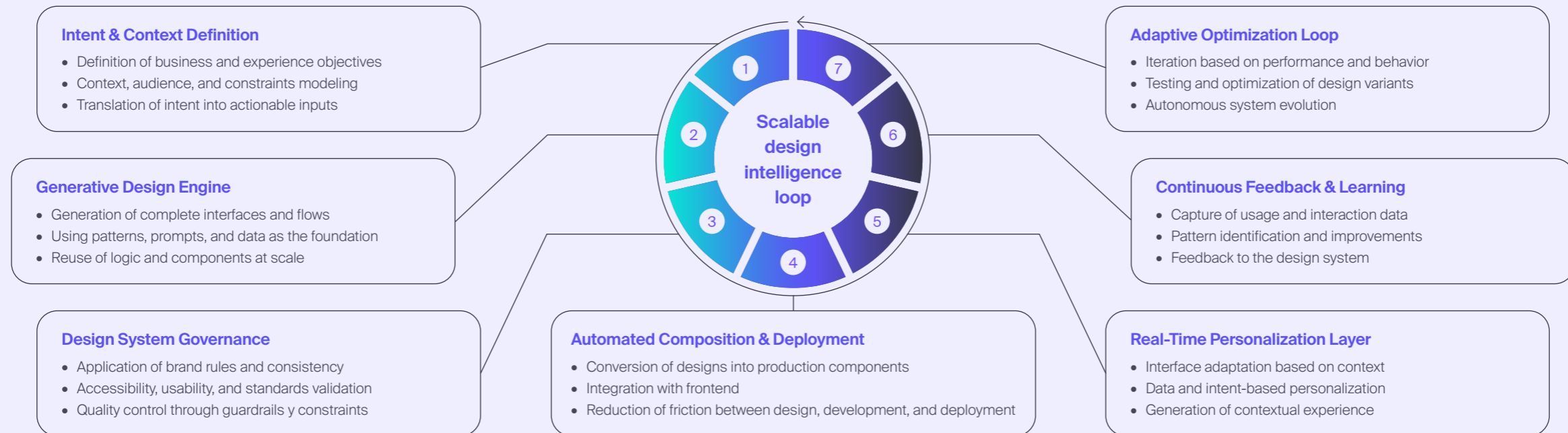
The explosion of digital channels, conversational interfaces, and personalization needs has turned design into a bottleneck for product delivery speed, now that interface generation is no longer limited by technology but by the ability to design and govern them. The **integration of generative capabilities within design systems enables scaling experience production** without proportionally increasing teams, shifting the focus from manual creation to **defining rules, patterns, and guardrails that ensure consistency and quality**.

### Trend prioritization matrix

7,0



## Transform intent, rules, and feedback into consistent, personalized digital experiences



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## Conversational & Multimodal UX

Redesign interfaces around intent to reduce friction and shorten user journeys

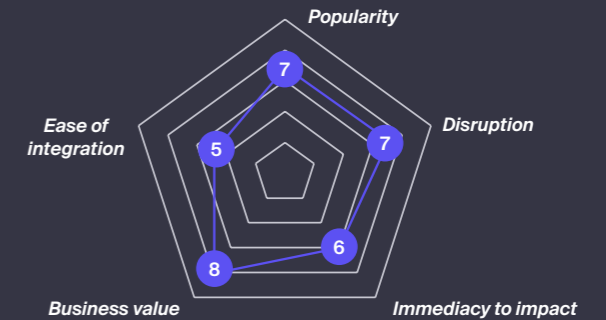
**Conversational & Multimodal UX** refers to the design of experiences where interaction moves beyond traditional graphical interfaces to be structured around **natural language and multiple modalities within the same flow**. Users interact through intent, while **systems interpret, respond, and generate experiences**. This involves creating new interaction patterns, components, and design principles.

### Insight

The shift toward conversational and multimodal interfaces entails redesigning how business tasks are executed, eliminating steps and reducing friction. In this model, **users express intent and the system directly orchestrates the resolution**, which enables shortening processes, increasing conversion rates, and improving operational efficiency. However, **its value depends on how these flows are designed**: without clear objectives, defined metrics, and control over automated decisions, conversational interfaces tend to degrade into inconsistent experiences that do not scale.

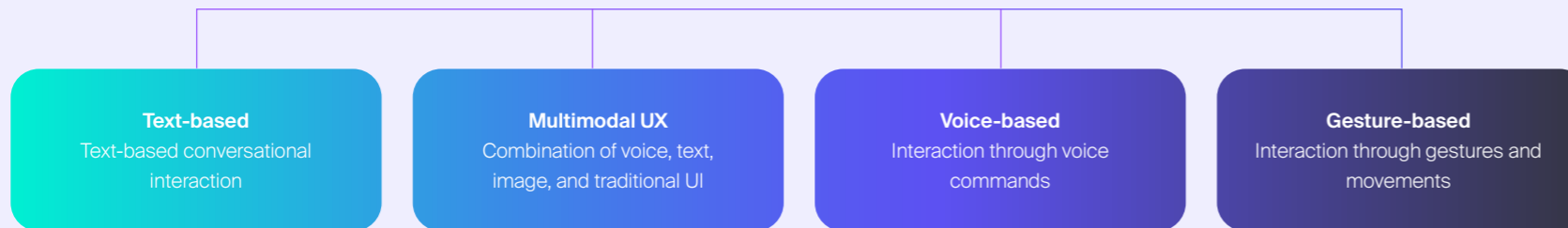
### Trend prioritization matrix

6,6



## New Forms of Digital Interaction

### Types of conversational UX interfaces



### Common Capabilities to Reduce Friction



### Strategic use cases

#### Sales qualification and lead conversion

The system filters and structures incoming prompts, guiding prospects through key questions. By leveraging previously gathered context, it shortens the sales cycle, **reduces customer acquisition costs, and increases conversion rates**.

