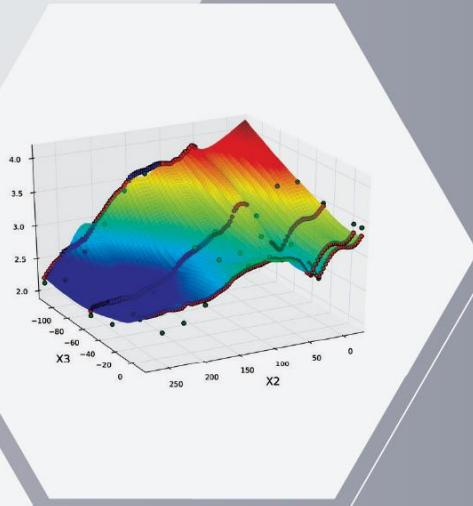
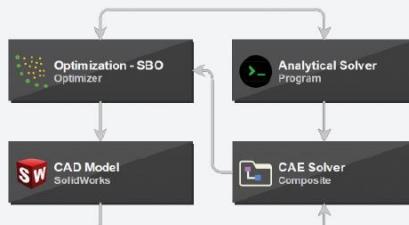




# EXPLORE AND OPTIMIZE YOUR DESIGNS



## pSeven platform

From research in algorithms  
to large scale industrial  
applications

DATADVANCE

April 2021

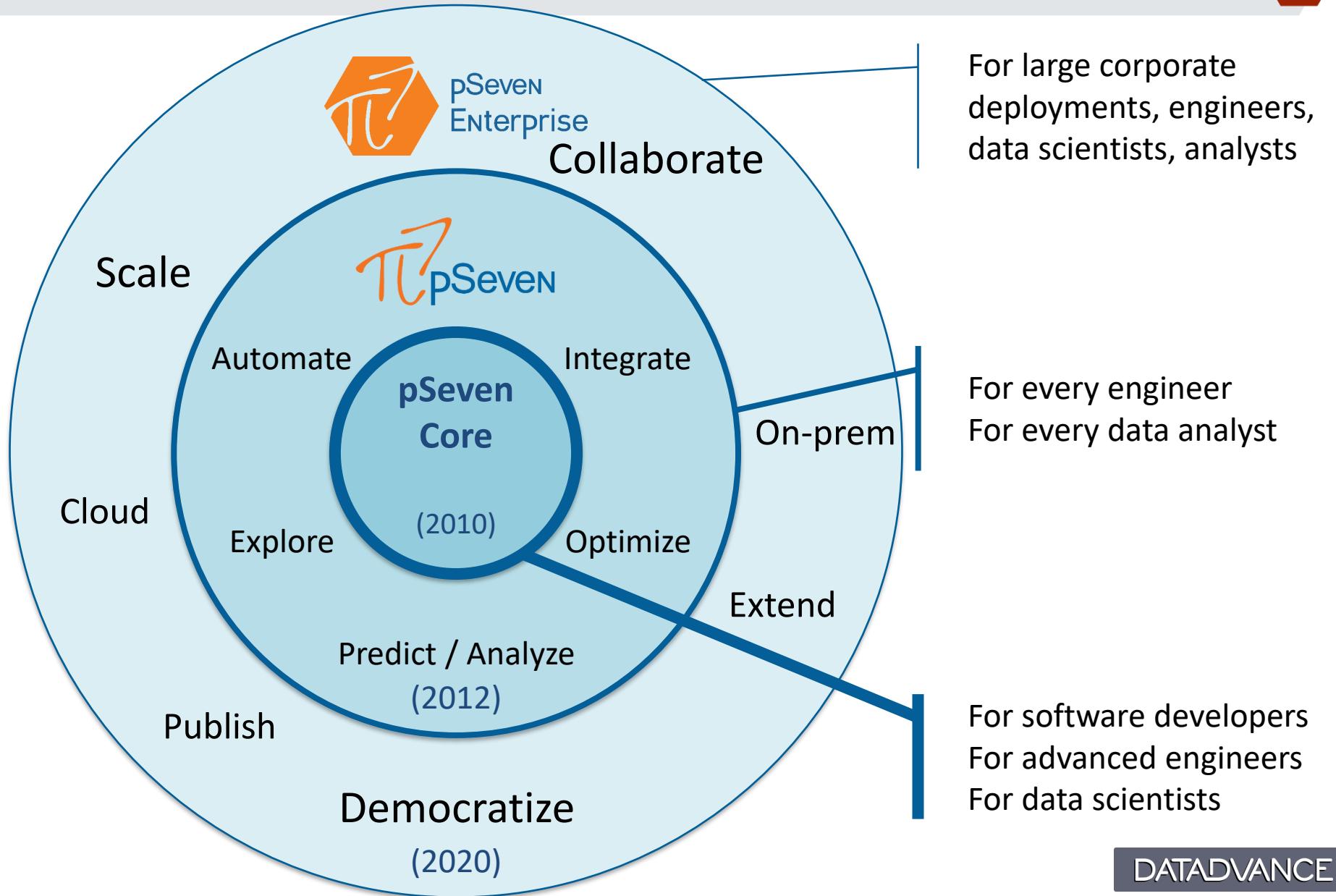
DATADVANCE



# About DATADVANCE

- We develop a low-code web-based software platform for predictive modeling and optimization.
- We serve customers from industries like aerospace, automotive, energy, marine and offshore and related industries.
- Our roots go to French aviation and Russian mathematics. We evolved out of a joint research program between
  - [Airbus](#) – a global leader in aerospace and defense industry
  - [Institute for Information Transmission Problems](#) – one of the leading mathematical centers in Russia with three Fields prize winners on the staff
- We were incorporated in 2010.
- We are located at
  - Toulouse, France (HQ and Operations)
  - Moscow, Russia (R&D)
- R&D team of 40+ highly skilled researchers, data scientists, software developers and application engineers, with 8 PhDs among them.
- World-wide sales through a network of Value Added Resellers.





