

# (Gen)AI for plastics and composites


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MINISTÈRE  
DE L'ÉCONOMIE,  
DES FINANCES  
ET DE LA SOUVERAINETÉ  
INDUSTRIELLE ET NUMÉRIQUE



« Enhance the competitiveness of French plastics and composites converters...  
... by carrying R&D and technological transfers...  
... an generate knowledge to be shared amongst stakeholders »

TRL 4 → TRL 7-8

**For an Ethical, Responsible and Environmentally-friendly plastics industry**

## Axe 1

Ecological and Energy Transition

1. Improving the **eco-design** of plastic products and develop **new usage models** (ex. reuse)
2. Ensuring the **quality of Recycled Materials** (PCR)
3. Developing **low environmental impact** alternative solutions for a **sovereign industry**
4. Protecting **human** health and natural **ecosystems**

## Axe 2

## Digital Transition and Industry of the Future

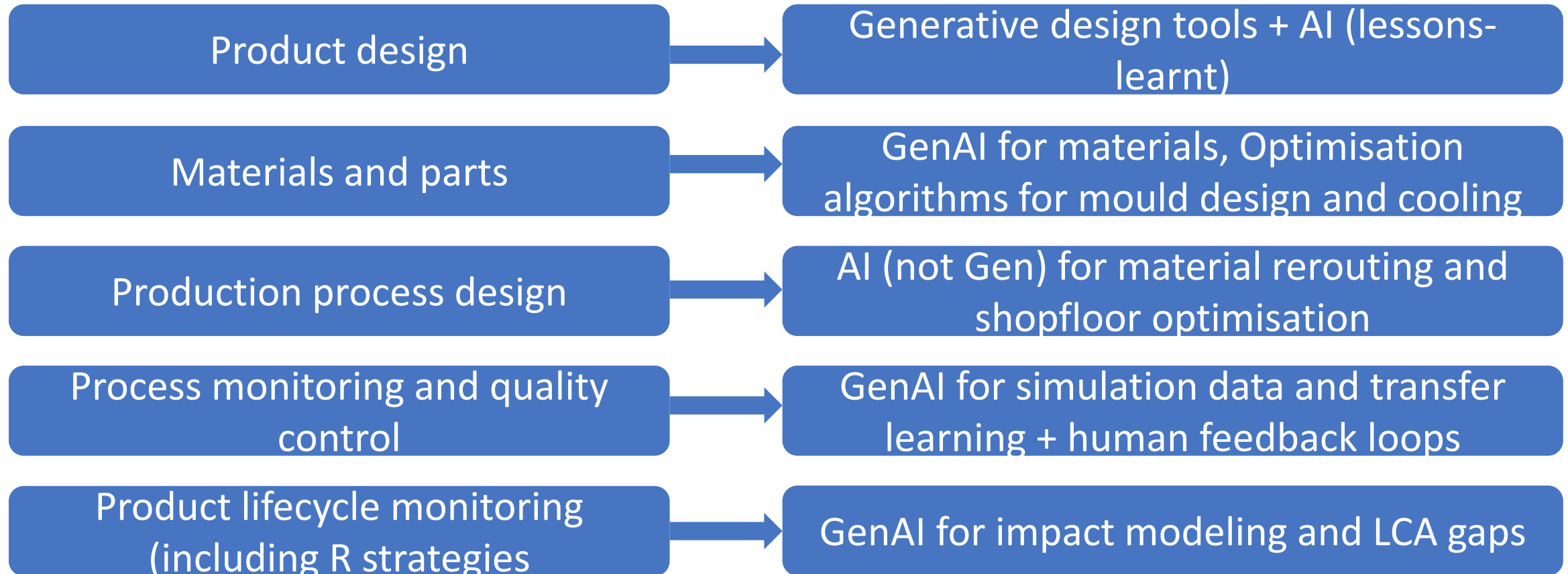
1. Developing **new processes and tooling** to produce better, in France
2. Implementing **digital solutions** for design, production and control
3. Creating **high-quality databases** and **secure their exchange** for a resilient industry
4. Deploying **new digital applications** to support efficient production