#### Customer support cost optimization

Through flows designed to resolve repetitive tasks and predictable ticket deflection, achieving a **lower cost per ticket, reduced resolution times y support team scalability**.

#### Product onboarding and activation

The experience breaks down initial setup into sequential steps, translating into **faster activations, lower dropout** during sign-up and **better trial-to-paid conversion** a payment.

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## Trust-Centric experience

Embed transparency by design to increase trust, enable adoption and ensure regulatory compliance

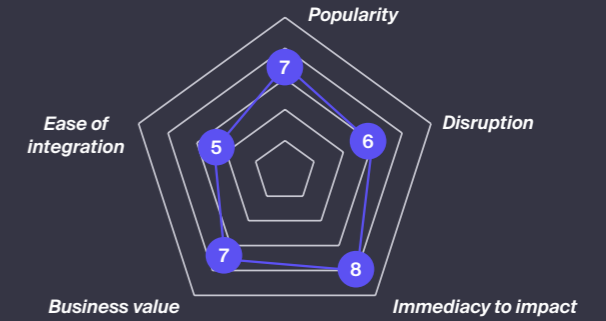
**Trust-Centric experience by design** is the discipline of designing AI-based digital products and services by placing transparency, explainability, and user trust calibration at the center of the process. If Explainable AI (XAI) is the technical capability to audit a model, being Trust-Centric is the organizational capability to translate that complexity into a user experience that generates security, control, and loyalty.

### Insight

Currently, **customers demand immediate answers and deeply distrust decisions they do not understand.** At the same time, the AI Act turns algorithmic transparency into a non-negotiable legal requirement for any organization. Incorporating transparency and explainability principles from the product design phase **enables organizations to unlock the true potential of automation in customer relationships.** Maintaining "black box" experiences exposes the organization to significant penalties, but above all erodes the return on heavy investments already made in generative AI.

### Trend prioritization matrix

6,6



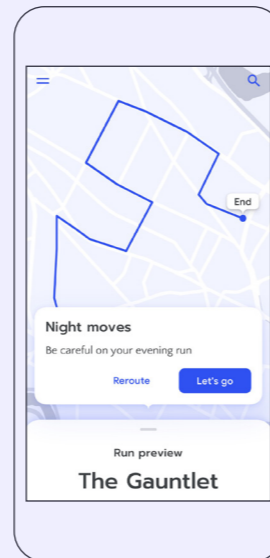
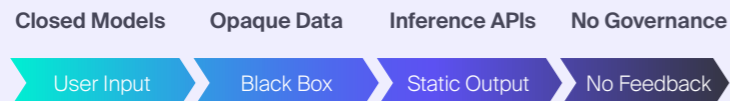
## Comparison between an opaque experience and a trust-centric experience, where AI explains its decisions, reveals the data used, and enables user control

### Opaque experience (black box)

Errors that erode trust and expose to regulatory risks:

- Presenting decisions without clear justification
- Hiding uncertainty or simulating absolute certainty
- Confusing transparency with technical exposure
- Designing experiences without considering risk level
- Showing outputs without explaining conditions
- Overloading interface with irrelevant explanations
- Eliminating the user's ability to intervene
- Breaking contextual continuity between interactions

#### Stack

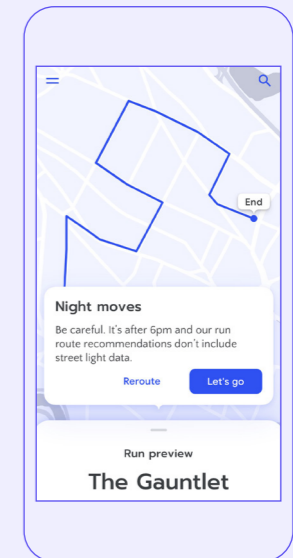
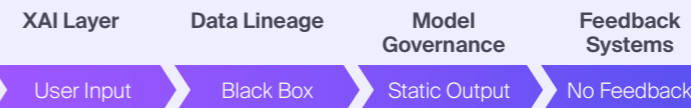


### Trust-centric experience

Actions that turn a solution into a reliable digital partner:

- Explicar "por qué" en decisiones con impacto
- Make used and missing data visible
- Translate uncertainty into clear signals
- Integrate correction and override
- Design explanations linked to context and user
- Adapt transparency level to case risk
- Expose traceability in an accessible way
- Maintain contextual memory in every interaction

#### Stack



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## AI-Native Enterprise

Redesign operating models around AI to scale execution and reduce dependency on human labor

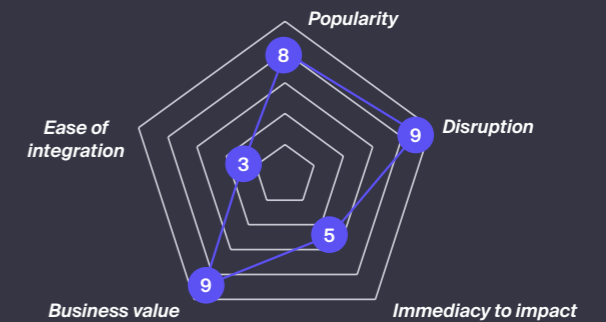
**AI-Native Enterprise** refers to an **organizational model in which AI is structurally integrated across the entire value chain**, from technology architecture to operational processes and decision-making. This approach materializes in agentic architectures, continuous data pipelines, automated workflows, and systems designed to learn and improve with use, **where AI is part of the base design of the product, operations, and experience.**