# pSeven Core, pSeven and pSeven Enterprise users

Aerospace



Automotive & Heavy Machinery



Energy & Turbomachinery



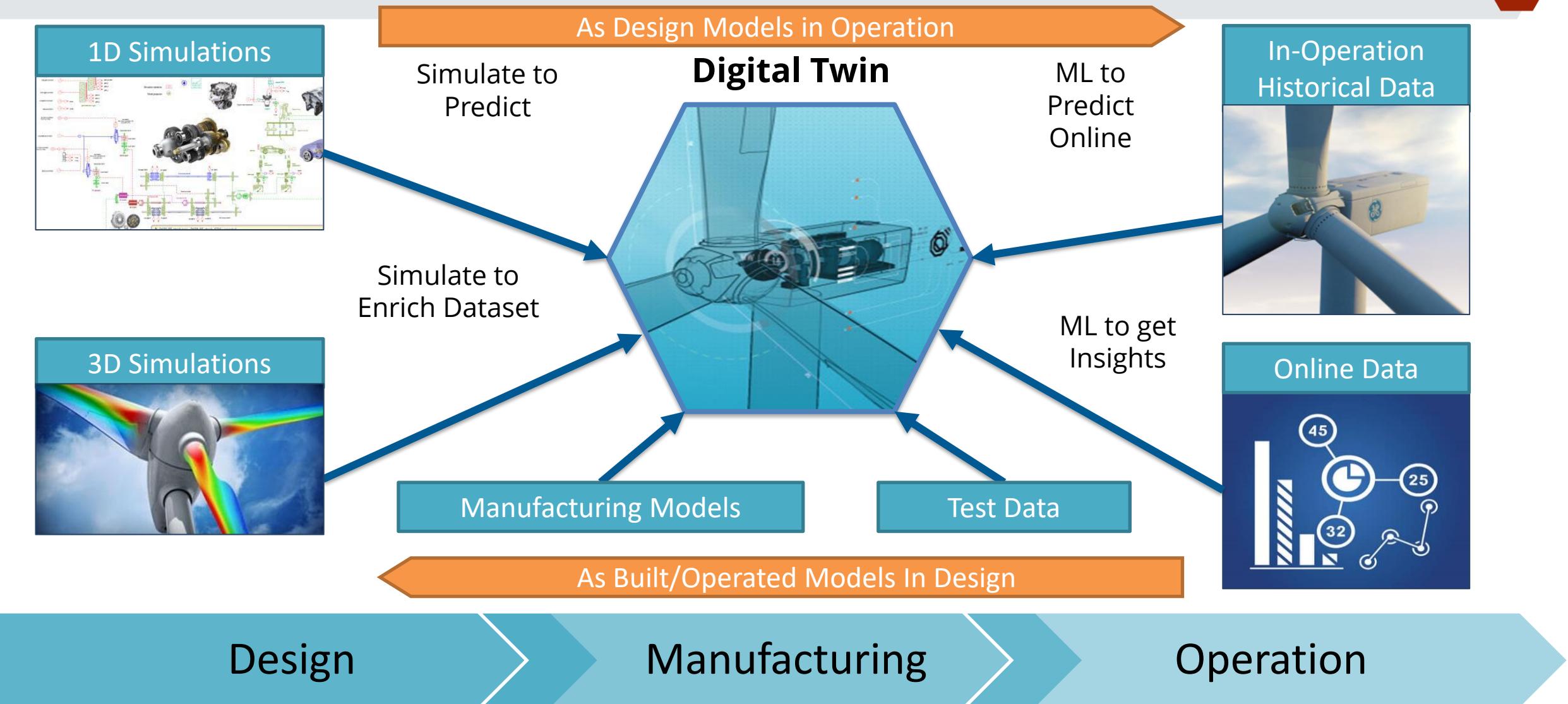
Shipbuilding & Offshore



Life Sciences & Other



# Digital Twin: Set of Simulation and ML Models Associated with Real Equipment



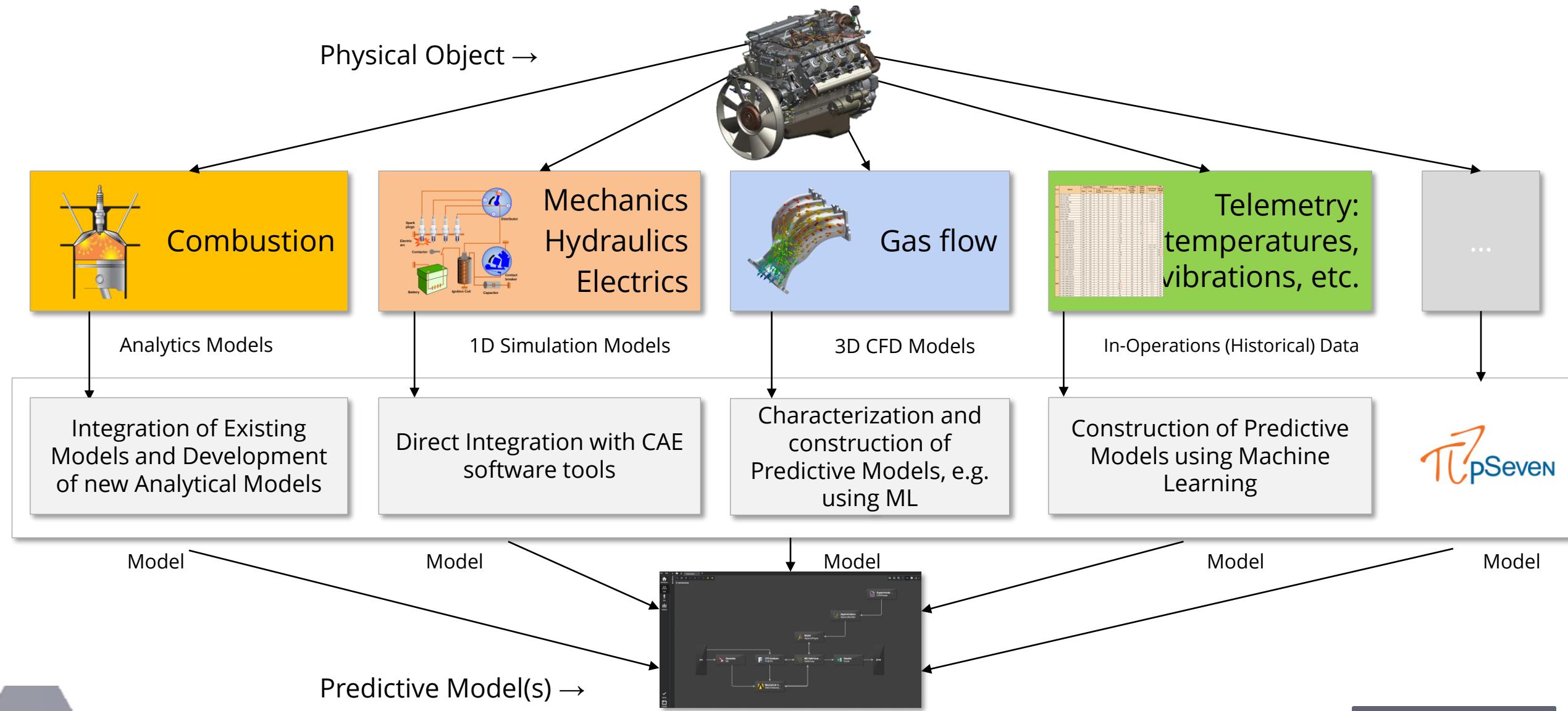


# Predictive Model

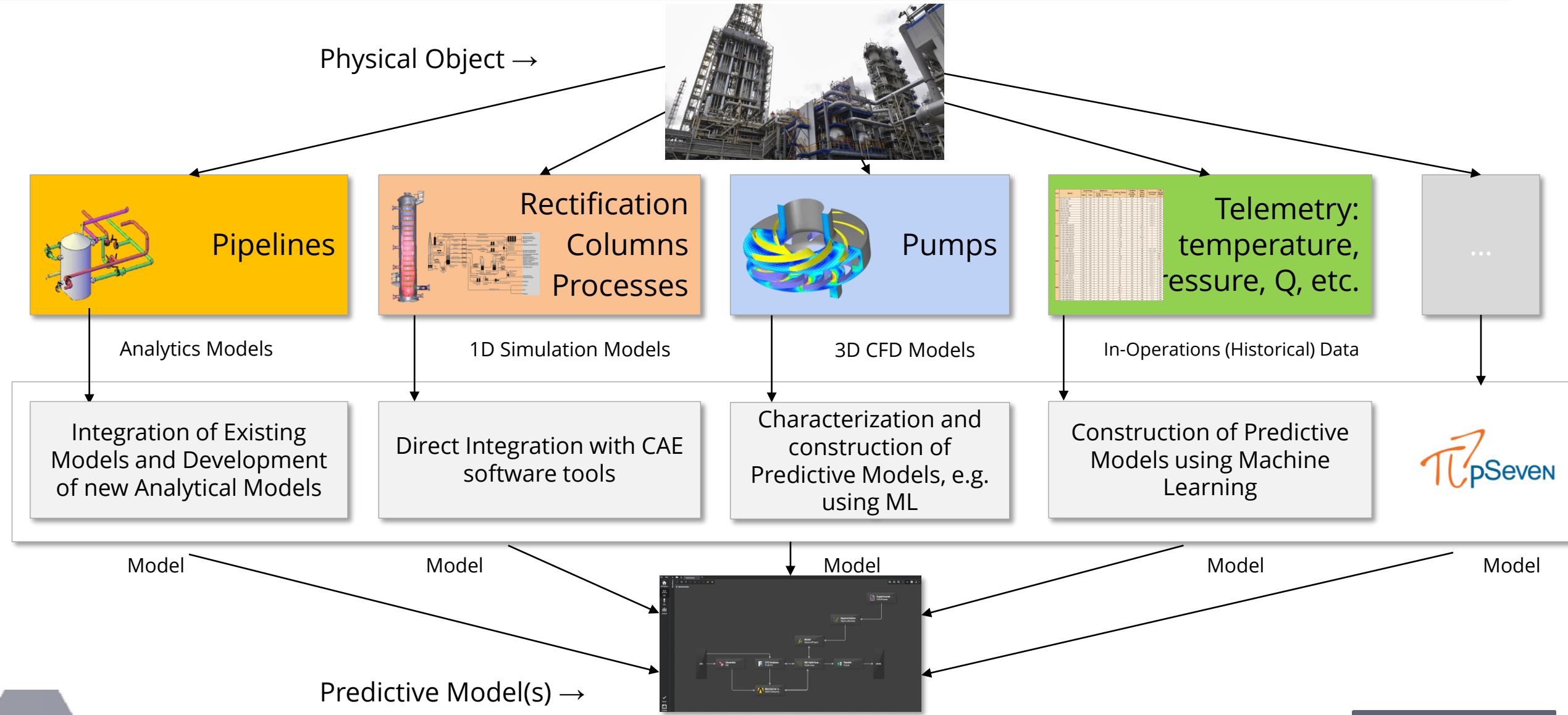
Machine Learning      CAE      Artificial Intelligence  
Big Data                  Predictive Analytics  
Industrial AI              Digital Twin  
Sensors data              IoT      MBSE  
Simulations

We need a powerful yet easy to use low-code web-based platform to **build**, **explore** and **operate** physics-based or data-driven predictive models, to **automate** and **orchestrate** processes, data and models, and to **deploy** models and applications.

