

## INTRODUCTION

## AI IN MANUFACTURING



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## INDUSTRY 4.0 In Manufacturing

The Fourth Industrial Revolution, or **Industry 4.0**, becomes each day more relevant in **manufacturing**. This new wave of technology is very appealing to **producing companies** since it makes it very simple to **increase productivity** in operations.

With its rising popularity, manufacturers that fail to adapt to the newest technology in the industry will be at a disadvantage towards competitors that managed to **digitize** their production along the way.

## **AI and Machine Learning**

work well with manufacturing data. Hundreds of variables influence the production process. While it is very complex for humans to analyze them, machine learning models can forecast the impact of individual variables in such circumstances. Machines still operate below human skills in other industries involving language or emotions, limiting them to adapt to the A.I. field.





# 2. In Manufacturing

### **Predictive Maintenance**

Predictive Maintenance techniques enable manufacturers to predict potential machine downtime. By analyzing data provided by sensor devices, manufacturers use A.I. technology to forecast when the machines will fail. This allows them to plan optimal repair and maintenance dates, in a way that assures equipment efficiency, as well as avoids extra costs.

## **Predictive Quality**

Predictive Quality enables companies to make data-driven predictions of product- and process-related quality. The goal is to optimize quality by using predictions as a basis for deciding on actions. This includes not only action measures derived by the user himself, but also recommended actions provided prescriptively by a machine learning data analytics model. This results in less scrap and reduces quality inspection costs.





# 3. In the industry

The benefits of A.I. in manufacturing are widely known. However, producing companies still hesitate to adopt these technologies in their facilities.

#### Here is some data

Less than 14% utilize A.I. to increase productivity and sustainability. About 75% of these companies hesitate to use A.I. in production, due to the missing software experts that understand the use-case, or because the return of the investments is still unclear. Finally, they also worry about data safety and are also often conservative about clouds. Production engineers need solutions that are:



faster and truly **drive productivity**.





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## **APOLLO SOFTWARE** For Predictive Maintenance



Our APOLLO software is for predictive maintenance of production units. It is easy to set up and connect machine tools, robots, or measurement systems. It is 100% automated, predicts downtime and optimal maintenance dates. It tracks operations, conditions, events, and alerts. Of course, it is a low-code platform that runs on-premise or in any cloud.



## **ARES SOFTWARE** For Predictive Quality



Our ARES software is for predictive quality and process optimization. You can load the data from any source. It automatically finds and utilizes all relations in the data. You can easily integrate predictions and optimizations into your live processes. Of course, it comes as a **nocode platform** that runs onpremise or in any cloud.





# 7. Process at IconPro



### Introduction

Processes and Data Sources Analysis.

### **Test Study**

Data Assessment and Feasibility Analysis.

#### Implementation

Software Interfaces Implementation and Validation.

### Deployment

Results Deployment, Software Installation and General Recommendations.

### Monitoring

Software / Algorithm Validity and Results Sustainability.



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