### Insight

The advancement of AI is redefining the operational limits of organizations, making it unfeasible to compete with models based on intensive human execution. This is shifting the focus toward their replacement by autonomous systems, **drastically reducing operating costs and increasing execution speed**. Consequently, organizations shift to scaling through intelligence, generating cumulative advantages that are difficult to replicate. In this new context, technology ceases to be a cross-cutting enabler and becomes **the core of the business model, capable of capturing value directly and redefining how to compete in the market.**

### Trend prioritization matrix

6,8



## How to Evolve Toward an AI-First Operating Model



Dimension	Digitally enhanced operating model	Ai-first operating model
Model engine	People as core, with digital support	AI agents as core, with human oversight
Processes and decision-making	People-centered processes, AI as support	AI-centered processes, automated decisions
Organization and roles	Traditional structure, enhanced static roles	Flat hierarchies, roles redefined around AI
Governance and risk	Reporting-based governance, ad hoc AI policies	Real-time governance with integrated AI and guardrails
Culture and focus	Efficiency-oriented	Speed, adaptability, and trust-oriented

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## Service as Software (SaS)

Shift from software delivery to outcome-based services to capture value directly from execution

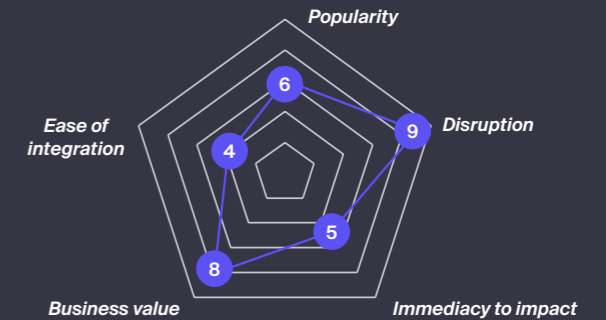
**Service as Software (SaS)** is a model in which **organizations deliver complete services executed by software, rather than providing tools for the user to execute them.** This approach is based on the **combination of artificial intelligence, automation, and autonomous** workflows to operate end-to-end processes, integrating data, decision logic, and execution in a single layer.

### Insight

AI's ability to automate decisions, interpret context, and execute complex processes is shifting value from software usage to the direct delivery of results, making license-based models lose relevance against outcome-oriented schemes. This **reduces the need for client expertise, eliminates operational friction, and enables scaling services** much more efficiently. In this way, companies stop capturing value through access to functionalities and **shift to doing so through the impact generated, transforming the logic of pricing, differentiation, and competition.**

### Trend prioritization matrix

6,4



## From SaaS to Software-Delivered Services

Moving from selling tools to delivering complete outcomes through AI and automation

Dimension	SaaS	SaS
Differentiation	Low – standardized workflows	High – contextualized and adaptive execution
Speed of change	Periodic release cycles	Real-time reconfiguration
Interface	Fixed UI (screens, forms)	Conversational, multimodal, and generative
Governance	Based on static roles and rules	Based on context and dynamic policies
Integration	Manual integration through APIs	Automatic orchestration through agents
Cost model	Subscription per user/license	Pay per task, usage, or outcome

### Integration implications by role

- CIO** Architecture evolves from isolated applications to capabilities orchestrated by intelligent agents
- COO** Processes cease to be static and begin adapting dynamically according to operational context
- CFO** Financial model migrates from licenses to cost directly linked to generated results
- Business** Decisions cease to be based on standards and are optimized in real time according to context and interaction

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## Resilient & Autonomous Supply Networks

Deploy AI-driven supply networks to anticipate disruption and maintain operational continuity

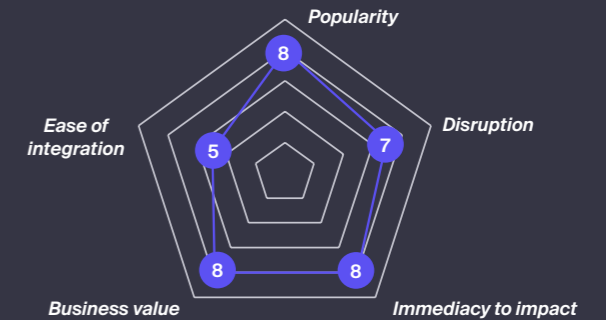
**Resilient & Autonomous Supply Networks** is the evolution of supply chains toward intelligent networks capable of anticipating disruptions, dynamically adapting, and executing decisions autonomously. This **integrates real-time data** (IoT, operational systems), **predictive analytics, scenario simulation, and AI agents** that automate key processes, enabling continuous, connected management oriented toward resilience.