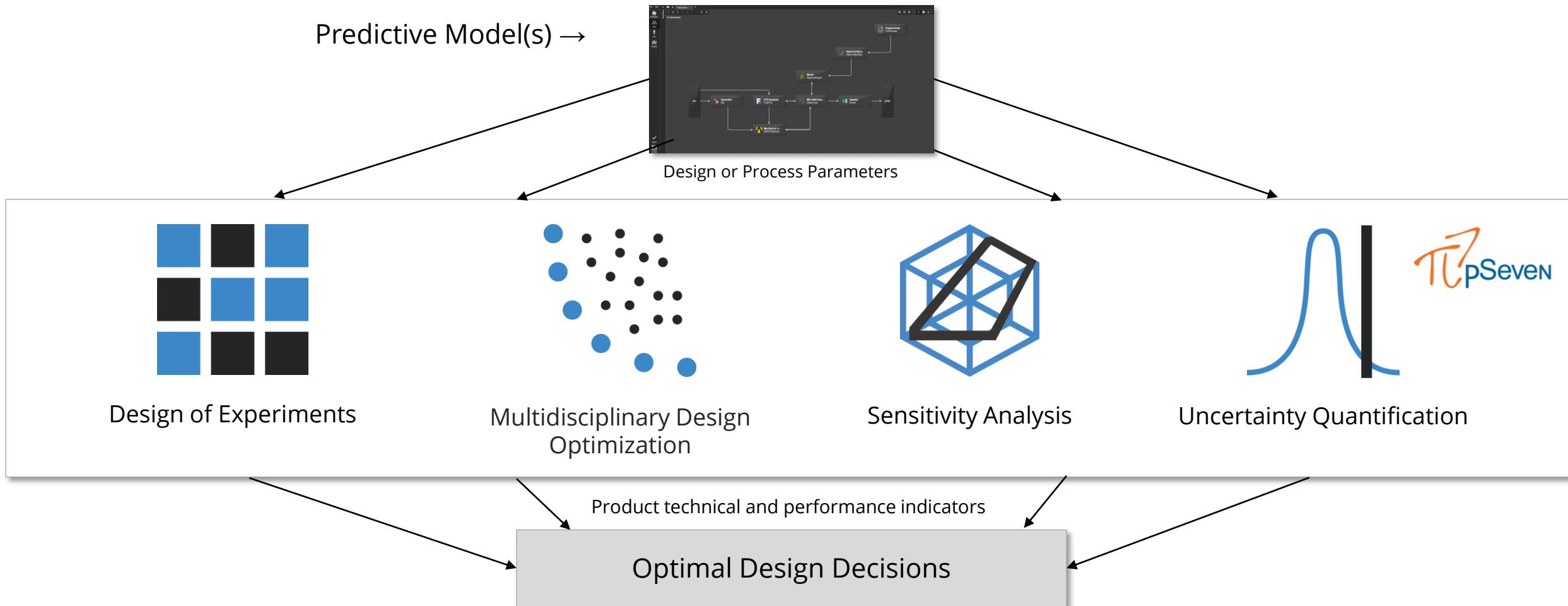
# Building Digital Twin: Illustration - Engine



# Building Digital Twin: Illustration – Refinery



# Design Space Exploration\* with Digital Twin: Illustration



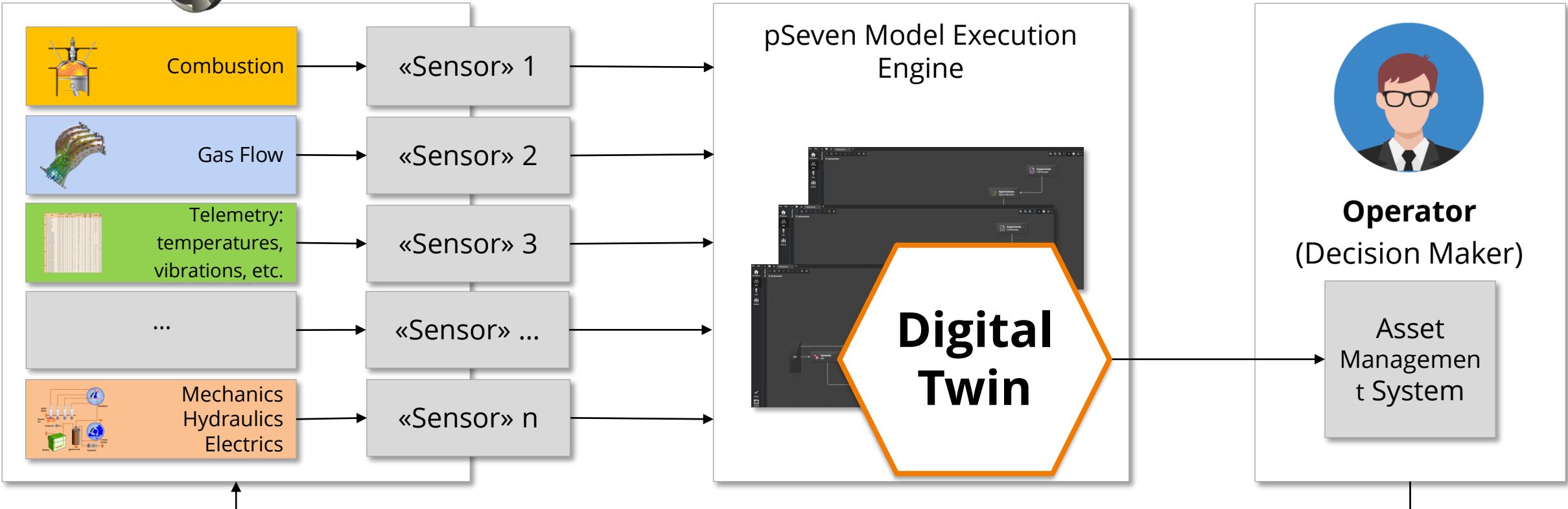
\* **Design Space Exploration** is a way to *systematically and automatically* explore very large numbers of design alternatives to find *optimal* performance parameters.