## Axe 3

High value-added products

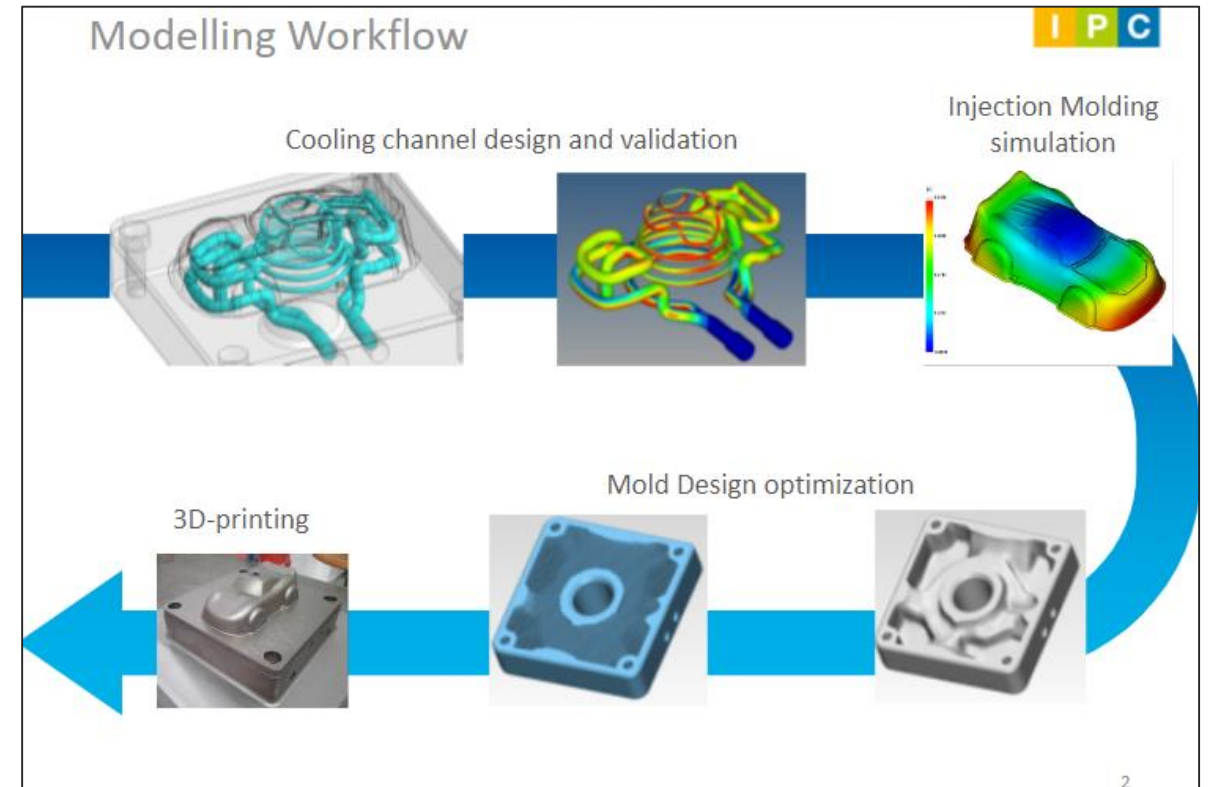
1. Improving the **life cycle of composites** from design to end-of-life
2. Enhance the technical, functional and environmental **performance of composites**
3. Deploy plastronics in an innovative and **responsible way**
4. Give **new properties and functionalities** to the surfaces and volumes of plastic object

