

The background features a dark blue gradient with large, overlapping, semi-transparent shapes in shades of purple, pink, and orange, creating a modern, abstract design.

# AWS re:Invent

NOV. 27 – DEC. 1, 2023 | LAS VEGAS, NV

MFG106

# Product innovation and customer engagement with Ferrari and Autodesk

## Alex Francois-Saint-Cyr

(she/her)

BD Head, Product  
Engineering & Development  
Amazon Web Services

## Mauro Coletto

(he/him)

Head of Business Analytics &  
Data Science  
Ferrari

## Heather Kerrick

(she/her)

Fusion Machine Learning  
Manager  
Autodesk

## Giovanni Longobardi

(he/him)

Cloud Operations Manager  
Ferrari



# Agenda

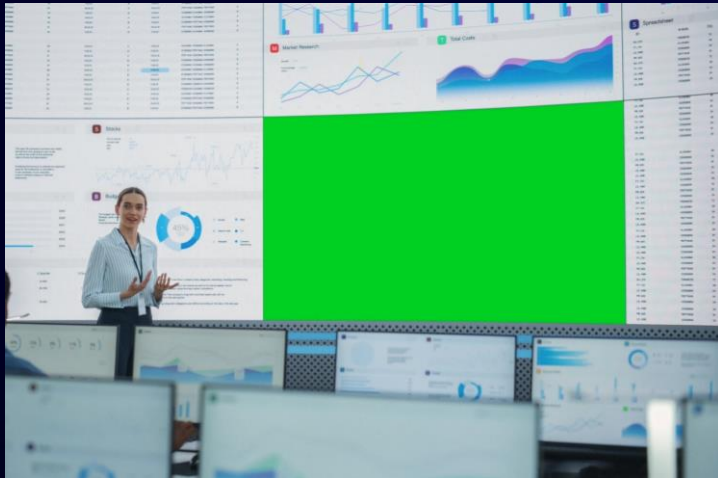
**01 Product engineering and development challenges**

**02 Embracing operational and product complexity**

**03 Hear it best from our customers:**

- Ferrari – Mauro Coletto and Giovanni Longobardi
- Autodesk – Heather Kerrick

# Product development challenges



**Performance and scalability**



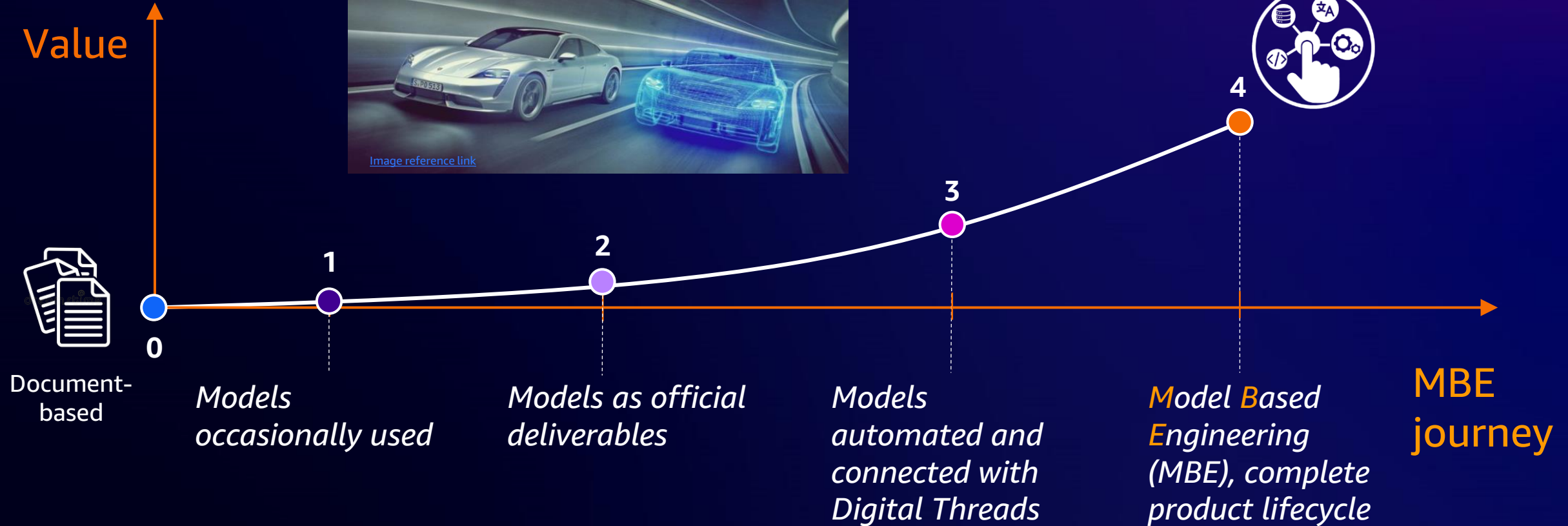
**Data friction**



**Global collaboration**

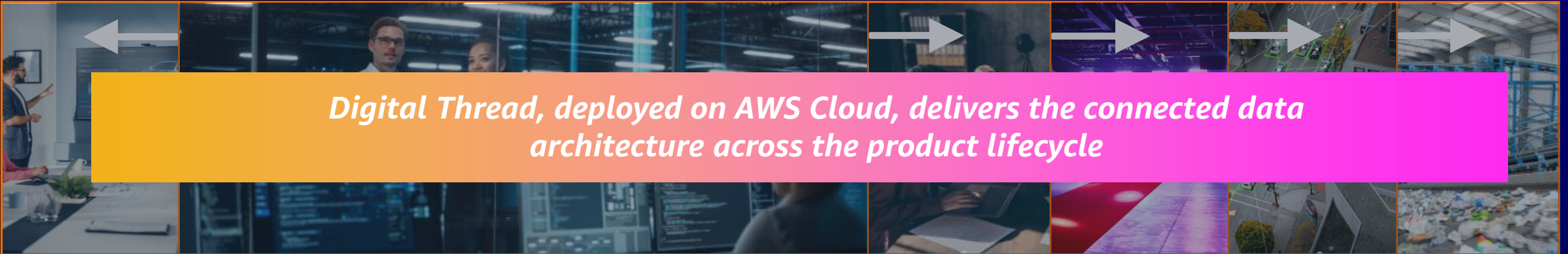
# Model-based engineering (MBE) journey

Where are you today?



# Product lifecycle

INCREASING COMPLEXITY

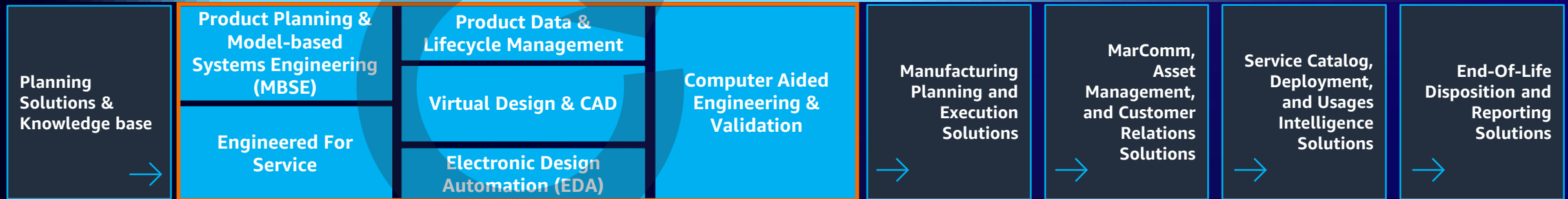


# Product engineering and development

EMBRACING COMPLEXITY



## Product Engineering Governance



### AWS Services (\*)

### AWS Solutions (\*)

### AWS Partners (\*)

Amazon AppStream		Amazon SageMaker		Research & Engineering Studio (RES)
NICE DCV		AWS IoT Core		Scale-Out Computing on AWS (SOCA)
Amazon S3		AWS Lambda		Twinflow
Amazon RDS		AWS ParallelCluster		


(\*) Sample List

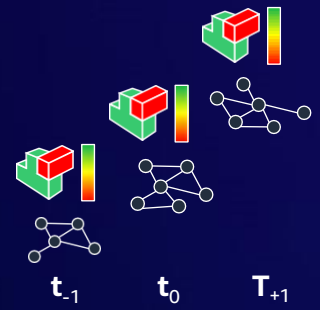


# Key technology enablers

Black box > 

White box > 

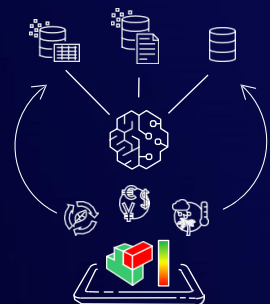
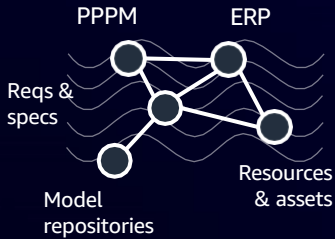
Grey box > 



**Secure global collaboration**

**High-performance computation (HPC)**

**End-user computing (RD/VDI)**



**Data lake/data mesh/  
Digital Thread**

**Automation &  
intelligence (AI/ML)**

**Spatial review/  
X-reality**



# Artificial intelligence in product engineering and development

## CAPABILITIES

- Speech to text conversion
- Advanced data search
- Ideation
- Design exploration
- Multidisciplinary optimization
- New materials



