

A Bespoke Global Al Advisory & Consulting Firm



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Overview

The manufacturing industry hasn't seen such challenges in the recent past. Globally, the sector is being impacted by multiple forces and trends. Not only must they deal with natural, macroeconomic and political volatility, they must also adapt to an ever-changing cohort of disruptive digital technologies — predictive analytics, additive manufacturing, smart manufacturing and the Industrial Internet of Things (IIoT) to name a few.

Manufacturing organizations must constantly reimagine how they function in this digital era, from creating a connected workforce to enabling predictive maintenance. And they must do all these in the middle of raising Industrial Consumerism – wherein Industrial companies are expected to become more and more consumer-centric moving from B2B to B2C to drive growth.

In the advent of Industry 4.0, The manufacturing industry has made significant progress in the adoption of advanced technologies for manufacturing operations and processes. Manufacturing organizations are implementing AI-powered intelligent solutions for enhancing operational efficiency. Thus, there is a considerable growth in the demand for AI for intelligent manufacturing processes or smart manufacturing solutions that offer decentralized decision-making, optimizes logistics & supply chains, enhances production outcomes, advances process effectiveness, and minimizes operational costs, enabling an intelligent manufacturing environment.

In 2019, Al met the Industrial Internet of Things and the use of artificial intelligence in the Industrial Industry began. Now a global transformation is underway to empower manufacturing with Al.

Al can help businesses elevate process automation, formulate forecasts of market trends, schedule production, and improve the efficiency of inspections. Al will be a pivotal technology to drive growth and innovation in the manufacturing sector.

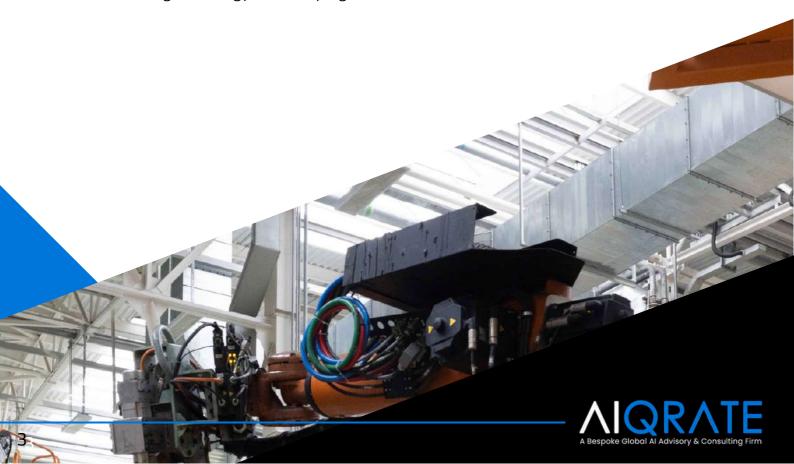


As the global manufacturing sector continues to be bombarded by a host of disruptive forces, manufacturers are under ever greater pressure to innovate and find new sources of growth. And now that AI technologies are coming of age, they're demonstrating they can provide a much-needed productivity boost, as well as help in reducing operating costs and transform customer experiences.

Al is believed to be the gamechanger in the industry of manufacturing. In manufacturing plants, the amount of data obtained from various sensors, software, IoT driven systems and their complex nature are making it impossible for humans to analyse. Adopting Al is considered to be the most efficient solution which enables the user to analyse and predict consumer behaviour, predict maintenance to prevent unwanted shutdown, detect abnormalities in the production process and much more.

Al also facilitates the use of real-time analytics to enable real-time decision making. Moreover, in the coming years, the increasing volume of data gathered through various devices, coupled with the widespread availability of high-speed broadband networks and the upcoming implementation of 5G technologies will further contribute to increased Al adoption in the Manufacturing industry.

There are many innovative technologies like Computer Vision, are revolutionizing manufacturing, making it more intelligent, productive, and efficient while emphasizing employee safety. Increased competition between regions and nations around the world has led to a revolution in manufacturing technology, and that progress will continue into the future.