# Usage of Digital Twin in Operations: Illustration - Engine



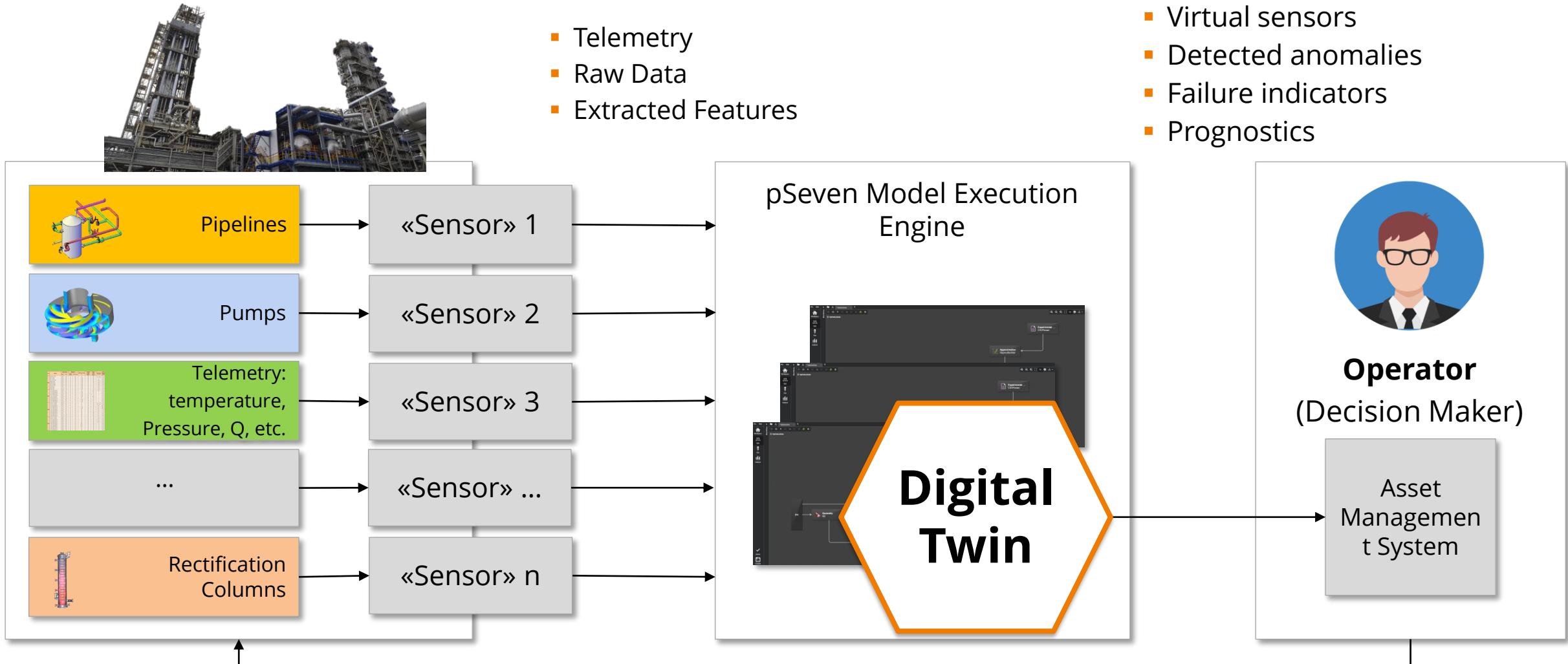
- Telemetry
- Raw Data
- Extracted Features

- Virtual sensors
- Detected anomalies
- Failure indicators
- Prognostics



- Action, e.g. predictive maintenance
- Change management
- Business processes optimization

# Usage of Digital Twin in Operations: Illustration - Refinery



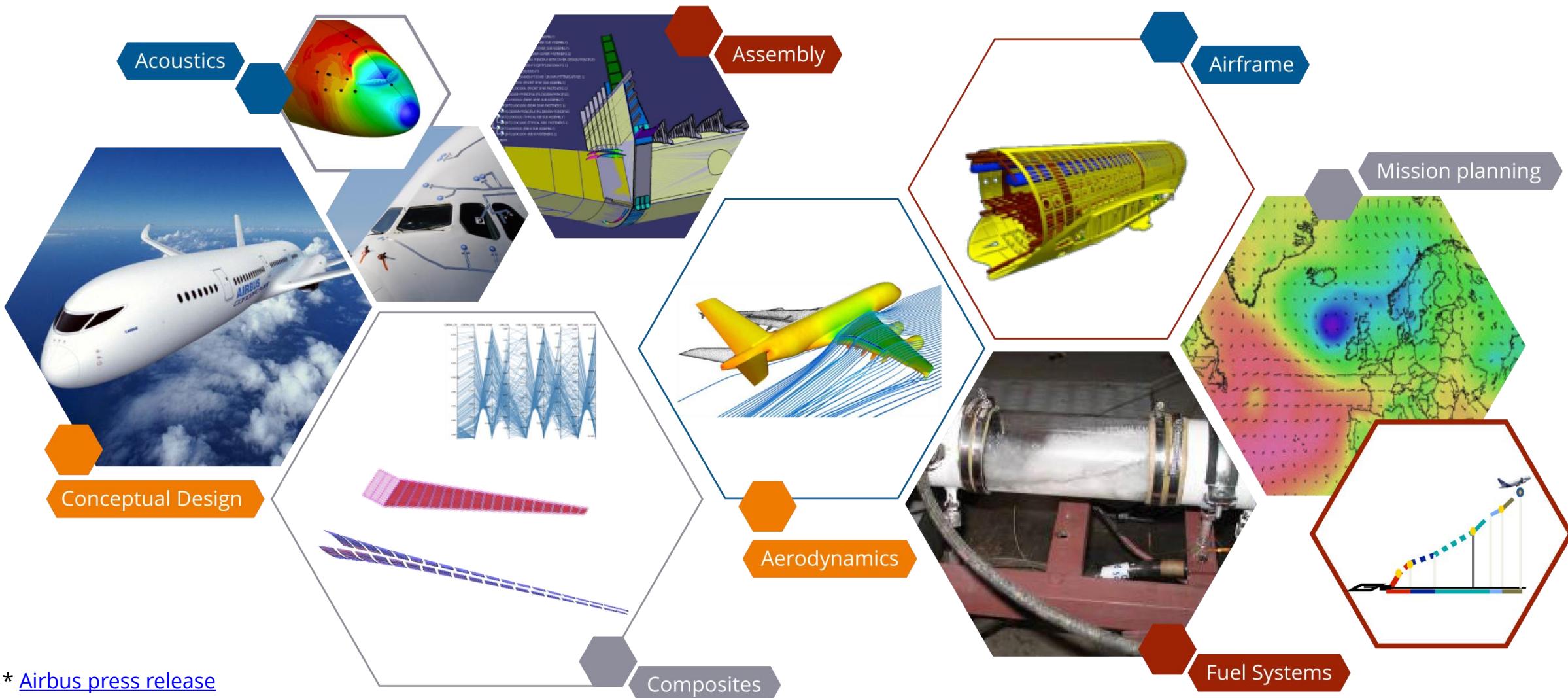
# Digital Twins: New opportunities for business transformation

- Significant **reduction of design time and cost** while improving the technical and operational characteristics of products (quality, reliability, safety, etc.)
  - New aircraft design lead time reduction by up to 10% with pSeven, - [Airbus](#) in 2013
- **Increase of product reliability and reduction of maintenance costs** through online monitoring, diagnostics and **predictive maintenance** (as a value added service for consumers).
  - Digital Twins allow to reduce the time to the first production by 1-3 years. At the same time, capital costs in the development of deposits are reduced by 20-30%, - Leading provider of products and services to the energy industry