## BENEFITS



Productivity



Innovation



Quality



Manufacturability



Sustainability

# Ferrari

## **Mauro Coletto**

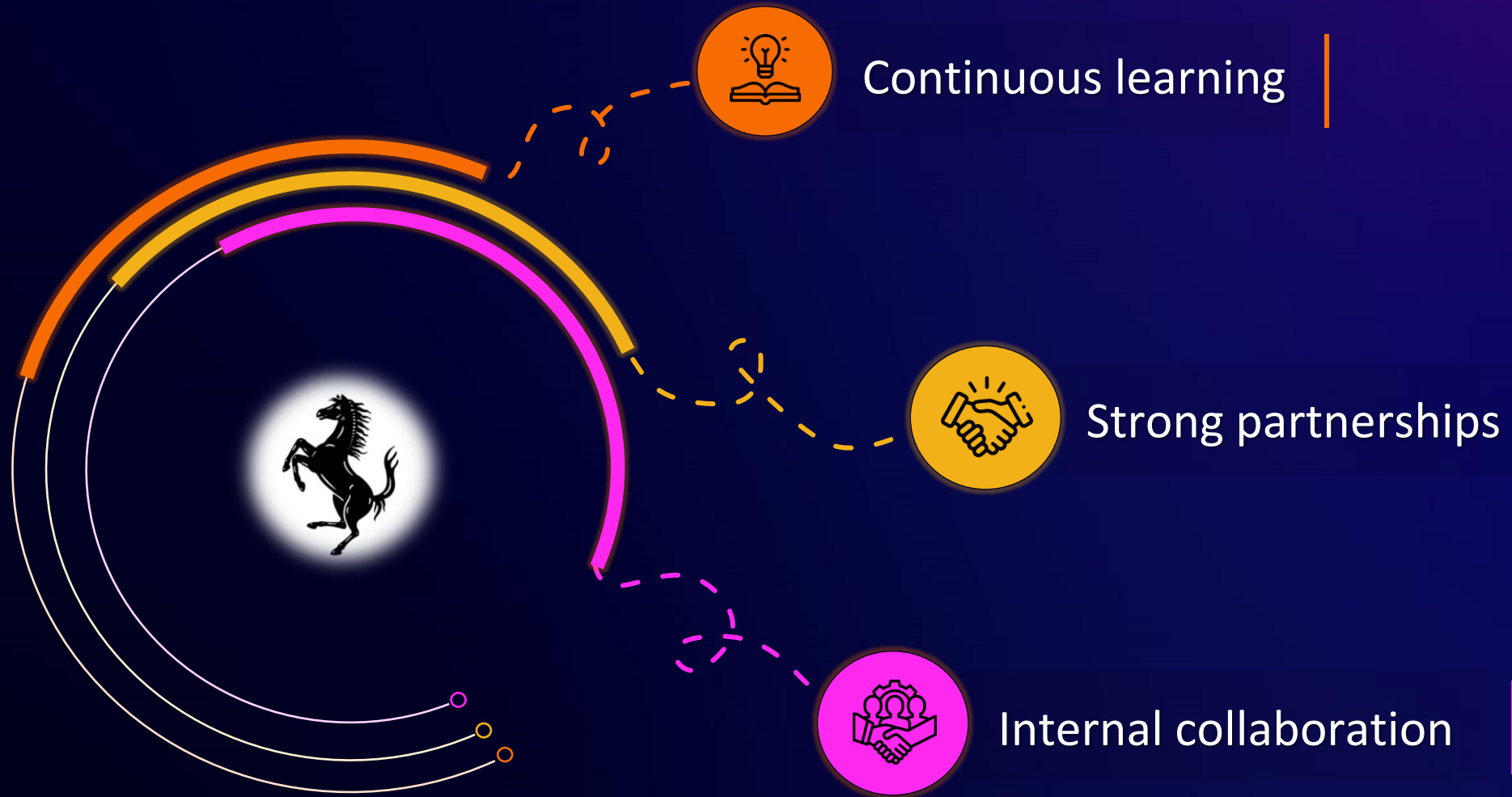
Head of Business Analytics  
and Data Science

## **Giovanni Longobardi**

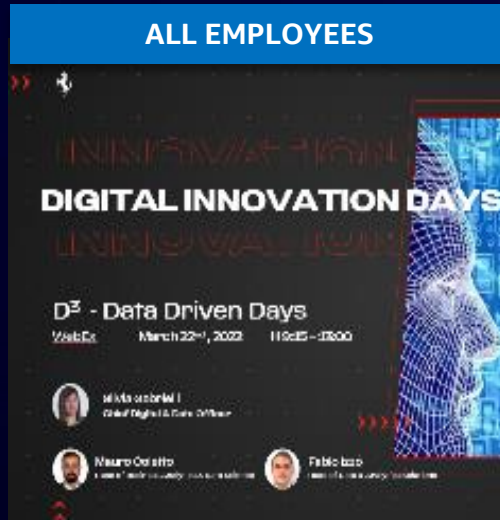
Cloud Operations Manager



# The guiding principles



# Continuous learning: Data-driven environment

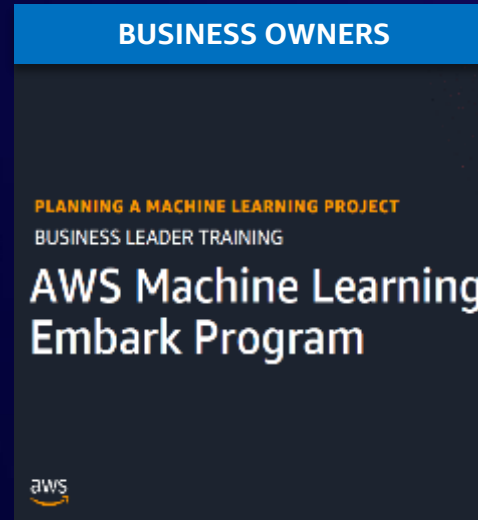


## DATA-DRIVEN DAYS

Inspirational and educational data-focused events open to all Ferrari employees

- Value of Data
- Machine Learning
- Data Visualization

**1700 participants**



## ML BUSINESS EMBARK PROGRAM

For R&D, manufacturing, communication, marketing, sales, and lifestyle business leaders

- How to drive an ML project
- Applied ML projects

**50 participants**



## ML TECH PROGRAM

Technical trainings

- Computer vision models
- Forecasting models
- Recommending systems
- Anomaly detection
- Predictive maintenance

# Partnership and strategic collaborations



# Collaboration: Building an enterprise data experts community

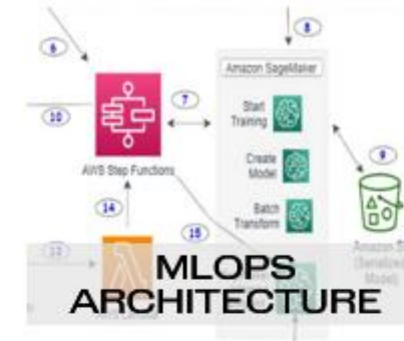


## FERRARI DATA SCIENCE HUB

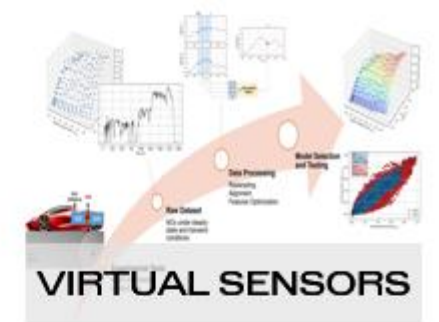
- Connect data analysts, ML engineers, and data scientists across all departments
- Networking and support
- Training and regular workshops



AWS AI/ML SERVICES



MLOPS ARCHITECTURE



VIRTUAL SENSORS



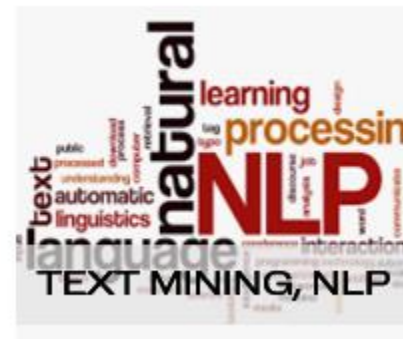
FORECASTING



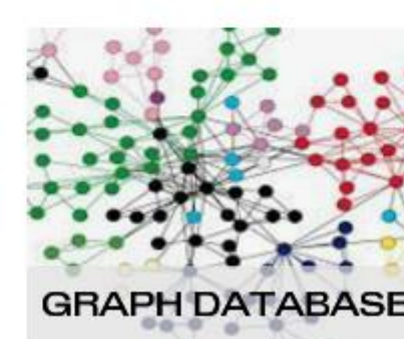
VIDEO DATA RETRIEVAL



COMPUTER VISION, OBJ DETECTION



TEXT MINING, NLP



GRAPH DATABASE



GENERATIVE AI

# The role of data at Ferrari



# Ferrari federated platforms and analytics tools

APPLICATIONS

BUSINESS INTELLIGENCE  
TOOLS



DATA SCIENCE LAB



Digital Manufacturing Platform

Vehicle Data Platform

Connected Car

Customer Data Platform & NBA

Digital Channels Analytics

DATA SOURCES

ERP

HR

PLM

MES

SAT

CRM

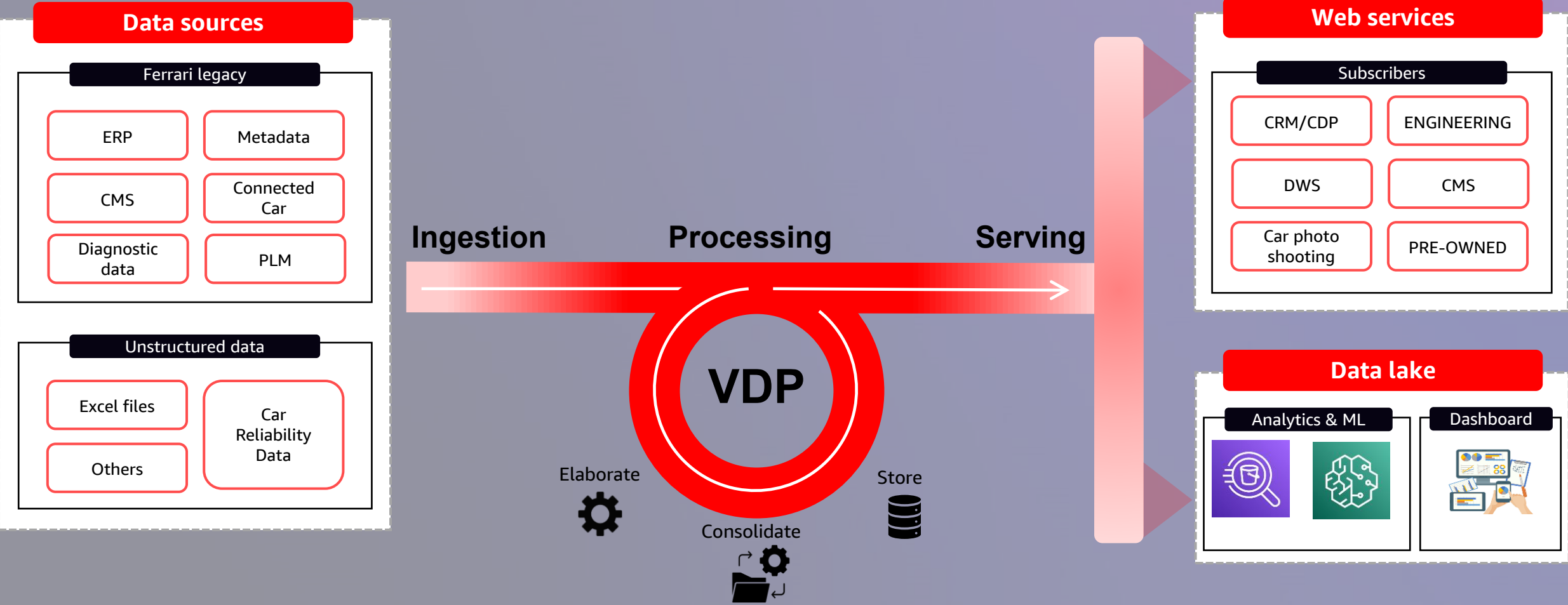
...