Business Value Chain: Al Adoption Areas

- 1. Maintenance
- 3. Supply Chain Optimization
- 5. Production
- 7. Product Development/R&D
- 9. Failure Probability Modelling
- 11. Customer Service

- 2. Quality Control
- 4. Design Process
- 6. Demand Planning
- 8. Process Control
- 10. Price Forecasting
- 12. Yield Enhancement

1. Maintenance

Predictive maintenance of devices allows the manufacturer to cut device repair or maintenance costs. Using ML-powered predictive solutions, Al tools for manufacturing can predict when machinery requires maintenance services.

Similarly, cloud and the IoT sensors are also playing a vital role in modernizing the manufacturing industry. The AI systems and applications are trained to explore insights from the device performance data which is collected from connected IoT sensors. These insights help manufacturers to know the performance of individual devices. Such insights will assist companies in predicting the failure of the devices that has to happen in the future.

IoT sensors can monitor factors that affect equipment conditions across industries such as oil temperature, salinity levels, and vibration levels. IoT sensors can give operators critical insights into wear-and-tear as well as emergency issues, allowing them to shut down equipment to prevent catastrophic failure or take other appropriate actions.

2. Quality Control

Now-a-days, assembly lines are being data-driven, interconnected, and autonomous networks. These assembly lines work based on a set of algorithms that provide guidelines to produce the best possible end-products. All systems can detect the differences from the usual outputs by using machine vision technology since most defects are visible. When an end-product is lower quality than expected, All systems trigger an alert to users so that they can react and make adjustments.

Mectron's SQ-7500 uses a camera to do 360 visual inspection of cylindrical parts. A patented light tunnel aides in surface inspection to catch scratches, dents, splits and cosmetic defects. The parts traverse the machine by being rolled along an angled rail by a conveyor belt so that the camera can get 360 degrees of images to inspect the part.



3. Supply Chain Optimization

Al's contributions to the manufacturing sector don't stop at the production line. Algorithms can help companies improve how they deliver their products to their consumers via predictive analytics. Better informed firms are able to shift from a reactionary model to a more profitable and predictive one. IoT sensors can collect a myriad of data along the industrial supply chain, from transportation and energy consumption to raw material cost fluctuations to weather patterns and other market conditions that can have an impact on a company's bottom line.

4. Design Process

Generative design uses machine learning algorithms to mimic an engineer's approach to design. Designers or engineers enter parameters of design (such as materials, size, weight, strength, manufacturing methods, and cost constraints) into generative design software and the software provides all the possible outcomes that can be created with those parameters. With this method, manufacturers quickly generate thousands of design options for one product.

5. Production

Industrial robots have been in manufacturing plants since the late 1970s. With the addition of artificial intelligence, an industrial robot can monitor its own accuracy and performance, and train itself to get better. Some manufacturing robots are equipped with machine vision that helps the robot achieve precise mobility in complex and random environments.

Cobots are another robotics application that uses machine vision to work safely alongside human workers to complete a task that cannot be fully automated. Feel free to learn more about cobots with our comprehensive guide.

Edge analytics provides fast and decentralized insights from data sets collected from sensors on machines. Analysis of this data is performed on edge itself to reduce time to insight. Edge analytics has three use cases in manufacturing:

- Production quality and yield improvement
- Detect early signs of deteriorating performance and risk of failure
- Track worker health and safety by using wearables

6. Demand Planning

All enables organizations to optimize product availability by decreasing out of stocks and spoilage. All can also help with getting a better understanding of sales patterns.

For example, L'Oréal uses Al algorithms to predict demand based on a wide variety of data gathered from social media, weather, and financial markets.



7. Product Development/R&D

All enables organizations to expediate product development and R&D by reducing the test times and driving more concrete insights from customer data and demands.