FROM THE ASSET BASED ... TO A SERVICE DRIVEN





\* [Airbus press release](#)



## Objective

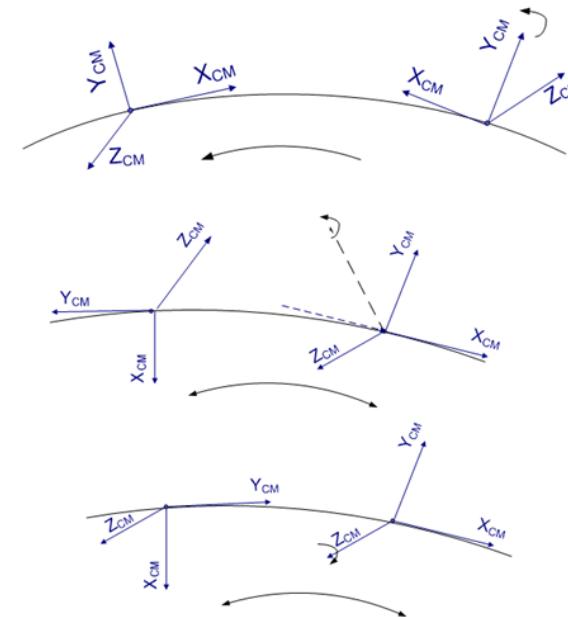
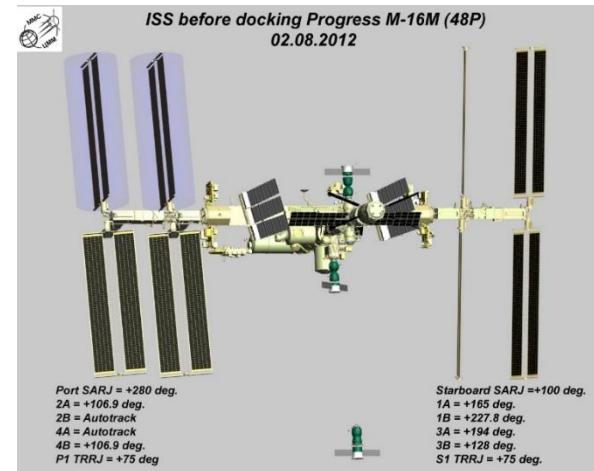
- Find optimal control parameters of ISS large angle maneuvers under the influence of external forces and with constraints on flexible modes

## Solution

- A set of program modules were developed in pSeven aimed to:
  - Model arbitrary ISS maneuvers
  - Exhaustively parameterize control signals in each channel
  - Determine fuel-optimal control cyclograms, type and number of thrusters

## Result – Proven by the a set real ISS maneuvers in Dec 2019

- Fuel efficiency of developed approach is about **x40** more effective than the existing method:
  - 180° turn on Ycm axis - from 80 to 4 kg of fuel
  - 180° turn on Euler axis - from 58 to 9.7 kg of fuel
  - 90° turn on Zcm (pitch) axis - from 17 to 1 kg of fuel
- Developed approach is **x5** more effective than upcoming NASA methodology





# : Multistage Aerodynamic Steam Turbine Gas Path Optimization



## Objective:

- Maximize efficiency of High Pressure Cylinder (HPC), Intermediate Pressure Cylinder - 1 (IPC-1) and Intermediate Pressure Cylinder - 2 (IPC-2) subject to geometry and mechanical stress constraints

## Problem:

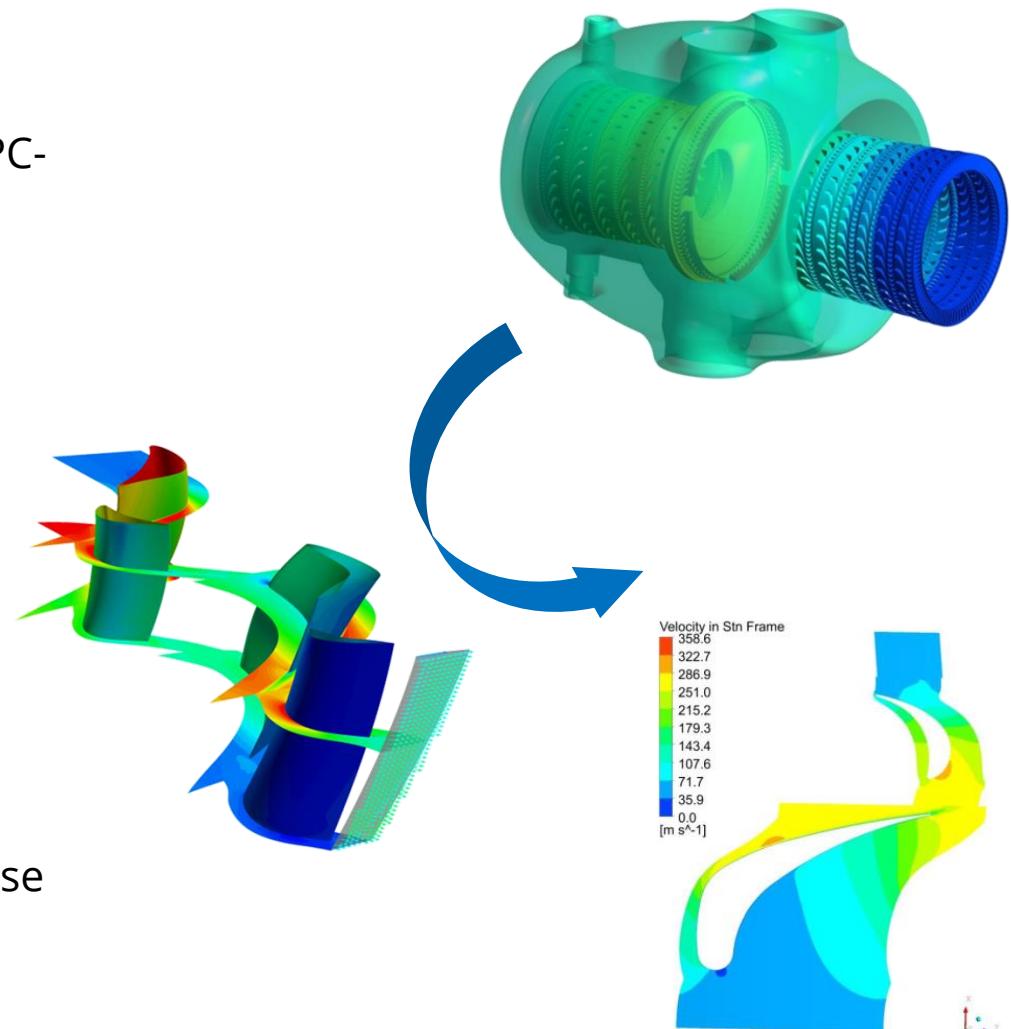
- HPC – 9 stages (69 parameters)
- IPC-1 – 10 stages (90 parameters)
- IPC-2 – 6 stages (72 parameters)
- 3D simulation in ANSYS CFX using HPC cluster