# Overview of opportunities along the value chain



# Example 1: Gen AI for design optimisation

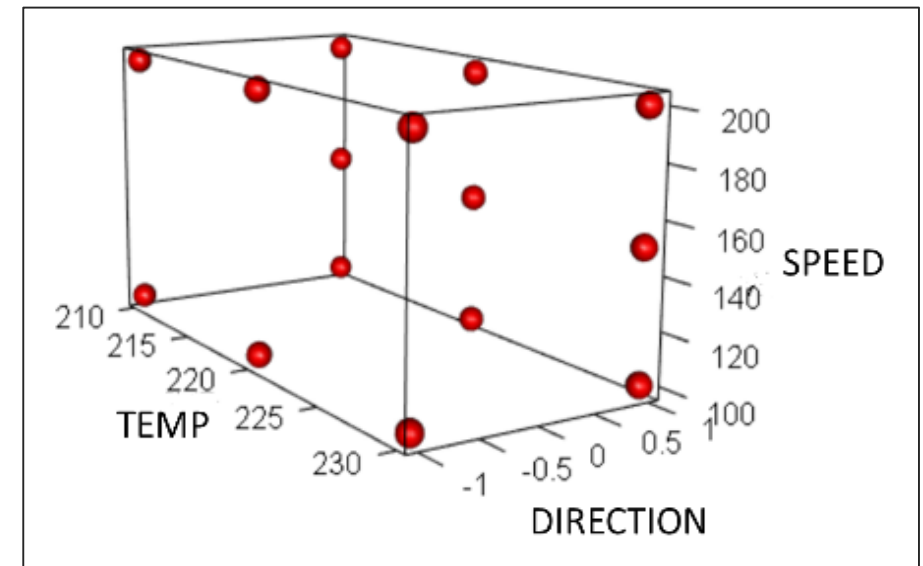
- Goal: use algorithms to optimize cool flow tubes around a mold
- Generate options based on simulation data and practical experience
- Test virtually and optimize flow rates depending on several parameters (temp, visc, geometry etc).



## Example 2: transfer learning for injection moulding

- Goal: optimize quality prediction and 'recipe' changes for process control
- Based on new sensors (IoT) and existing data on viscosity, temp and others (database IPC, moldflow) develop and train predictive model
- Digital DoE - generate data to optimize via NNs