For example, Intel is using big data and AI platforms to create tests for hard to validate functionalities improving the targeted coverage by 230x as compared to standard regression tests.

8. Process Control

Al can help organizations optimize processes to achieve production levels with enhanced consistency, economy and safety.

For example, Unilever uses AI to influence operations by predicting outcomes and improving efficiency levels to optimise output.

9. Failure Probability Modelling

Failure probability modelling has won its place in the manufacturing industry. The efficiency of the machine learning algorithms in the failure prediction is undoubtful. Active application of failure probability modelling helps to increase performance, predict occasional failures in the functioning and as a result reduce maintenance costs.

Harnessing a computer vision system, General motors analyses images from cameras mounted to assembly robots to spot any early indications of a failing robotic part. This helped the organisation prevent downtime costs which can reach upto US\$20,000 per minute.

10. Price forecasting

To manufacture products, it is necessary to purchase the necessary resources, and sometimes the prices can get affected due to a number of factors. With the rapid changes in prices, sometimes it may be hard to assess when it's the best time to buy resources. Knowing the prices of resources is also necessary for companies to estimate the price of their product when it's ready to leave the factory. Using AI, the system is able to provide accurate price recommendations just like in the case of dynamic pricing that's used by e-commerce businesses like Amazon where machine learning algorithms analyze historical and competitive data to always offer competitive prices and make even more profit.



UK startup ChAI uses AI and machine learning to predict the price fluctuations of the world's key raw materials, such as oil, aluminum, copper and others. Most suppliers only consider and model historical data sets, which takes time and can be too static. ChAI's unique approach is to use AI technology to analyse diverse economic and alternative data. Unlike its competitors, ChAI does not just analyse market data, but also uniquely valuable data sets such as satellite imagery and maritime transport data. This enables businesses to plan and forecast more effectively and to be more resilient to market fluctuations.

11. Customer service

In manufacturing, the importance of customer service is often overlooked – which is a mistake as lost customers can mean millions of dollars in lost sales. Al solutions can analyze the behaviours of customers to identify patterns and predict future outcomes.

For example, in 2018, Nokia unveiled the latest version of its Cognitive Analytics for Customer software, providing powerful new capabilities so service provider business, IT and engineering organizations can consistently deliver a superior real-time and personalized customer experience. The software allows service providers to quickly identify issues and prioritize improvements.

12. Yield enhancement

Manufacturers can now use AI systems to decrease scrap rates from defective products and get more value out of the materials that go into the production process. These gains are made possible by using AI systems to identify causes of yield losses that can be avoided by changes to production processes or product designs. The payoff can be huge. For example, in the semiconductor industry, decreasing scrap rates and testing costs can lead to a reduction in yield detraction of up to 30 percent of the total production cost.

programs and improve patient engagement.

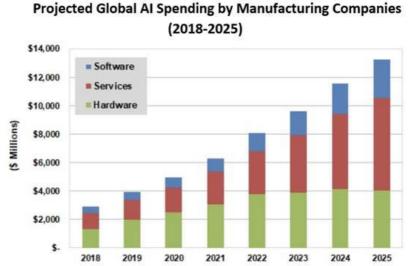
Manufacturing is estimated to generate about 1,812 petabytes (PB) of data every year, more than communications, finance, retail and several other industries.



Spending on Al

The Artificial Intelligence market size in manufacturing is valued at USD 2.3 billion in 2022 and is anticipated to be at USD 16.3 billion by 2027; growing at a CAGR of 47.9% from 2022 to 2027.

Annual worldwide manufacturing sector investment in Al software, hardware, and services is projected to increase to \$13.2 billion by 2025 from \$2.9 billion in 2018.