### Insight

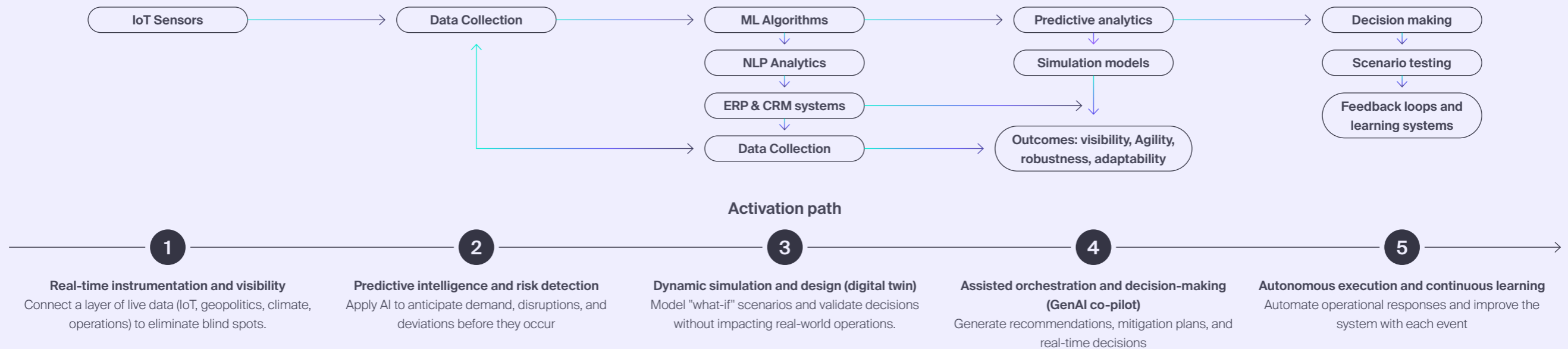
The growing geopolitical, climate, and economic volatility has turned disruption into a structural condition of supply chains, making models based on static planning and manual reaction ineffective. AI integration enables shifting from reactive management to **anticipatory capability, where risks are identified before materializing and decisions are adjusted in real time**. This transforms the supply chain into a competitive advantage, **capable of maintaining continuity, optimizing resources, and constantly adapting to a changing environment**.

### Trend prioritization matrix

7,2



## How to connect data, predictive analytics, simulation, and AI-assisted decision-making to anticipate risks and execute autonomous responses



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### Preemptive Cybersecurity

Shift from reactive defense to predictive prevention to reduce attack impact and exposure

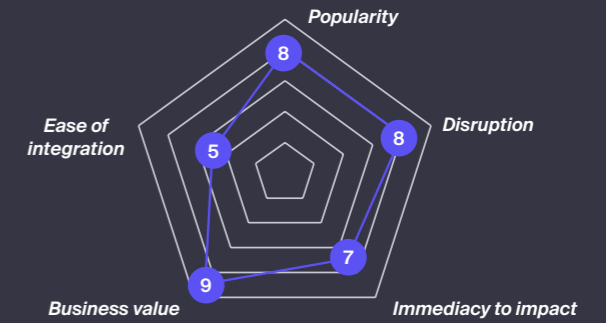
**Preemptive Cybersecurity** is a security approach that **seeks to anticipate, deflect, and neutralize threats before they execute**, rather than detecting them once they have compromised the system. This model **integrates AI-based predictive intelligence, continuous exposure management, deception techniques, and dynamic architectures** such as moving target defense that introduce unpredictability into systems.

#### Insight

The automation of attacks through AI is drastically reducing the time between reconnaissance, exploitation, and propagation, making a defense model based on detection and response unviable, where the attacker always has a temporal advantage. Therefore, **security can focus on preventing the attack from materializing**. Organizations that adopt **this approach will be able to reduce the operational and economic impact of attacks by having cybersecurity with a preventive rather than reactive capability**, where the objective is no longer to manage incidents but to reduce their likelihood of occurrence.

#### Trend prioritization matrix

7,4



### Security shifts from responding to incidents to preventing attacks from ever being executed



20

## Non Human Identity (NHIs) & Zero Trust Architecture

Secure machine identities to control access and enable safe automation at scale

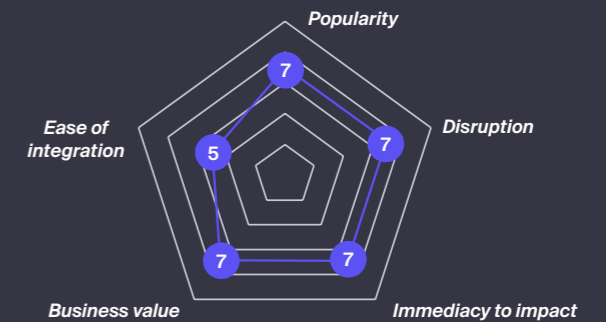
**Non Human Identity (NHIs) & Zero Trust Architecture** is a security model where identity, both human and non-human, becomes the core of access control in distributed systems. It integrates **machine identity lifecycle management** (APIs, agents, workloads, devices) with **Zero Trust principles**, establishing continuous validation, least-privilege access, and complete traceability in every interaction.

### Insight

Enterprise security is moving from being a network problem to becoming an identity problem at scale. In environments where machines far outnumber humans and execute most operations, managing who accesses what and under what conditions becomes critical to avoid systemic breaches. Machine Identity + Zero Trust redefines the security architecture by **shifting control from the perimeter to every interaction, enabling operation in distributed, dynamic, and autonomous environments without compromising security or compliance**, reducing the attack surface and enabling secure automation at scale.