Putting in place multiple DoEs

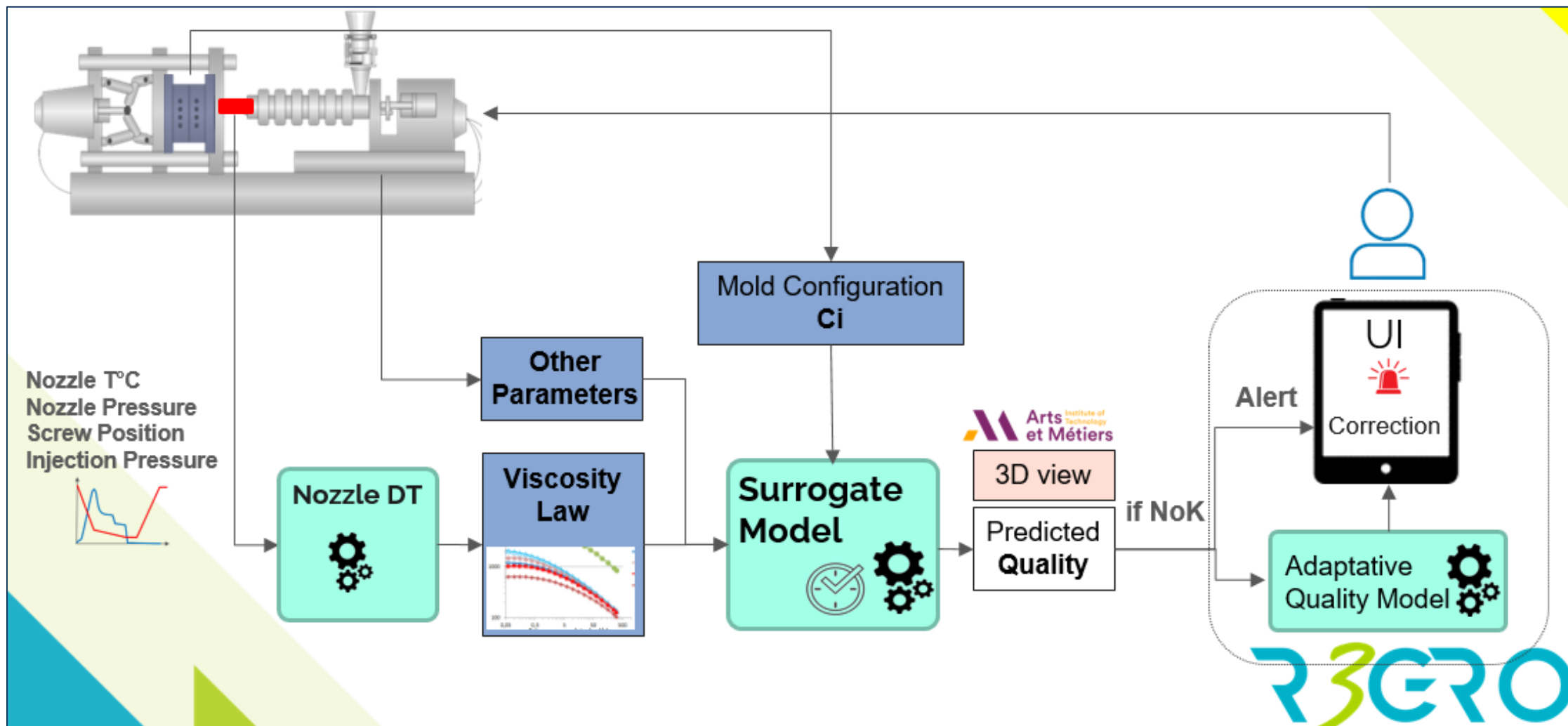


3 variables :

- Temperature
- Rotation speeds of the injection screw
- Material mix

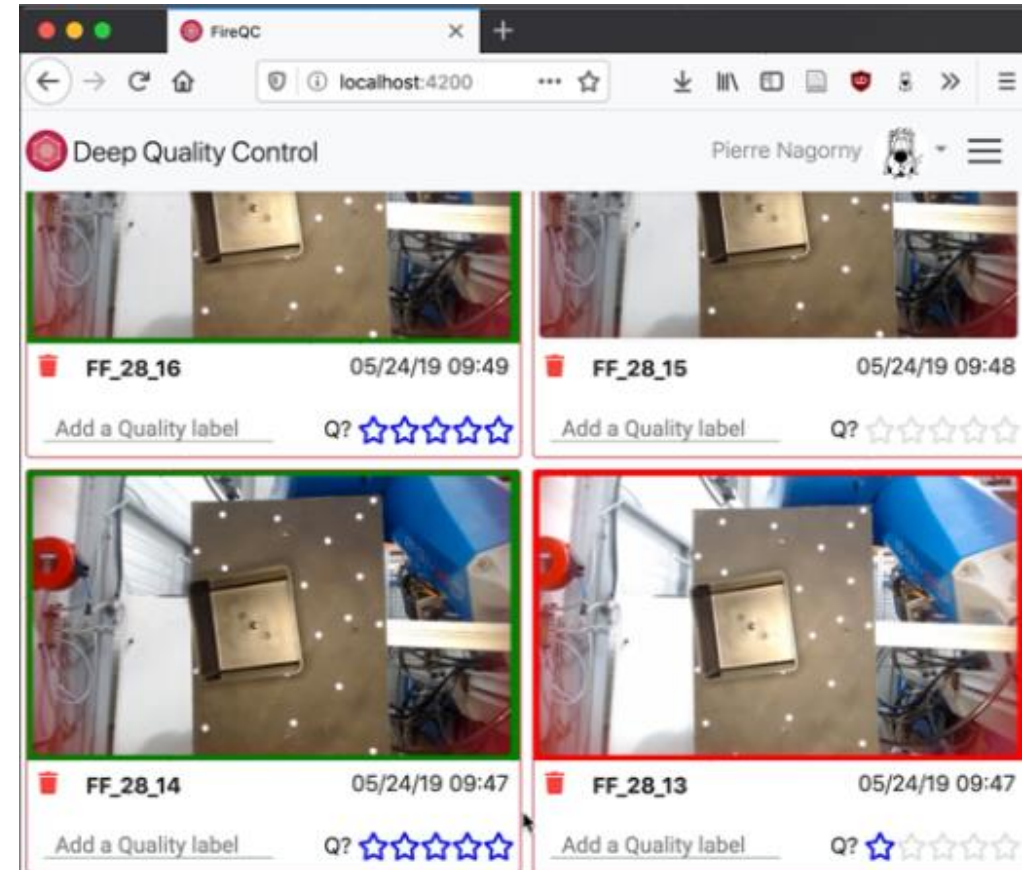
2 states :