A few examples of spending & application of AI by Manufacturing companies on AI are given as follows: -

- GENERAL MOTORS (US): General Motors collaborated with Autodesk to implement generative design algorithms that rely on machine learning techniques to factor in design constraints and provide an optimized product design. Having constraint-optimizing logic within a CAD design environment helps GM attain the goal of rapid prototyping. Designers provide a definition of the functional requirements, materials, manufacturing methods and other constraints.
- NOKIA (Finland): Nokia has introduced a video application that uses machine learning to alert an assembly operator if there are inconsistencies in the production process. Nokia launched the video application that uses machine learning to monitor an assembly line process in one of its factories in Oulu, Finland.
- UPTAKE (Chicago): Uptake designs and develops enterprise AI software for many industries, including manufacturing. The artificial intelligence combines material, process and equipment data to offer suggestions on how to holistically improve the operations process. The AI specifically helps improve energy costs as well as optimize labour and product reliability.



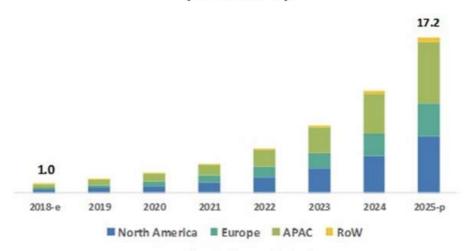
- VEO ROBOTICS (Waltham, Mass): Veo Robotics combines 3D sensing, computer vision and Al to make robots safer to work with in a manufacturing setting without the need for cages or operations shut-downs to perform inspections. The company's Al constantly monitors a robot's surroundings and temporarily shuts down a bot if a nearby human or a malfunction is sensed.
- MYTHIC (Austin, Texas): Mythic is a creator of intelligent processing units (IPUs) for a slew of
 devices, including some in the manufacturing industry. The company wants to create IPU's
 powerful enough to let robots and drones communicate conversationally so they can
 respond and react to situations in real-time.
- SIGHT MACHINE (San Francisco): Manufacturers use Sight Machine's analytics platform to gain real-time insights into every part of their manufacturing process and optimize operations as a result. The Sight Machine platform culls information from multiple sources, analyzes it, creates a digital "twin" of a company's infrastructure and then highlights areas where operations can be improved.
- INVISIBLE AI (San Francisco): How it's using AI in manufacturing: Invisible AI's platform gives
 manufacturers real-time visibility into its operational practices. The company installs smart
 cameras in crucial factory areas to capture the movements of a businesses' workforce and
 study its operational functions. The AI then analyzes the footage and suggests ways to
 improve everything from staffing and safety to efficiency.
- **OQTON (San Francisco):** Oqton is building an Al-integrated open factory operating system. The cloud-based platform couples design requirements with hardware capabilities so it can recommend production instructions and machine combinations for optimal factory efficiency.
- **FERO LABS (New York):** Fero Labs is an industrial machine learning application that helps global brands optimize their manufacturing processes. The company's artificial intelligence can discover hidden factors that are hindering production, predict maintenance issues and even optimize energy usage.
- LANDING AI (Palo Alto, California): Landing AI creates customizable artificial intelligencebased platforms to solve manufacturing needs. The company's Landing Light platform collects and labels data, develops project objectives, performs predictive modelling and offers manufacturers continued support to maintain optimal levels of production.

As manufacturing becomes more cost-sensitive and customers demand quality, manufacturers are using AI to enhance the performance of equipment, reduce downtime, and improve the quantity and quality of products.



Al Adoption across Regions

Artificial Intelligence in Manufacturing Market, By Region (USD Billion)



e: estimated; p: projected

- North America holds a significant share in the global artificial intelligence in manufacturing industry, and Asia-Pacific is projected to register the highest CAGR of 57.2% during the period 2020-2025, followed by North America.
- Al implementation can make robots smarter, reduce the downtime of machines, and increase the productivity.
- Hon Hai Precision Industry Co., Ltd (China) replaced 60,000 factory workers by robots.
- This high rate of adoption of industrial robots in manufacturing plants is expected to drive the growth of the AI in manufacturing market in APAC.
- The market growth is credited to the highly developed manufacturing plants in the countries such as Japan, South Korea, and China. The rapid adoption of the industry 4.0 revolution in the region also promotes the adoption of AI solutions. Moreover, the increasing investment in the AI technology in the emerging economies such as India and China are driving the market growth.