IOT



# Vehicle data platform infrastructure



# The role of CAE at Ferrari



# Why computer-aided engineering

## Research & Development and Digital & Data Synergy



Application support



HPC support



Enterprise architecture



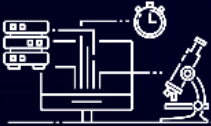
Operations & infrastructure

Enable engineers to virtually assess new car design and performance

## Business



CAE users



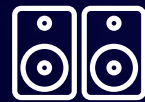
Optical simulation



FEM



CFD



Multibody



Electromagnetic simulation

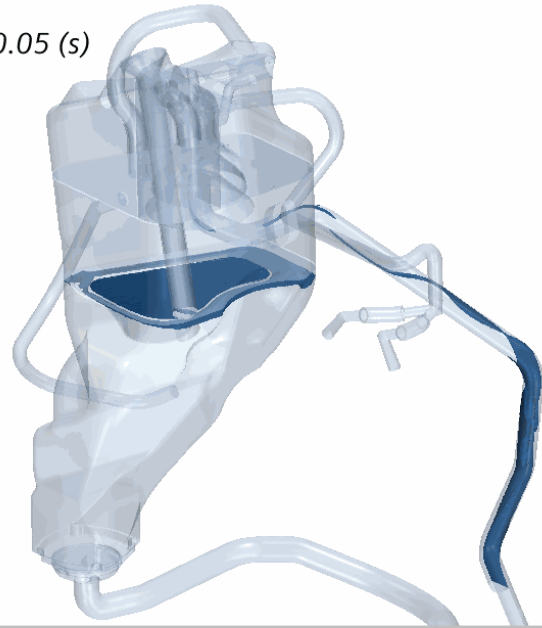
Improve scalability for peak management

Reduce computational time

Enhance user experience

# Virtual prototyping examples

Solution Time 0.05 (s)



## Simulations on cloud HPC

Structural

Computation fluid dynamics

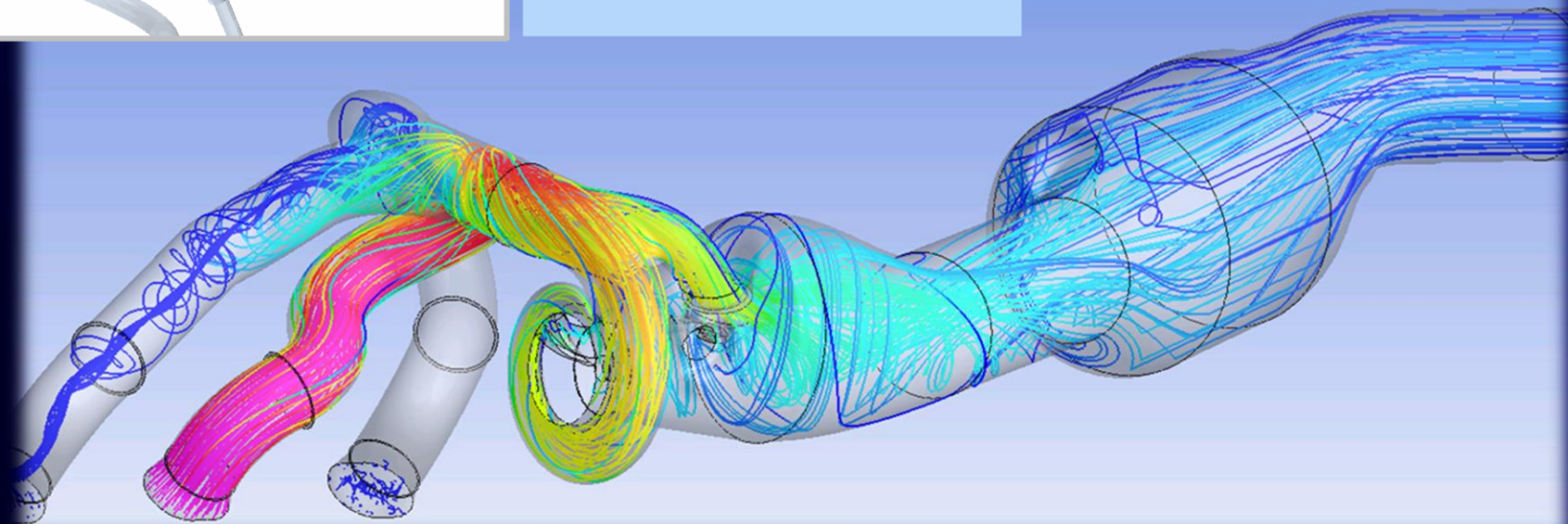
Electronic/electromechanical/thermal simulation

Multiphysic simulation

Optical simulation

Temperature  
Streamline 1

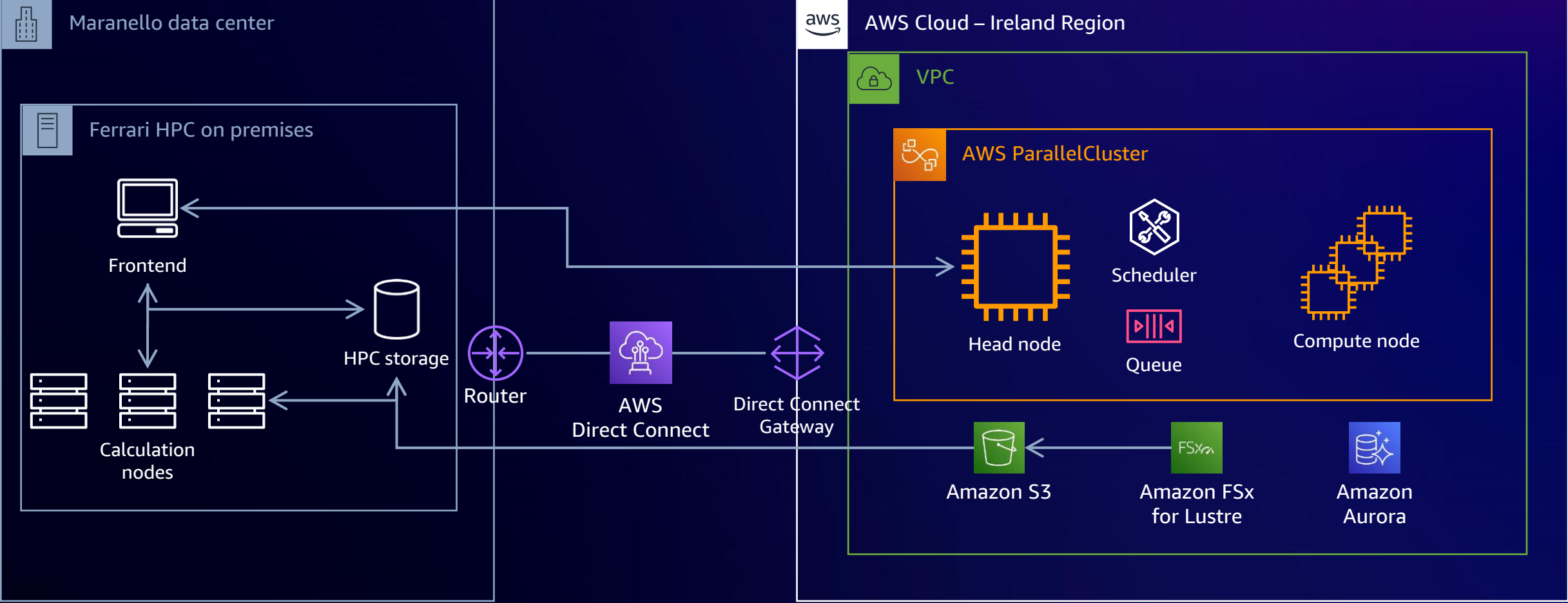
[K]