## Solution:

- Local gradient optimization method implemented in pSeven

## Results:

- **Efficiency boosted by 2%-4%** with capacity increase by 3%-6% because of geometry optimization (rotor and stator blades) and changing of operating mode



# Adaptive DoE for Simulation Based Testing of Automated Driving Systems

## Intro

Simulation Based Testing of Automated Driving Systems, provided by FDTech GMBH, has several advantages compared to the real world testing. The objective of this use case is to help FDTech ensure the benefits of virtual testing like large scalability and variability, automated generation of driving scenarios, traceability and reuse of the testing data

## Objective

- Enable scalable test beds by:
  - Controllable Test bed automation layer (TAL)
  - Traceable and reusable testing scenarios data
  - Automation of test preparation, run and analysis
- Ensure full coverage of important driving scenarios with lowest possible number of test points

## Challenges

- Number of scenarios increases exponentially with number of parameters
- Different scenarios have different importance

## Solution

- Workflow automation in pSeven
- Adaptive Design of Experiments technique to drastically reduce number of simulations
- Predictive modelling and Machine Learning to cover the whole test space

## Result

- Fully automated and scalable workflow for virtual test execution
- Adaptive exploration of the driving scenarios to reveal most important with minimum test runs

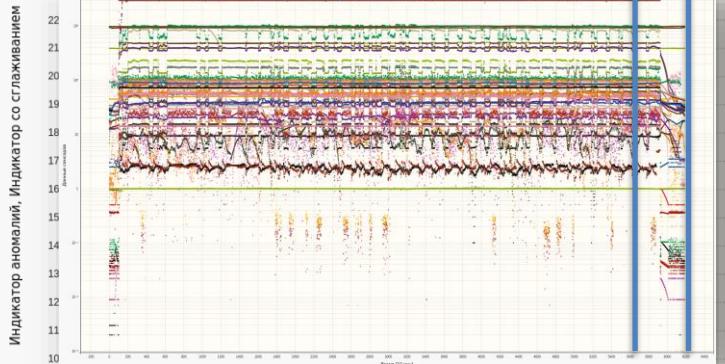




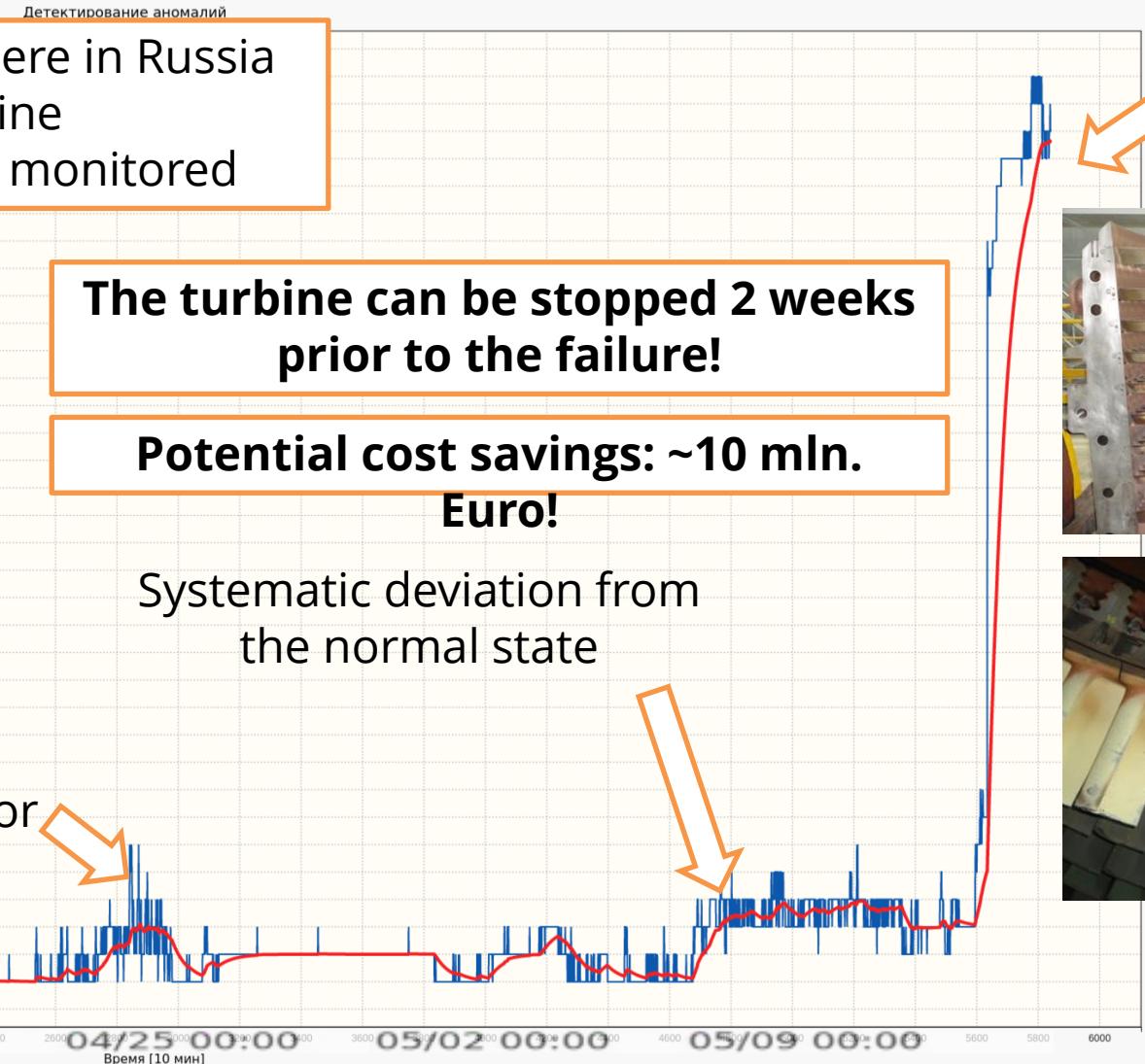
# POTEK : Post-mortem analysis of accident on power plant gas turbine



Location: Somewhere in Russia  
Object: Gas Turbine  
Sensors: 100+, 54 monitored



Failure early precursor



No domain specific knowledge was used. Just pure data analysis!