Al Adoption scenarios in this industry across US, Europe & Asia are explained as follows: -



Al Adoption across US

GENERAL MOTORS

Harnessing a computer vision system, General motors analyses images from cameras mounted to assembly robots to spot any early indications of a failing robotic part. For an organisation the size of General Motors this helped the organisation to prevent massive downtime costs which can reach at least US\$20,000 per minute.

BOMBARDIER

Partnering with Aurora, Bombardier has strengthened its resource planning and scheduling capabilities using AI enabled tools. With the help of AI Bombardier has the capability to schedule its airplane assembly operations quicker and handle able production rate changes more effectively.

BOEING

Harnessing computer vision for its aircrafts, Boeing is using the technology to scan various parts of an airplane via an augmented reality headset and other hardware to capture images. The images are sent to a back-end processing platform to detect certain abnormalities.

SPARKBEYOND

Sparkbeyond is an AI-enabled research engine that detects patterns in complex datasets. It developed an AI manufacturing technology for helping manufacturers to get the data related to input such as text documents and geospatial, business data and information related to World Bank and, weather conditions. Using all this data, Sparkbeyond generates reports that assist manufacturers in making the best business and final decisions.

AUTOMATION ANYWHERE

It is one of the biggest Automation companies and builds Robotic Process Automation (RPA) platforms for various industries. Manufacturing companies are using Al-enabled RPA platforms to automate the procurement, get insights, and manage and monitor the manufacturing processes efficiently.

BRIGHT MACHINES

Bright Machines brings together intelligent software with flexible factory robots and machine learning to help our customers meet the growing demands of the next generation of manufacturing.



KELLOGG'S

Kellogg's has launched an Al system that helps customers decide which recipe should be chosen to make a product of their choice on their website Bear Naked. This technology helps the consumer giant come up with final products that the consumers actually want.

RETHINK ROBOTICS

Rethink Robotics helps manufacturers meet the challenges of an agile economy with an integrated workforce, combining trainable, safe and cost-effective robots with skilled labour. Its Baxter robot, driven by Intera, an advanced software platform, gives world-class manufacturers and distributors in automotive, plastics, consumer goods, electronics and more, a workforce multiplier that optimizes labour.

OSPERITY

Osperity's technology provides Al-driven intelligent visual monitoring for industrial operations that can result in improved safety, reduced carbon footprints, and more.

KINTA AI

Kinta AI is an artificial intelligence platform which empowers the dynamic and digital factory of the future.

KUKA

KUKA, the Chinese-owned German manufacturing company, is one of the world largest manufacturers of industrial robots in the world. One use case of AI they have been investing in is helping to improve human-robot collaboration. Most industrial robots were very strong and stupid, which meant getting near them while they worked was a major health hazard requiring safety barriers between people and machines.

CATERPILLAR

Based on operation and performance data collected from shipboard sensors, Caterpillar's intelligent asset management system uses machine learning algorithms to optimize hull cleaning frequency by analysing cleaning cost and performance improvements under different levels of cleanliness.



Al Adoption across Europe

THALES SA

Leading supplier of electronic systems to aerospace and defence companies in France — is currently collecting historic and current data relating to part failures. That data is then utilised by Al algorithms to predict potential problems in the future in order to make proactive maintenance decisions for customers.

BMW

Currently within its operations the BMW Group is using AI to analyse component images from its production line for real time identification of deviations from the standard product quality

CARLSBERG

Implementing a 'Beer Fingerprinting' project, Carlsberg is developing sensors that have the capability to differentiate between different flavours of beer. Carlsberg processes the data via Al and uses the information to develop new beers as well as enhance the quality of existing beers.