# Ferrari HPC hybrid architecture

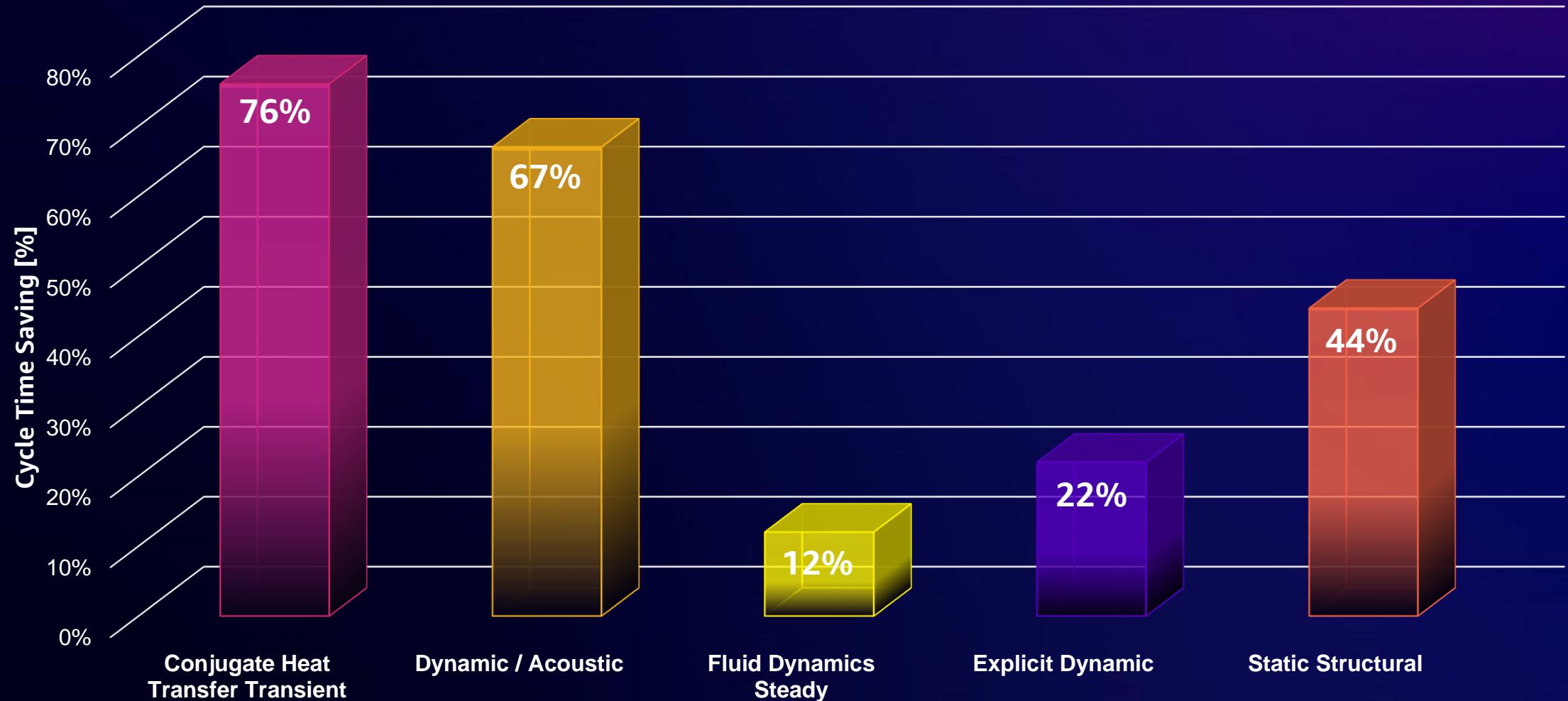


CAE users



# Benchmark results

## AWS HPC SIMULATION PERFORMANCE (TIME)



# Benchmark summary



# Benchmark summary



## Improvement

- 4x data transfer → Improved 75% of our use cases
- End user experience
- Scalability

# Benchmark summary



## Improvement

- 4x data transfer
- End user experience → Improved 91% of our use cases
- Scalability

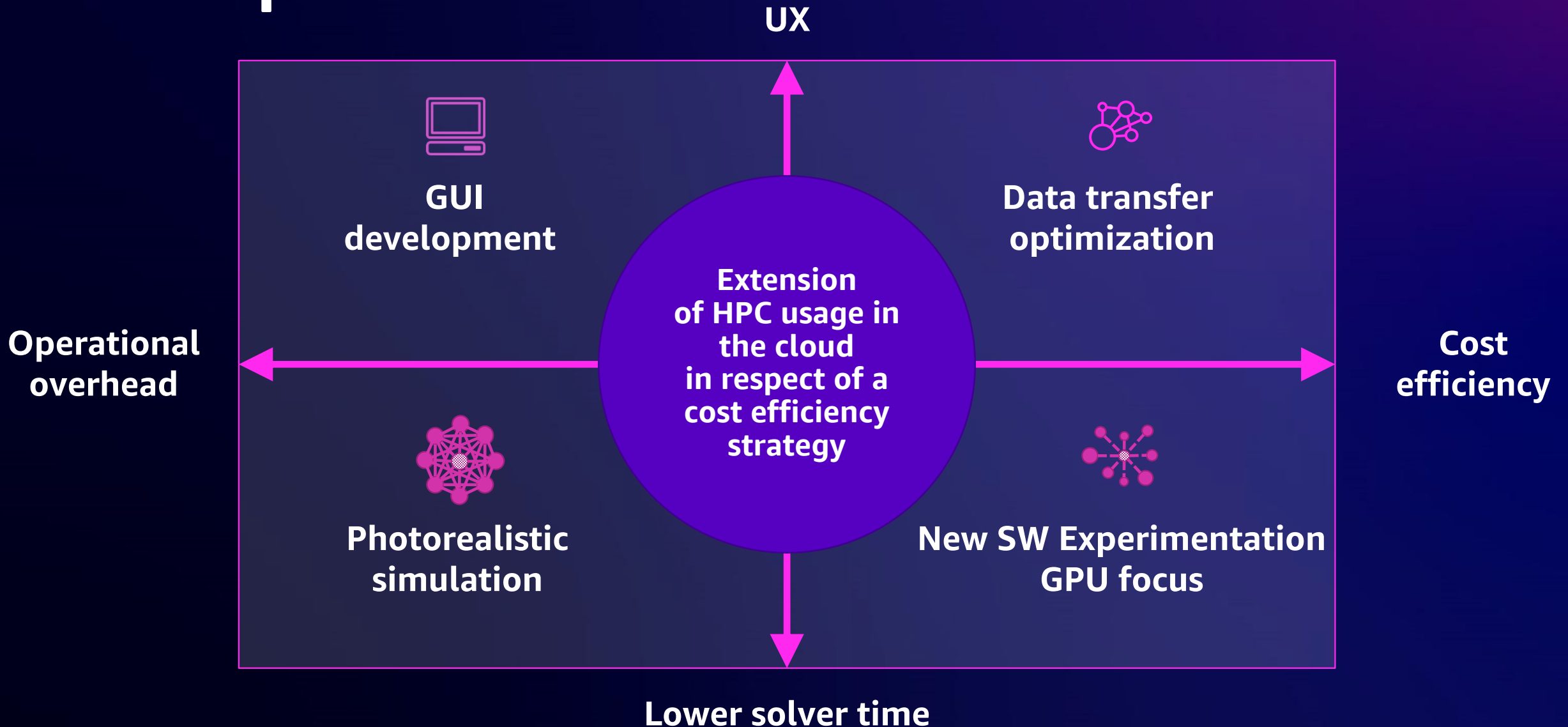
# Benchmark summary



## Improvement

- 4x data transfer
- End user experience
- **Scalability** → Improved 100% of our use cases