### Trend prioritization matrix

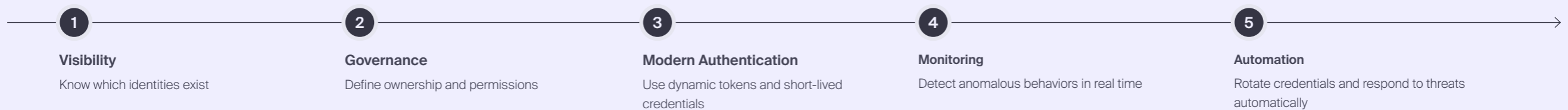
6,6



## Extend Zero Trust to non-human identities to strengthen access control, continuous verification, and automated protection



### Steps to Secure Non-Human Identities



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## Tokenized Asset Infrastructure

Tokenize assets to improve liquidity, transparency and programmability of ownership

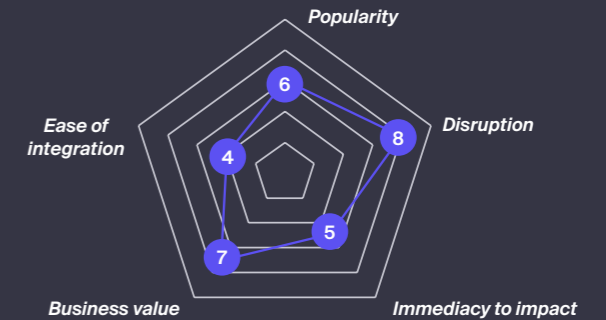
Tokenized Asset Infrastructure is based on the **representation of real-world assets** (RWAs) such as debt, real estate, commodities, or financial instruments **on blockchain infrastructures**, where ownership, transfer, and compliance are managed through programmable logic. This model enables fractional ownership, near-instant settlement, integration with digital financial ecosystems, and process automation.

### Insight

The tokenization of RWAs marks the transition to the **digitalization of ownership and rights over them**. In an environment where traceability, operational efficiency, and interoperability are critical, organizations need **infrastructures that enable managing assets beyond closed systems and fragmented jurisdictions**. Regulatory maturity and institutional adoption are pushing tokenization into multiple industries, turning it into a common layer for representing, transferring, and governing assets programmatically.

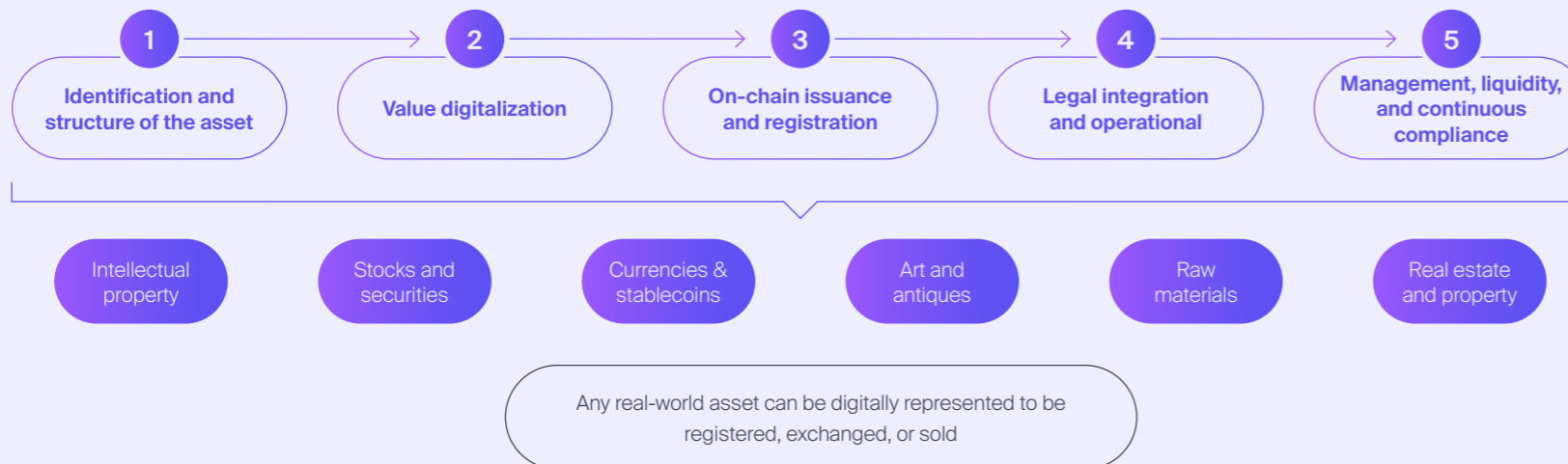
### Trend prioritization matrix

6,0



## Convert real-world assets into digital tokens to record ownership, execute transactions, and manage liquidity

### How a Real-World Asset Is Tokenized



### Use cases by sector



#### Finance

Tokenization of bonds, funds, and private credit transforms financial markets through on-chain issuance, real-time settlement, and programmable ownership



#### Real Estate

Tokenization of buildings or energy projects enables fractional investment, secondary liquidity, and continuous asset financing



#### Healthcare

Tokenization of clinical data and trials enables secure, auditable, and monetizable among ecosystem actors



#### Public sector

Tokenization of property records or digital identity enables secure management, fraud reduction, and administrative automation

22

## Green AI & Sustainable Computing

Optimize AI energy consumption to reduce costs and enable sustainable scaling

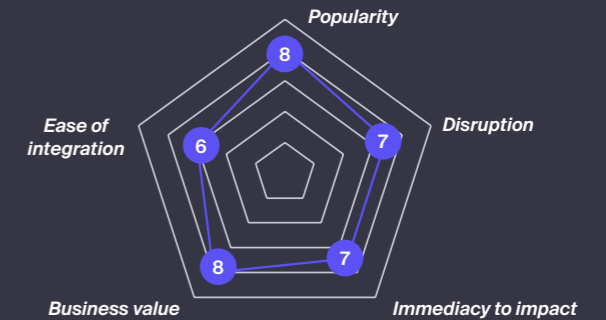
**Green AI & Sustainable Computing** is the capability to design, deploy, and operate AI systems under energy efficiency, resource optimization, and operational sustainability criteria, integrating: **Green in AI**, focused on reducing the AI stack's own consumption; **Green by AI**, oriented toward using AI to optimize processes and improve efficiency; and **Sustainable Computing**, which structures the infrastructure to support AI scaling within physical, economic, and regulatory limits.