SIEMENS

The German conglomerate Siemens has been using neural networks to monitor its steel plants and improve efficiencies for decades. The company claims that this practical experience has given it a leg up in developing Al for manufacturing and industrial applications.

CLOVER GROUP

Clover Group develops predictive analytics platform for industrial companies. It makes predictions about the technical condition of equipment and the manufacture of defective products.

BRIDGESTONE

Tyre manufacturer Bridgestone has introduced a new assembly system based around automated quality control, which resulted in more than 15 per cent improvement in uniformity of product.

PRESENSO

The Presenso solution uses advanced Artificial Intelligence to provide real-time asset failure predictions based on monitoring sensors' signal data in the cloud. With its proprietary adaptive algorithms, Presenso can analyse sensor behaviour, automatically learn how machines behave and use this learning to predict machine failures before they occur.



KONUX

The KONUX system is an end-to-end solution which uses IoT devices and artificial intelligence to improve network availability, extend asset lifetime and reduce costs. It continuously monitors and analyses the health of key switch components such as the track bed, and frog, and provides actionable recommendations. It ultimately allows for better maintenance planning by helping infrastructure managers anticipate failures before they happen and know the optimal time and type of maintenance needed.

Al Adoption across Asia

NISSAN

In order to ensure it complies with new regulations, Nissan is currently trialing the use of AI to design its cars. The technology is used to modify an existing car, but keeps in mind the effect of the modification.

KONUX

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FANUC

Fanuc, the Japanese company, which is a leader in industrial robotics, has made a strong push for greater connectivity and AI usage within their equipment. Fanuc is using deep reinforcement learning to help some of its industrial robots' train themselves. They perform the same task repeatedly, learning each time until they achieve sufficient accuracy.

• MITSUBISHI ELECTRIC

Mitsubishi deploys multiple sensors and AI technologies on its industrial robots, allowing them to grasp targeted objects within 3.5 milliseconds and rapidly recognize and adapt to changing conditions while maintaining precise operations.

WEIQIAO TEXTILE

Coupled with sensors installed on production assets, Weiqiao acquires and integrates all its textile workshop data, including machine operating data, product quality information, personnel information, equipment power, room temperature and humidity, etc., into a big data cloud platform for further analytics. This enables manufacturing system and data management on a fully automated textile production line.

AQUANT

Aquant's predictive AI technology monitors equipment, aggregates data, and helps spot equipment failure before it happens. The company's systems operate across several verticals including Manufacturing, Life Sciences and Medical, HVAC, Communications, Computer and Office Equipment, and Home Appliances.

FLUTURA

Flutura is an Al Solutions company focused on improving two core business objectives of "Asset Uptime" and "Operational Efficiency". Flutura does this with Cerebra, their Al Platform tuned for IIOT in Process Manufacturing, and Heavy machinery manufacturing industries, powering connected asset and connected operations use cases.

- North America possess high growth potential, due to increase in adoption of industrial robots in the manufacturing sector of the region
- North America is the major contributor in the AI in manufacturing market, wherein the US accounted for the largest share
- Cross-industry participation in the manufacturing domain, along with a significant increase in venture capital investment, has further propelled the growth of the Artificial Intelligence in manufacturing market in North America
- APAC to account for significant share followed by North America in Artificial Intelligence in manufacturing market during the period 2020-2025
- APAC is also considered to have the greatest number of manufacturing plants in the world. There are a few dark plants in China, where only robots work, and no human is required
- Europe AI in manufacturing market is anticipated to grow at a CAGR of over 44% over the period 2020-2025.

As a world manufacturing hub, Asia has great potential for industrial application of artificial intelligence. Among other countries, China, Japan, and South Korea who are more competitive regarding policies, research and development (R&D) capabilities, data, and talent are viewed as leading the way on Al development in Asia.