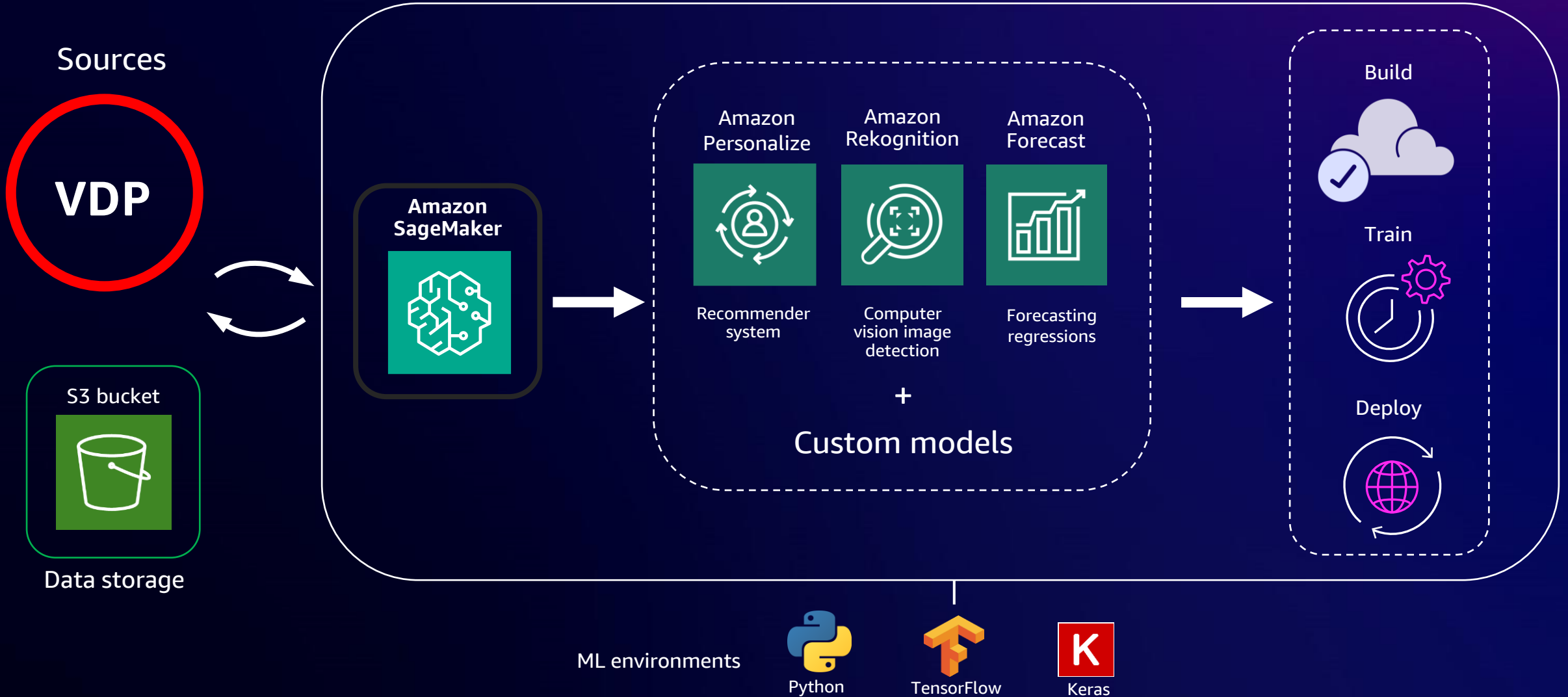
# Next steps



# Ferrari machine learning use cases

# Machine learning architecture

Workflow

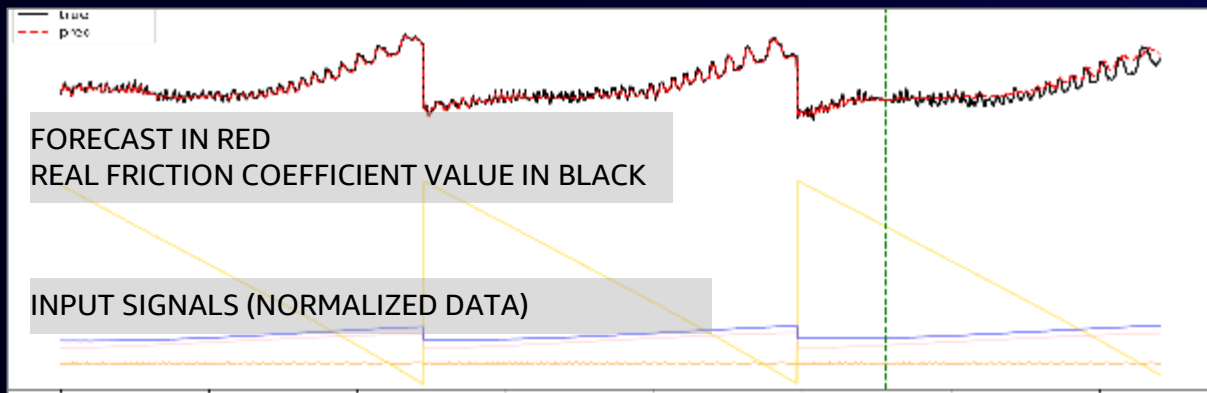
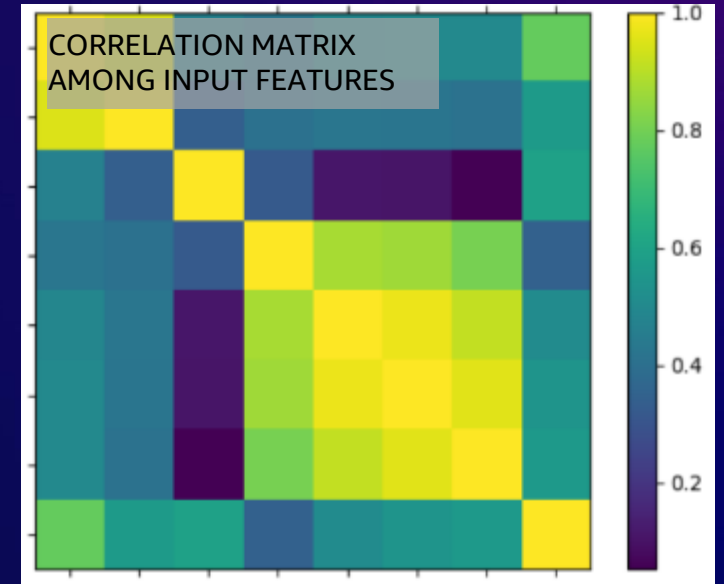


# ML use case example: Friction coefficient forecast

**Challenge:** Estimate the friction coefficient between disk and pad to optimize the vehicle dynamics

**Solution:** An ML model built on test benchmark data that can be used as a vehicle control system

- Must deal with **50Hz sampled data**
- Must be suitable for **general vehicle usage**
- Simulation data useful to build a more robust model
- Must be **deployable on the vehicle**



# Computer vision for content auto-labelling

**Goal:** Introduce an Automatic Tagger, with human supervision, to make unlabelled data from owned and earned media available for deeper analysis/indexing/consumption pipelines

## CAR MODEL CLASSIFICATION



Car Model:  
Ferrari Roma



Car Model:  
Ferrari 458  
Spider



Car Model:  
Ferrari  
Portofino



## PIPELINE

Custom labels



Celebrity recognition



Label detection



Face comparison



## OBJECT DETECTION



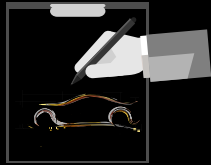
- Cars: 6
- People: 6
- Settings: **OUTDOOR**
- ...

# Generative AI: A potential game changer

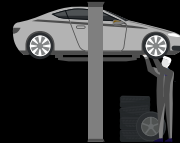
EXPERIMENTAL PHASE: LEVERAGE GENERATIVE AI TO CREATE WEIGHT- AND COST-OPTIMIZED PARTS

## Traditional approach

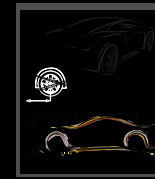
Part design takes weeks based on multiple iterative design and testing cycles conducted by engineers. No foundation models can currently perform this task. Algorithms available based on reinforcement learning.



Design



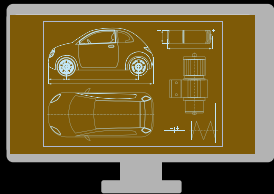
Testing



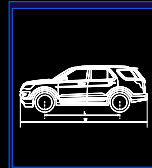
Adaptation



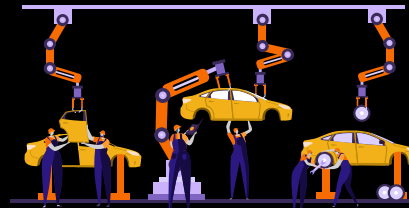
## Generative AI approach



Algorithm creates several models and automatically tests them against constraints



Best designs are further evolved until absolute best is found



Engineers review results and finalize the shape for production and physical testing

Value at stake of generative AI



2-3x quicker design process of complex parts, accelerating overall product development process

# Autodesk

## Heather Kerrick

(she/her)

Senior Manager, Fusion Machine Learning  
Autodesk



# Our software helps innovators design, build, and manufacture like never before

## Our industries

---

Architecture, engineering, and construction

---

Manufacturing

---

Media and entertainment



# Our company



Founded  
1982



Headquarters  
San Francisco, CA



Employees (2023)  
13,700+ worldwide

## In FY23\*

**\$5.01B**

Total revenue  
(up 14% year on year)

**6.74M**

Total subscribers  
(up 11.6% year on year)

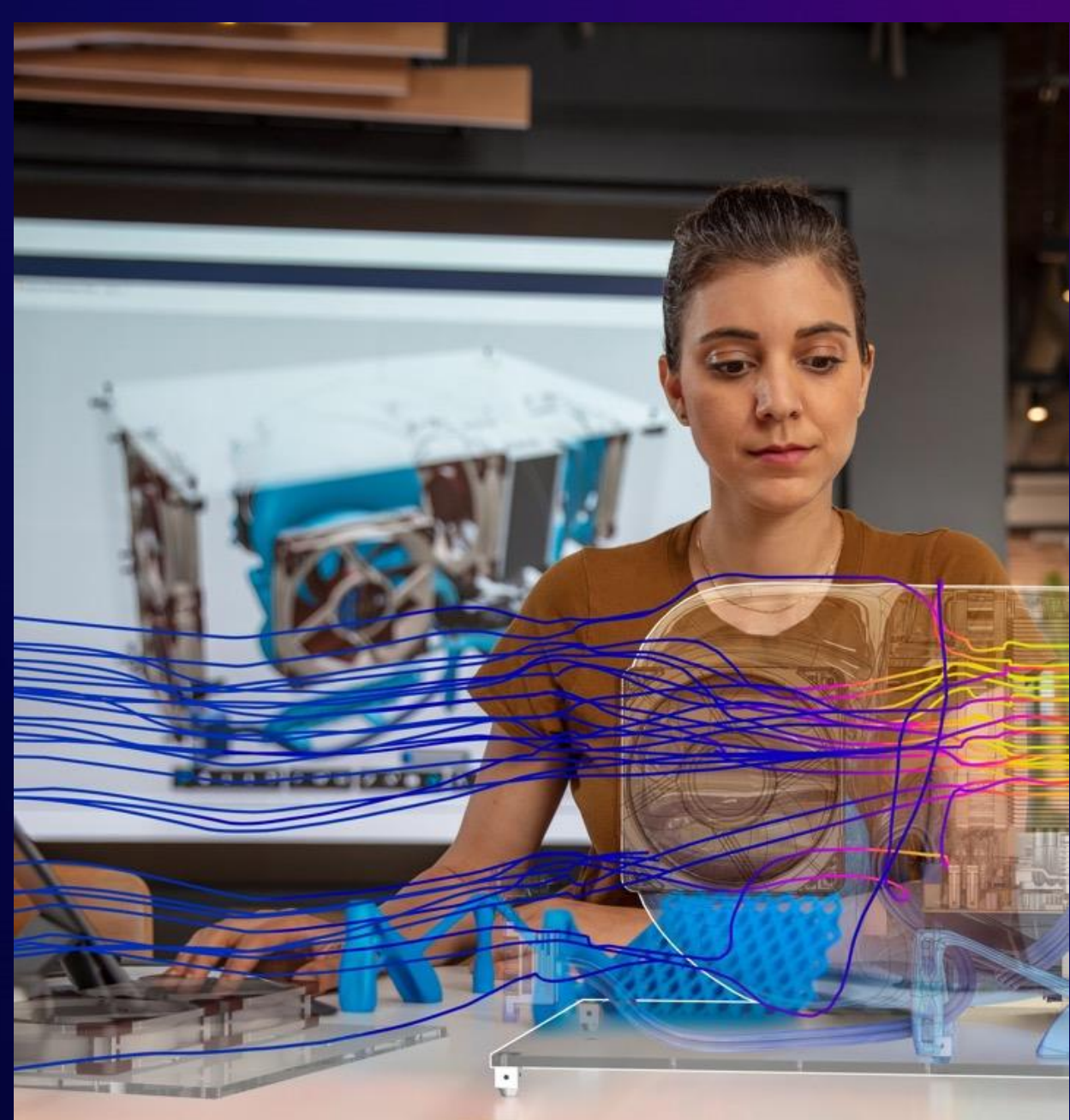
**\$1.2B**

R&D investment  
(20% of net revenue)

*\*Figures are from our fiscal year 2023, which began February 1, 2022 and ended January 31, 2023.*



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# Autodesk is changing how the world is designed and made