### Insight

The exponential growth of AI is shifting the bottleneck from innovation to deployment. Thus, Green AI moves beyond ESG to become the **capability that determines which organizations can scale intelligence and which cannot**. On one hand, Green in AI redefines how intelligence is built and executed. On the other, **Green by AI** transfers that intelligence to business, optimizing supply chains, energy consumption, industrial operations, or networks, turning sustainability into measurable operational efficiency.

### Trend prioritization matrix

7,2



## How to Reduce AI's Energy Cost

### Green in AI: Making the AI stack more efficient

#### Carbon-aware scheduling

execution based on carbon intensity and energy availability

#### Hybrid compute optimization

Maximización de uso de GPU/TPU y balanceo de cargas

#### Hardware efficiency

Maximization of GPU/TPU usage and load balancing

#### Inference optimization

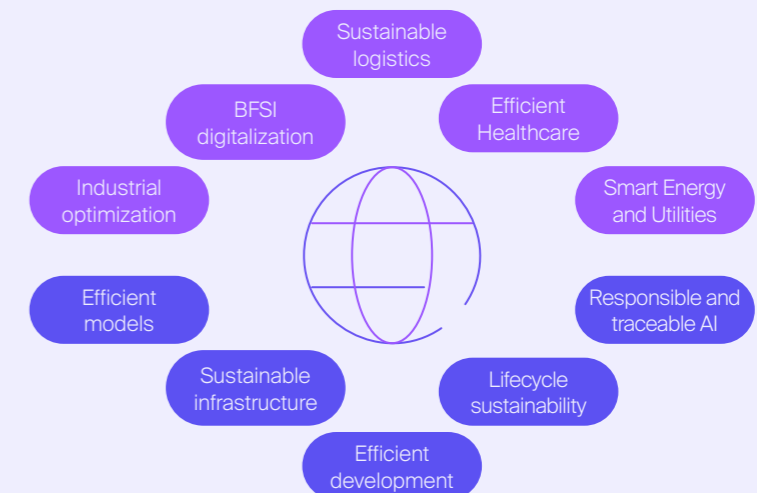
Reduction of cost per token and efficient memory usage

#### Energy telemetry

Real-time monitoring of consumption and performance

Sustainability evolves from a CSR metric into a driver of operational efficiency

### Green by AI: using AI to optimize business processes and consumption



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## Sustainability-as-a-Service

Automate ESG measurement and action to turn sustainability into an operational capability

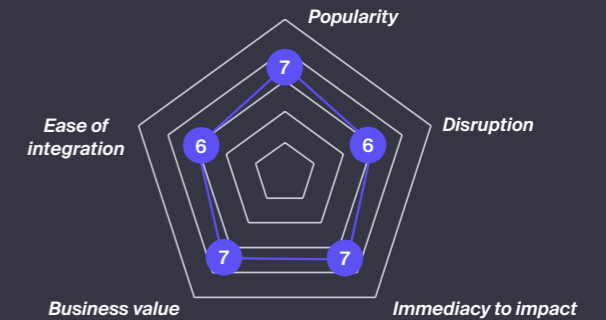
**Sustainability-as-a-Service** combines AI, operational data, and managed service models to transform sustainability into a continuous, measurable, and externalizable function. It integrates **automated ESG data capture, impact calculation** (emissions, resources, biodiversity), **and the execution of reduction** initiatives within an as-a-service model, where organizations consume sustainability as an end-to-end service instead of building it internally.

### Insight

Sustainability is shifting from a declarative exercise to an operating system that requires precision, traceability, and continuous execution. But **organizations lack the capability, data, and talent to measure and act** at scale on their real impact. Thus, Sustainability-as-a-Service emerges: **a model that industrializes sustainability through data platforms, AI, and specialized services that cover the entire cycle, from measurement itself to full intervention.**

### Trend prioritization matrix

6,6



### Measure, digitize, and automate ESG actions to move from reporting to continuous improvement



### Benefits



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## AI-Augmented Workforce

Redesign roles and workflows to integrate AI agents and scale productivity beyond human limits

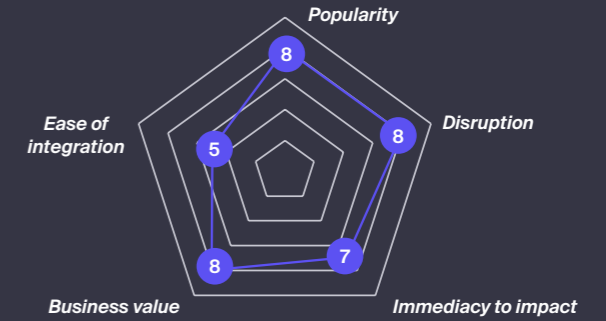
**AI-Augmented Workforce** redefines how teams are structured, **integrating AI agents as a new work category**, called **silicon-based workforce**, that **operates alongside human talent**. It is not about automating existing tasks, but about redesigning roles, processes, and responsibilities so that humans and agents collaborate in real time, leveraging their complementary strengths.