Impact on Revenue and Cost

The Artificial Intelligence revenue in manufacturing market is expected to grow to USD 17.2 billion by 2025, at a CAGR of 49.5% during the period 2020-2025.



Al will add \$3.7 trillion dollars revenue to the manufacturing industry by 2035.

- Increasing venture capital investments in AI are fostering growth of artificial intelligence in manufacturing market.
- Investments in the technology are led by tech giants and digital native companies including Google, Nvidia, and Intel.
- They are collectively investing massively in this industry in a wide variety of Al-based applications ranging from machine learning to robotics, assistance technology, and natural language to computer vision.
- The exponential growth in digital data is driving growth of artificial intelligence (AI) in manufacturing market. Today, it is estimated that approximately 1.7 megabytes of new data is created every second. It is further estimated to grow at an annual growth rate of 40% over the next 10 years.

Factors responsible for growth of AI in manufacturing market are:

- Huge availability of data sets and cloud storage solutions
- Increase in venture capital investments
- Evolving industrial IoT and automation is further supplementing the growth of the AI in manufacturing market.
- Widespread usage of machine vision cameras in manufacturing applications, such as machinery inspection, material movement, field service, and quality control drive the growth of the artificial intelligence in the manufacturing market.
- Moreover, key market players are adopting various strategies such as product launch and product innovation, to expand their existing product portfolio and maintain competitiveness in the rapidly growing AI marketspace.
- Oracle launched new artificial intelligence-based apps for supply chain, manufacturing, and other professionals. IBM launched Al-powered Watson Assistant for businesses. This product is a smart enterprise assistant powered with artificial intelligence (Al) features.



- GE launched Brilliant Manufacturing Suite to allow customers to realize their own vision of a Brilliant Factory.
- Furthermore, with the use of automation and AI in the manufacturing industry, risks during manufacturing processes are reduced and customers can get answers on an immediate basis, thereby enhancing the customer experience.

The extensive proliferation of advanced technologies, such as robotics and industrial IoT, in the manufacturing sector is the primary factor driving the market. The rising labour costs is also driving the adoption of advanced manufacturing technologies. The high investment by the major automakers to incorporate next-generation production technologies is also adding to market growth.



Challenges

The expectations are high for the application of artificial intelligence in manufacturing industry. Meeting the requirements is a real challenge to industrial partners. On the one hand training algorithms require lots of clean, bias free data sets, otherwise the result of the training would be wrong. On the other hand, cybersecurity is another threat that must be taken into consideration with the increasing use of connected technologies.

Some of the key challenges faced by the global manufacturing leaders in Al adoption are explained as follows:

1. Technology Gap

Many countries still face problems closing the technology gap between chip design and industrial software development for implementation of AI solutions in manufacturing. They must either import backbone digital products from foreign suppliers or build up their own tech capabilities, both of which can be costly and time consuming.

2. Data Scarcity

Machine learning models require a large high-quality dataset to train well-performing algorithms, yet many heavy industry facilities lack the right data type needed to properly train machine learning models

3. Manufacturing Standards

Manufacturing standards are different throughout the world, and this standard variance has become a major barrier to widespread AI deployment in the manufacturing industry.

4. Market Downturn

Compared with other industries, manufacturing has a relative low return on investment (ROI) and is less profitable in the short term. Consequently, many investors feel less confident about investing in new manufacturing technologies especially during a market downturn.

5. Data Inaccuracy

For machine learning to work properly, a lot of data is needed. Consumer data is easier to understand. Manufacturing data often is biased, outdated, and full of errors. For example, sensor data collected on the production floor in extreme, harsh operating conditions, where extreme temperature, noise and vibration variables can produce inaccurate data. Plants have historically been built using many proprietary systems, which do not talk to one another, where operational data also may be spread across multiple databases in multiple formats not suitable for analytics, requiring extensive preprocessing.

6. Predictions of less value

All predictions by Al solutions do not always mean better decisions. For example, industrial companies with ML platforms that predict faults of low risk areas. This results in wasting money replacing small parts with little Rol. It's critical to prioritize risk and fault predictions which can justify the Rol.