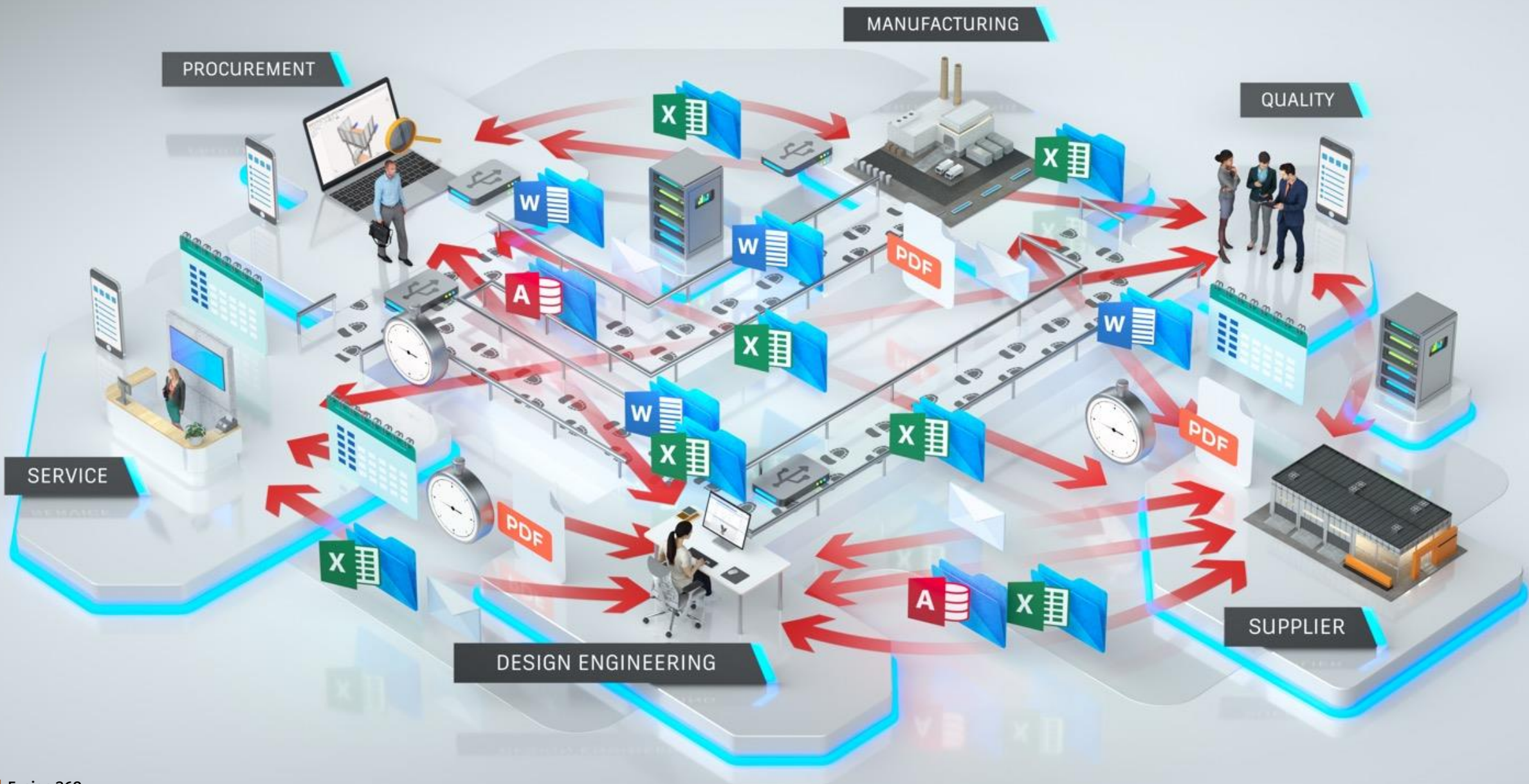




Increase size of wheel. Please design a new fender for the larger wheel



# Fusion





# AUTODESK AI

# Developer Requirements



**High performance**



**Secure**



**Reliable**



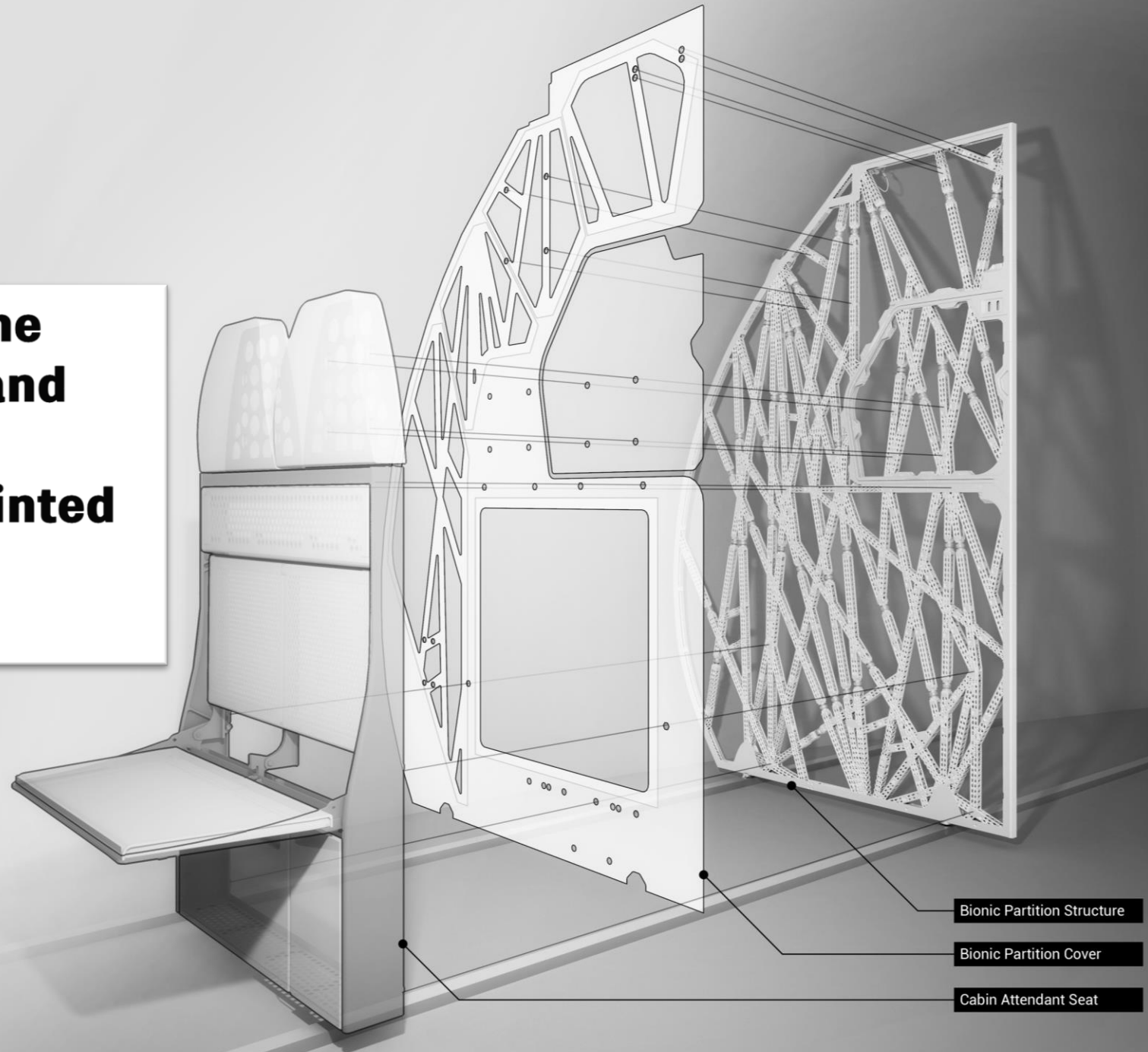
**Flexible**



**Scalable**

# Autodesk and Airbus show the future of aerospace design and manufacture in pioneering generatively designed 3D printed partition