### Insight

Current work models remain designed for human capabilities, limited in scale, availability, and execution, which prevents capturing the real potential of automation. The real change occurs when organizations **redesign their processes as hybrid systems, where agents take on continuous execution** and humans focus on oversight, decision-making, and value generation. Organizations that adopt this logic will be able to scale productivity and operational capacity without depending exclusively on human capital growth.

### Trend prioritization matrix

7,2

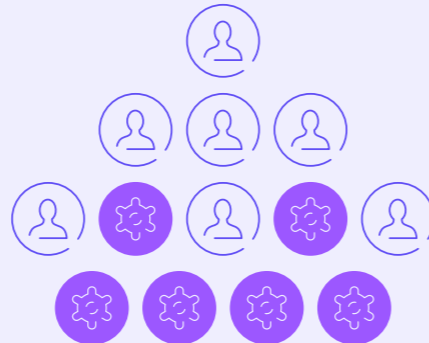


## Evolution from human teams augmented by AI to organizations where agents execute work in an integrated way

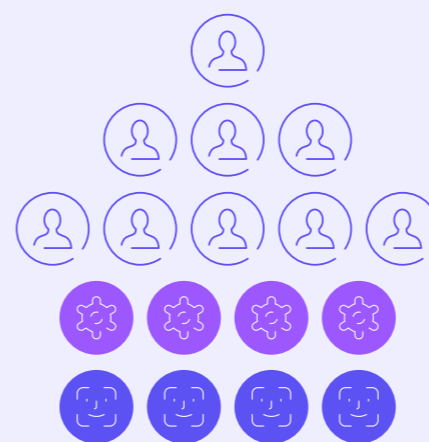
**Human-centric workforce**  
Humans with task-specific AI



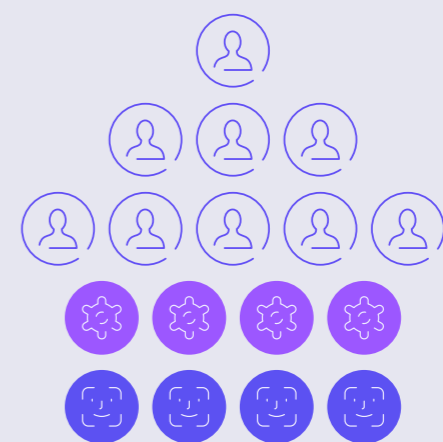
**AI-Supported work**  
AI automating tasks



**AI-Augmented workforce**  
Agents collaborating in workflows



**Agent-native organization**  
Agents integrated as a digital workforce



Humans
 Automation, AI, other tools
 Generative AI



25

## Full-Journey Process Automation

Redesign end-to-end processes for autonomous execution, reducing manual intervention and operational friction

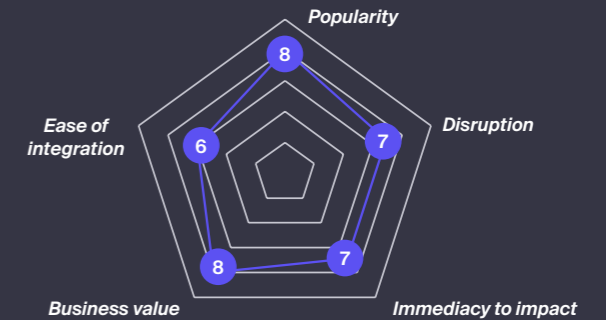
**Full-Journey Process Automation** involves **redesigning complete end-to-end processes to be executed by orchestrated AI agent systems**, capable of deciding, coordinating, and acting without constant human intervention. Unlike traditional automation focused on isolated tasks, this approach **treats the process as a continuous unit, where multiple agents collaborate to achieve an outcome**, managing exceptions and adapting in real time.

### Insight

Current operating models continue to organize work into tasks executed and supervised by humans, even when capabilities to automate complete processes already exist. The emergence of AI agents **enables reorganizing work execution around autonomously managed end-to-end processes**, where human intervention shifts toward exception oversight and system control. This redefines the operational structure: **from teams that execute tasks to organizations that manage autonomous execution systems, changing the basis on which productivity and operational capacity are measured**