7. Deploying Edge Computing

Industrial AI is built as an end-to-end system, where data is generated by sensors on the edge, served to algorithms, modelled on the cloud, and then moved back to the edge for implementation. In a manufacturing facility that crushes ores into platinum bars, bars that come out with the wrong consistency must be immediately detected in order to adjust the pressure at the beginning. Any delay means wasted material.

Edge computing allows manufacturers to process data locally, filter data and reduce the amount of data sent to a central server, either on site or in a cloud. Additionally, a key goal in modern manufacturing is to be able to use data from multiple machines, processes and systems to adapt the manufacturing process in real-time. The ability to deploy predictive models on the edge devices such as machines, local gateway, or server is critical to enable smart manufacturing applications.

8. Implementation cost and Rol

Manufacturing organizations adopting AI often face the challenge from high implementation costs across talent, technology & infrastructure. This makes it hard for them to justify adopting ML solutions and ensuing Rol.

The benefits of using AI in the manufacturing sector are clear. Whether they lie in digitally focused innovation, enhancing user experience, bringing new levels of operational efficiency, or a completely new competitive edge, the technology holds huge potential for companies willing to make the jump to intelligent operations. But the risks and challenges ahead should not be underestimated. A balance must be struck between using human employees and AI-enabled machines, including a full consideration of the ethics of AI development.



The Way Forward

For AI investments and strategies to remain competitive and differentiated, Manufacturing companies will have to move their thinking beyond short-term gains and in-house siloed deployment plans. Instead, they'll need to embrace a more holistic strategy based on identifying "best fit" AI partners, investing in AI strategy build along with broader ecosystem partnerships and collaborating efficiently within them. Rather than being tactical, limited to a single product line/plant/geo, these relationships will be strategic, focused on designing and implementing the AI technology path that the Manufacturing company needs to navigate.

Key steps for Al Adoption in this industry are:

1. Design Al Strategy

The first step in achieving scale involves designing AI strategy in conjunction with processing data in real time from the shop floor/production environment. To automate the collection of real-time, live data, the prototype needs to be integrated with legacy IT (MES and ERP) and industrial internet of things (IIoT) systems.

2. Data Governance Strategy

Put down solid foundations of data governance and AI talent to create a robust foundation for scale, and to encourage new implementations, manufacturers should design a data governance framework that defines critical processes related to the generation, management, and analysis of data. In addition, they need to deploy a data & AI platform — a central platform to store and analyse data using AI and to make it available to issue-specific AI applications.

3. Align with strategic objectives

Manufacturing companies need to ensure their AI deployments match their strategies and business goals, be that bringing new revenue, reducing costs, or enhancing operational efficiency. The key is to choose deployments of appropriate complexity to deliver business goals.

4. Define use cases

Ascertaining where a technology can outperform humans is the proper strategic approach to finding the right Al application scenarios.

5. Build data foundation

Al based on deep learning still relies on big data. A company's data foundations determine whether its Al project will work.



6. Establish teams and partnerships

If a company wants to develop AI capabilities, it needs a professional team with AI technology expertise, industry expertise and AI adoption expertise.

7. Run POC and implement at scale

After establishing a clearly defined scenario, complete data foundations, and a professional team, the next step is designing a prototype and running a proof of concept (POC) of the Al process. Iteration and large-scale implementation can be carried out if the prototype proves feasible. Performance needs to be continuously monitored for value generated, output quality and reliability.

The Manufacturing companies that take the lead in overcoming the challenges, and collaborate wisely with their wider ecosystems, are set to be the leaders of tomorrow. So far, the world has only seen a fraction of what AI has to offer. Manufacturing companies have barely scratched the surface of the potential and possibilities of Artificial Intelligence. That needs to change. It's time to make the leap to intelligent industrial manufacturing.





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