1 DEC 2015



Bionic Partition Structure

Bionic Partition Cover

Cabin Attendant Seat





# Multiple generative design technologies

DIFFERENT TOOLS FOR DIFFERENT OBJECTIVES

Form and function exploration with Automated Modeling

Generative design for structural components

Generative design for fluid paths



# Multiple manufacturing methods

MANUFACTURING-READY DESIGN

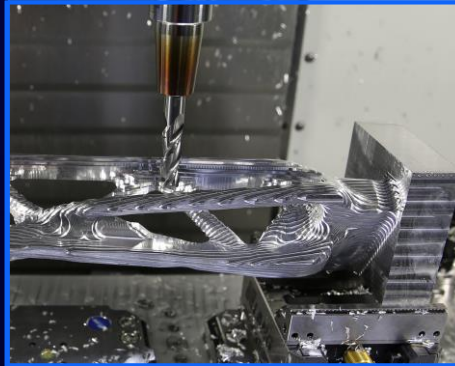
Additive  
manufacturing



Die casting



3 & 5 axis milling

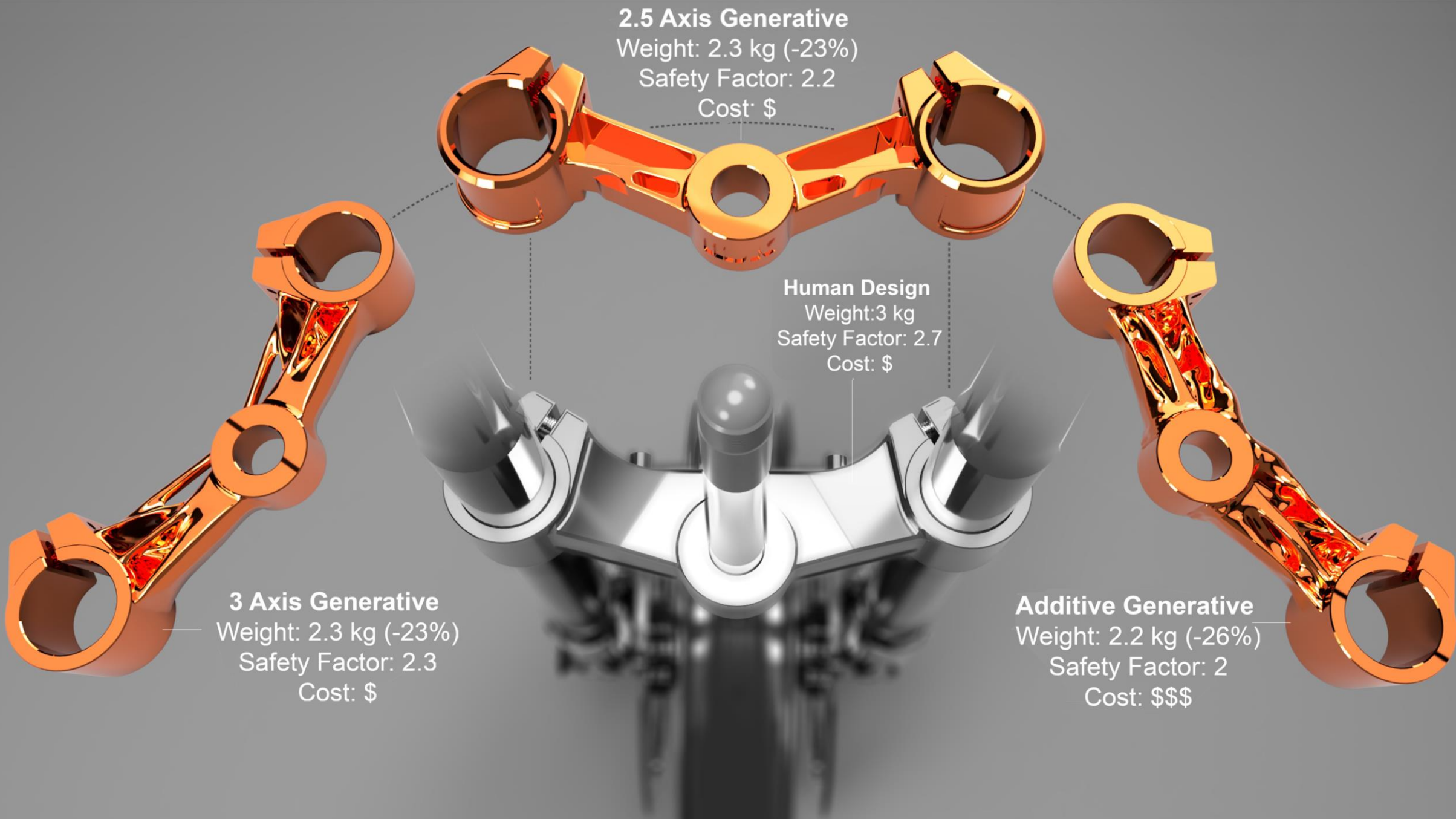


2.5 axis milling



2 axis cutting





**2.5 Axis Generative**  
Weight: 2.3 kg (-23%)  
Safety Factor: 2.2  
Cost: \$

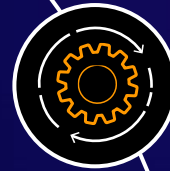
**Human Design**  
Weight: 3 kg  
Safety Factor: 2.7  
Cost: \$

**3 Axis Generative**  
Weight: 2.3 kg (-23%)  
Safety Factor: 2.3  
Cost: \$

**Additive Generative**  
Weight: 2.2 kg (-26%)  
Safety Factor: 2  
Cost: \$\$\$

# Performance Improvements Summary

GENERATIVE DESIGN FOR  
STRUCTURAL COMPONENTS



**1.7x**  
Faster solve times



**6**  
fully supported  
MFG Methods



**22%**  
increase in  
converged studies



**50%**  
reductions in  
failed studies





DESIGN

SOLID SURFACE MESH FORM SHEET METAL PLASTIC UTILITIES MANAGE

CREATE

AUTOMATE

MODIFY

ASSEMBLE

CONFIGURE

CONSTRUCT

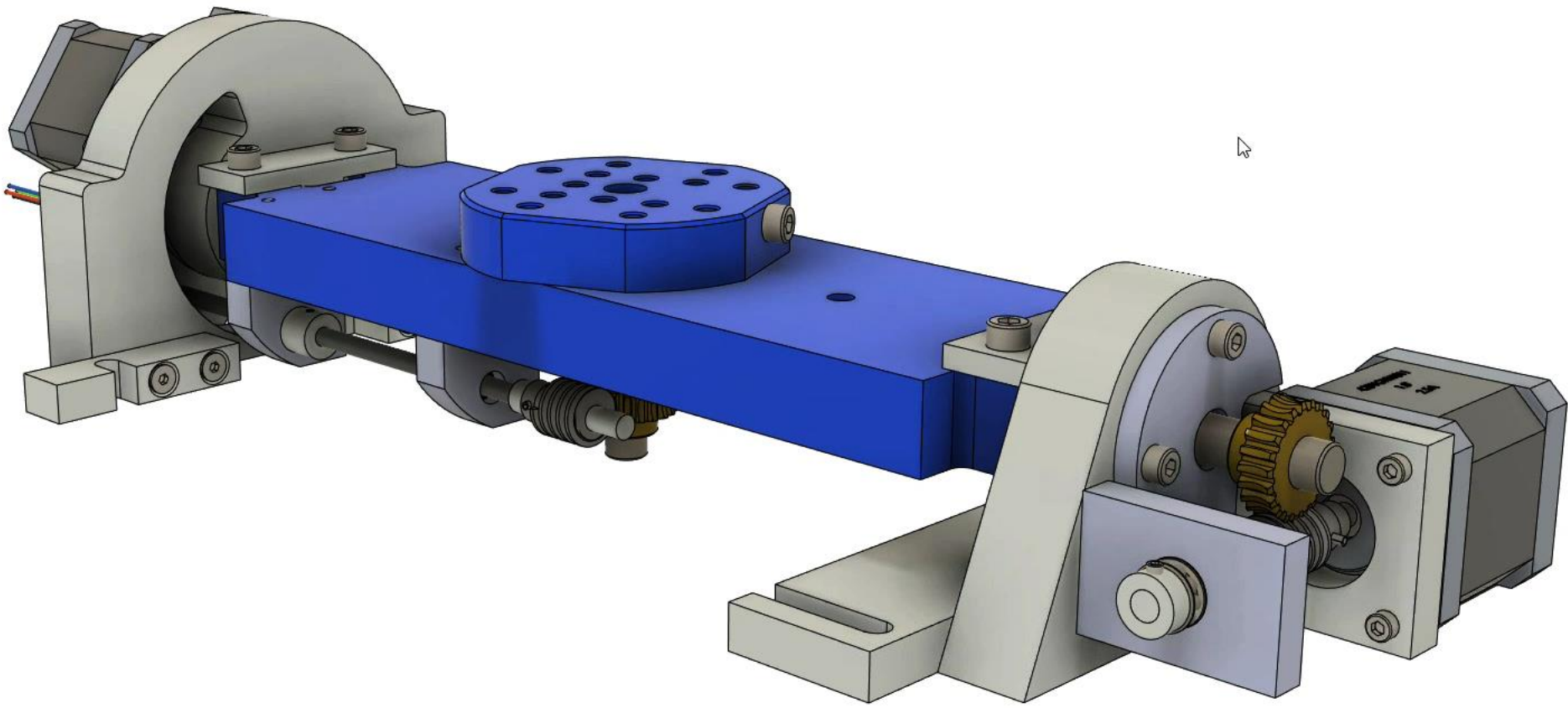
INSPECT

INSERT

SELECT

BROWSER

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  - TischACHalterR:1
  - CAchseBG:1
    - Origin
    - CWelle:1
    - AufspannPlatte:1
    - Stelling:1
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    - socket set screw cone point\_d...
    - 320\_007\_00\_schneckenraedel...
  - SAntriebA Gear:1
    - Origin
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    - NemaHalterL:1
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    - socket head cap screw\_din\_D...
    - socket head cap screw\_din\_D...
    - 320\_107\_00\_schnecken\_a\_17...
    - AxialLager13x6x5 bearing:1
    - AxialLager13x6x5 bearing:2
    - StellRing6:1
    - StellRing6:2
    - AAntriebsWelleZapfen:1
    - LagerHalterA:1
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    - socket head cap screw\_din\_D...
    - socket set screw cone point\_d...



COMMENTS

Navigation icons: Home, Back, Forward, Search, Zoom, Rotate, etc.





# Autodesk Machine Learning Platform

ENABLE MACHINE LEARNING  
DEVELOPMENT AND DEPLOYMENT



**Data**



**Train**



**Deploy**



**Monitor**

## THE PLATFORM

Help engineers and researchers leverage data; develop and train new models; deploy, monitor and retrain models in production

# Autodesk Machine Learning Platform

## THE VALUE

- Off-the-shelf ML models easier to use
- Access to increasingly powerful compute
- Train without copying data
- Control what data is where & who has access

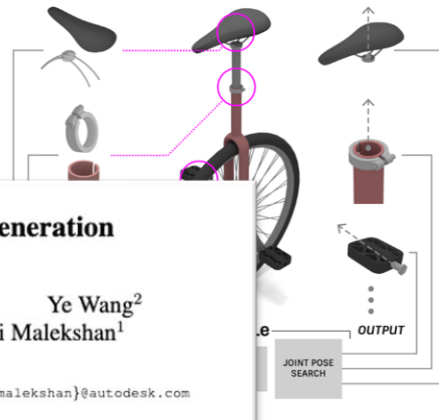
# Advancing the state of the art

## JoinABLE: Learning Bottom-up Assembly of Parametric CAD Joints

Karl D.D. Willis<sup>1</sup> Pradeep Kumar Jayaraman<sup>1</sup> Hang Chu<sup>1</sup> Yunsheng Tian<sup>2</sup> Yifei Li<sup>2</sup>  
 Daniele Grandi<sup>1</sup> Aditya Sanghi<sup>1</sup> Linh Tran<sup>1</sup> Joseph G. Lambourne<sup>1</sup>  
 Armando Solar-Lezama<sup>2</sup> Wojciech Matusik<sup>2</sup>  
<sup>1</sup>Autodesk Research <sup>2</sup>MIT CSAIL