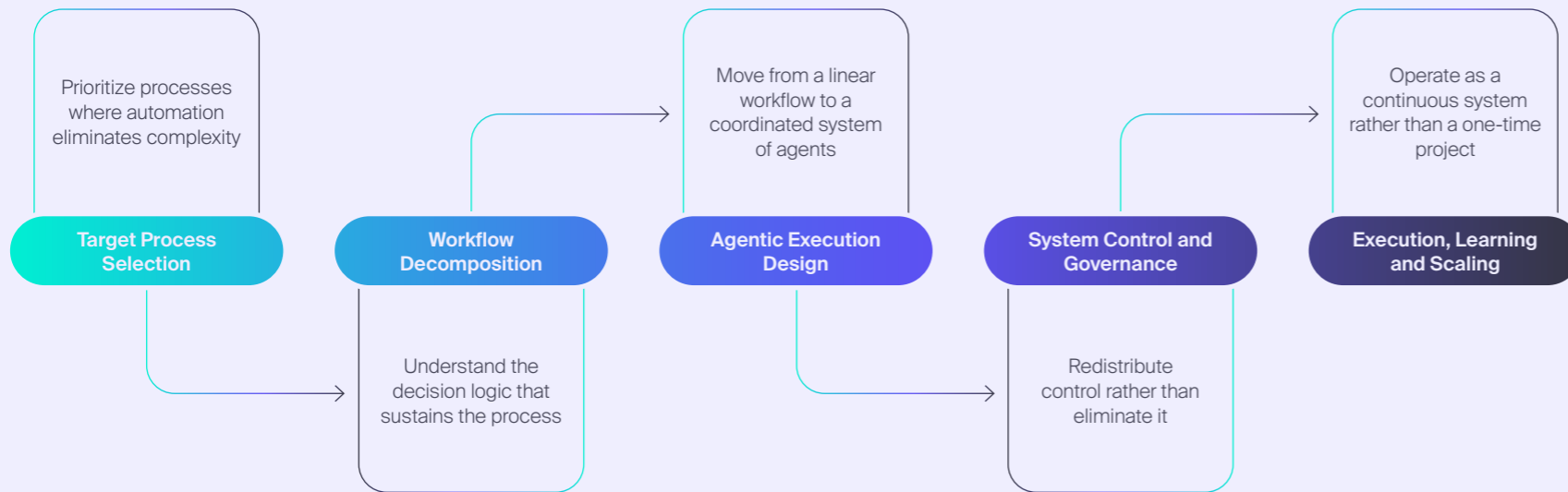
### Trend prioritization matrix

7,2



## From Linear Processes to Coordinated Agent Systems

Automation evolves from handling isolated tasks to orchestrating end-to-end processes as living, adaptive systems.



### Use cases by sector



#### Procure-to-pay Autonomous

Reduction of cycles from days to hours and elimination of manual bottlenecks



#### Autonomous incident response

Reduction of response time and real-time risk containment

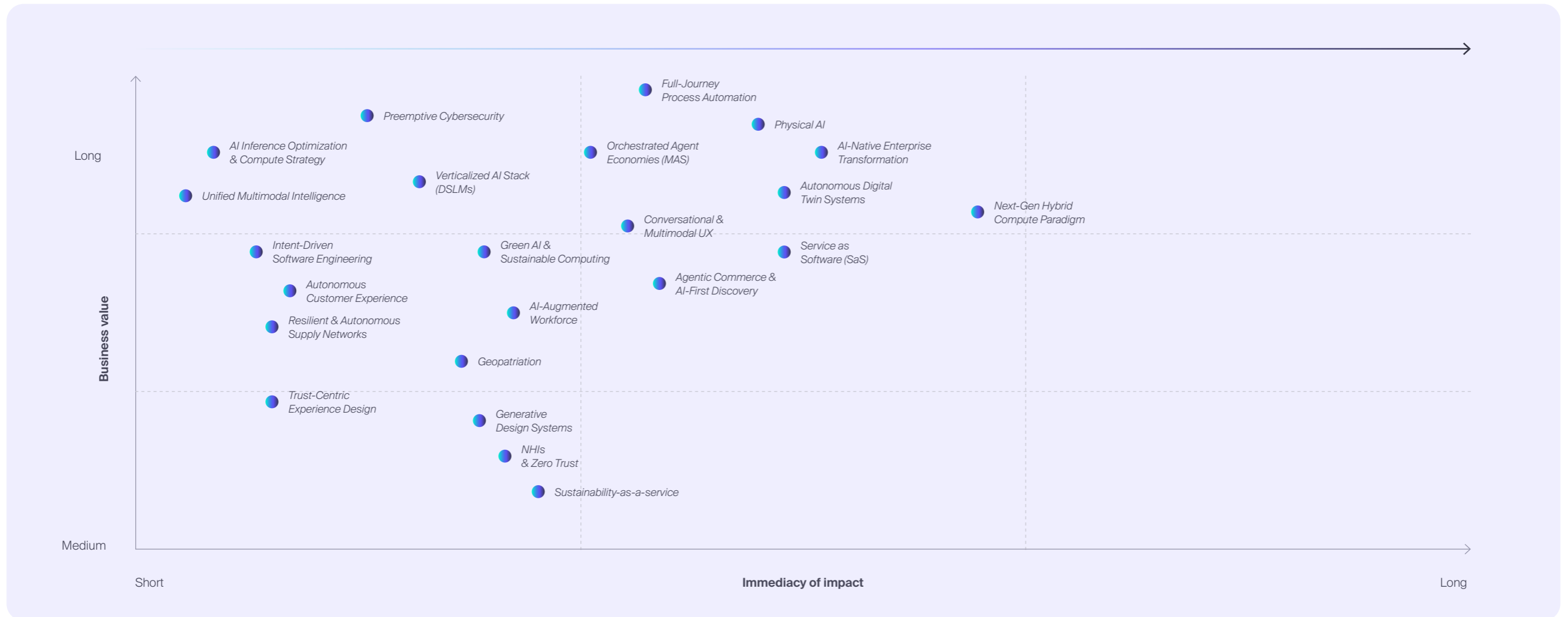


#### End-to-end order management

Scalability without proportional increase in operational costs

## Balance quick wins and long-term bets to maximize value across emerging technologies

The impact matrix illustrates the relationship between business value and time to impact for various emerging technologies. While technologies with high business value and short time to impact offer immediate opportunities for quick wins, those with high business value and longer time to impact represent significant long-term potential but require patience and sustained investment. Overall, the landscape shows a balance between rapid-commercialization solutions and transformative technologies that will develop over time.





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