- static
- dynamic



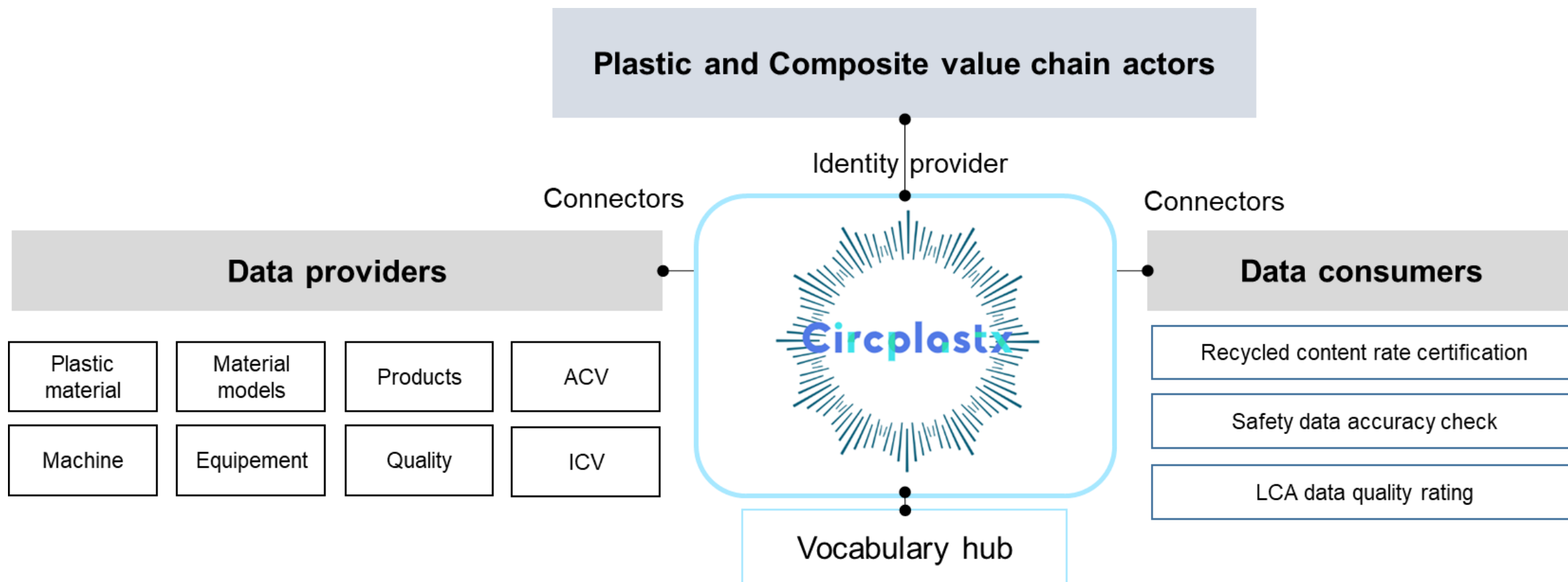
## Example 3: human in the loop

- Goal: improving quality inspection through human feedback
- GAN to generate thermal images
- Source of real thermal images
- Scoring of outcomes on quality inspection by algorithm – second layer of learning by humans in the loop
- Thesis project hosted at IPC
- Work continues in R3group project





# No AI without data...



# Limitations, Challenges and Outlook

## Limitations & Challenges

- Lack of data (but we have a data space for that!)
- Machine data still hard to work with (data act implementation?)
- No AI without understanding of the field (and hypotheses)
- Optimisation yes, discovery still low TRL

## Outlook

- Building high quality datasets
- DoEs for transfer learning
- LLM based process instructor
- Connection to actuators and closed loop production (difficult in some areas such as IM)