### Abstract

Physical products are often complex assemblies combining a multitude of 3D parts modeled in computer-aided design (CAD) software. CAD designers build up these assemblies by aligning individual parts to one another using constraints called joints. In this paper we introduce JoinABLE, a learning-based method that assembles parts together to



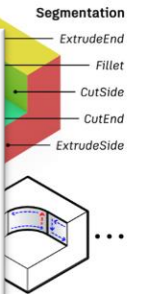
## BRepNet: A topological message passing system for solid models

Joseph G. Lambourne Autodesk Research Karl D.D. Willis Autodesk Research Pradeep Kumar Jayaraman Autodesk Research  
 Aditya Sanghi Autodesk Research Peter Meltzer UCL, Computer Science Hooman Shayani Autodesk Research

### Abstract

## Fusion 360 Gallery: A Dataset and Environment for Programmatic CAD Construction from Human Design Sequences

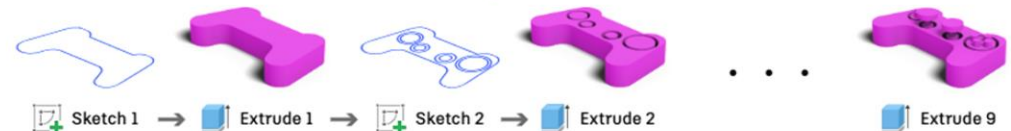
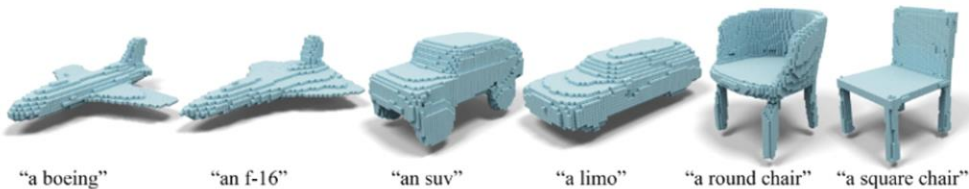
KARL D.D. WILLIS, Autodesk Research, USA  
 YEWEN PU, Autodesk Research, USA  
 JIELIANG LUO, Autodesk Research, USA  
 HANG CHU, Autodesk Research, Canada  
 TAO DU, Massachusetts Institute of Technology, USA  
 JOSEPH G. LAMBOURNE, Autodesk Research, United Kingdom  
 ARMANDO SOLAR-LEZAMA, Massachusetts Institute of Technology, USA  
 WOJCIECH MATUSIK, Massachusetts Institute of Technology, USA



## CLIP-Forge: Towards Zero-Shot Text-to-Shape Generation

Aditya Sanghi<sup>1</sup> Hang Chu<sup>1</sup> Joseph G. Lambourne<sup>1</sup> Ye Wang<sup>2</sup>  
 Chin-Yi Cheng<sup>1</sup> Marco Fumero<sup>1</sup> Kamal Rahimi Malekshan<sup>1</sup>  
<sup>1</sup>Autodesk AI Lab <sup>2</sup>Autodesk Research

{aditya.sanghi, hang.chu, joseph.lambourne, ye.wang, chin-yi.cheng, marco.fumero, kamal.malekshan}@autodesk.com



# Generative AI



## Concept Generator

Concept

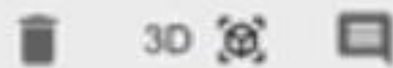
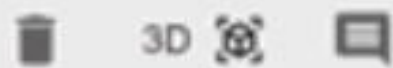
Dynamic



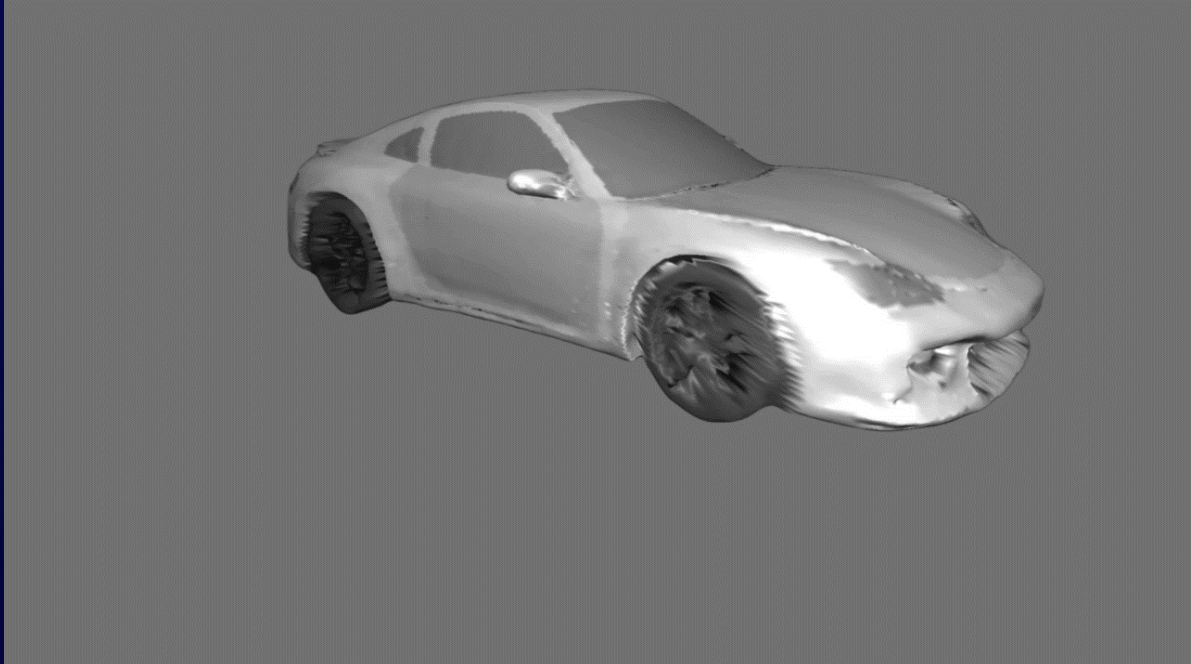
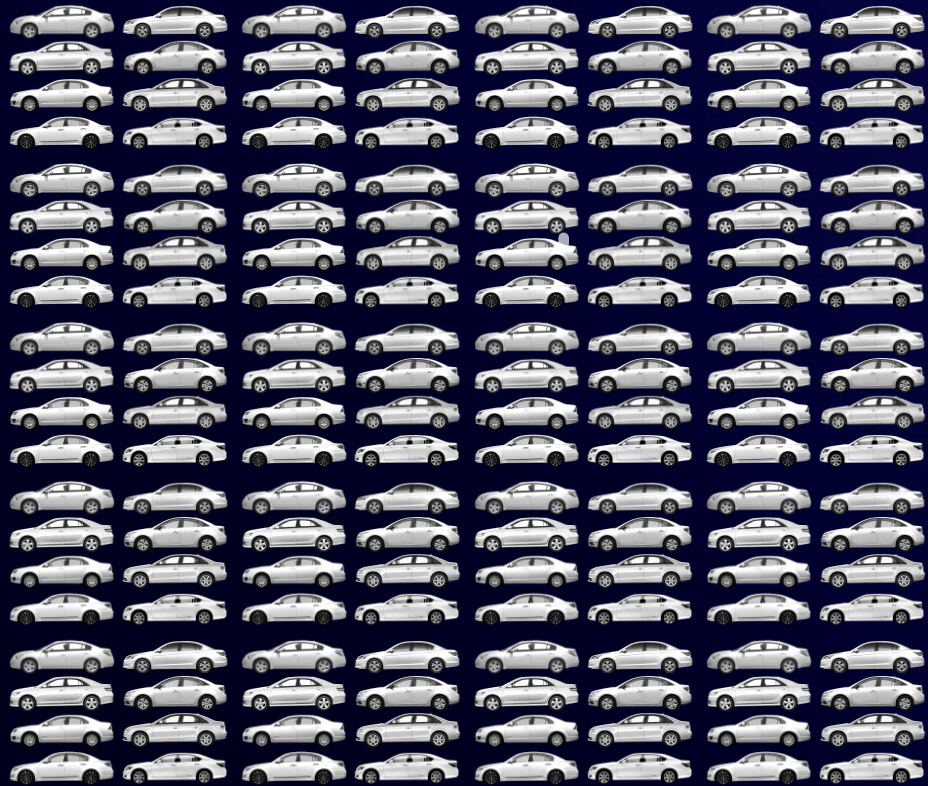
ADD NEW CONCEPT



Designs



# Blank AI



catalog

▼ TUNE +

Type ⋮

Coupe ▼

0.0

▼ BLEND +

Attributes

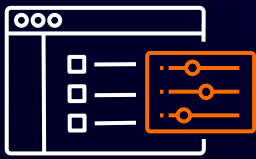
- Attributes ← ● →
- Dynamic
  - Modern
  - Bold
  - Simple
  - Sporty
  - Unique
  - Refined
  - Boxy
  - Elegant
  - Rugged
  - Aggressive



The logo features the Autodesk 'A' icon on the left, followed by the text 'AUTODESK AI' in a bold, white, sans-serif font. The 'AI' is rendered in a hollow, outlined style. The entire logo is centered horizontally against a dark blue background with a vertical gradient bar on the left side.

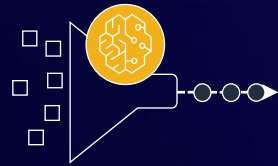
# AUTODESK AI

# Accelerating product engineering and development with AWS



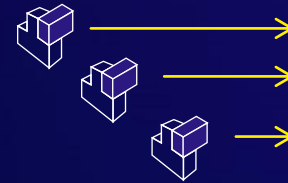
## Time to market

Reduce time to market through effective digital thread throughout value-chain and across enterprises



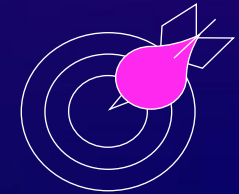
## Innovation pipeline

Increased ability to iterate on alternatives through efficient digitalization and smart digital validation



## Product and engineering capacity

Improved throughput through collaborative engineering, knowledge reuse, effective digital thread



## Quality, iteration, and rework

Reduce design iterations through cross-functional collaboration and new technologies like AI/ML

# Thank you!



Please complete the session survey in the mobile app

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