## ATA49

✓ APU core engine

## Goal

Predict failures of APU to improve maintenance procedure

## Result

Detection of failures with an accuracy of ~90% (9 correctly predicted failures account for 1 false alarm)

**Cost reduction** associated with downtime of plane due to unexpected failures by ~30%

## ATA79

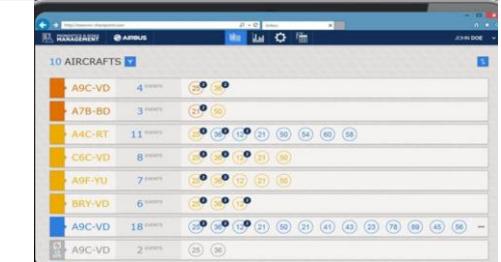
✓ Oil filter clogging

## Goal

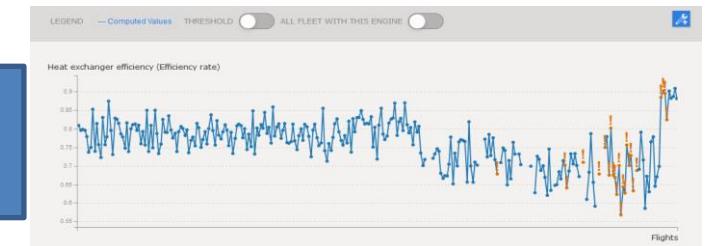
Optimize maintenance of an aircraft engine oil filter (minimize the number of checks and the costs of consumables)

## Result

**~50% reduction of the oil filter maintenance costs**



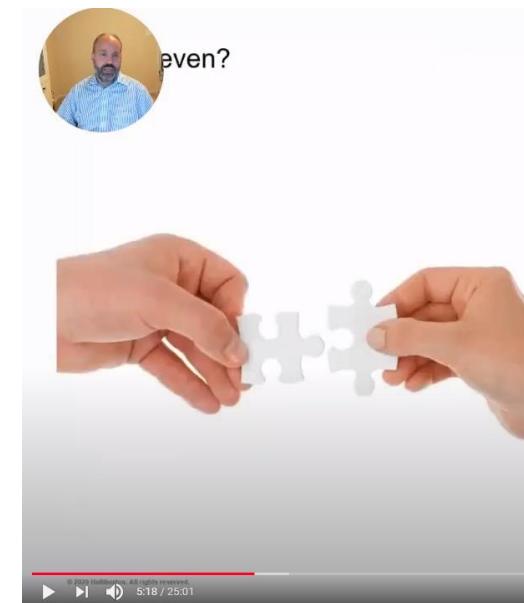
PRIORITY	ATA	MESSAGE
Low	21	Pack 1 heat exchanger efficiency low.



# Digital Oil Field with Halliburton Landmark\*

\* Information is publicly available at [Halliburton website](https://www.halliburton.com)

<https://www.datadvance.net/solutions/industry-solutions/resource-request/>



- Halliburton focus is on Oil & Gas
- pSeven Capabilities
  - High science
    - » Optimization
    - » Uncertainty analysis
    - » Predictive modeling
  - Visual programming
    - » Engineers are front and center
  - Cloud ready
    - » Web interface
- DATADVANCE's willingness to partner and grow together

<https://youtu.be/RS14qrwp-HQ?t=318>



Key driver: Process automation, Openness, Cloud-readiness, Algorithms

DATADVANCE



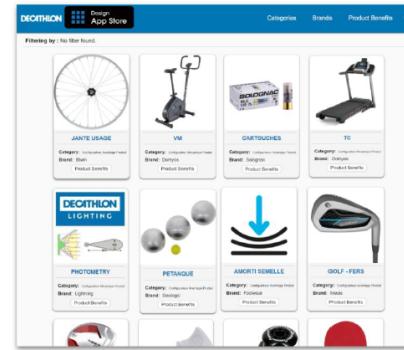
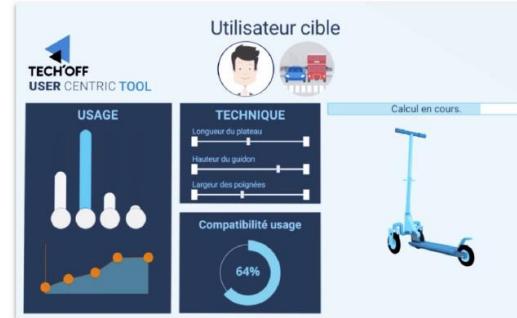
## Purpose

**CREATE SMART-ENGINEERING TOOLS  
DEDICATED TO PRODUCT DESIGN**

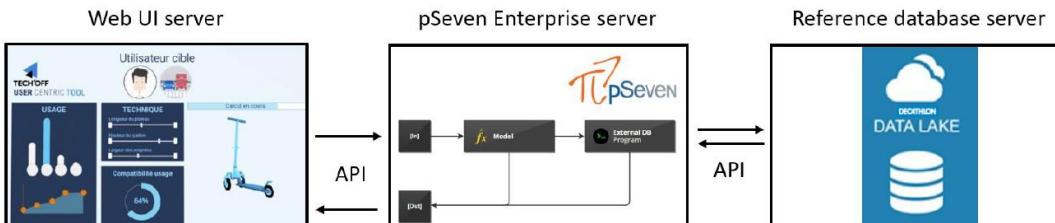
# Configurator  
**DESIGN APPLICATION STORE**

**PLATFORM TO HOST AND SHARE ENGINEERING  
KNOWLEDGE AND DESIGN INTERFACE**

- / Accessibility
- / Interoperability
- / User-Friendly



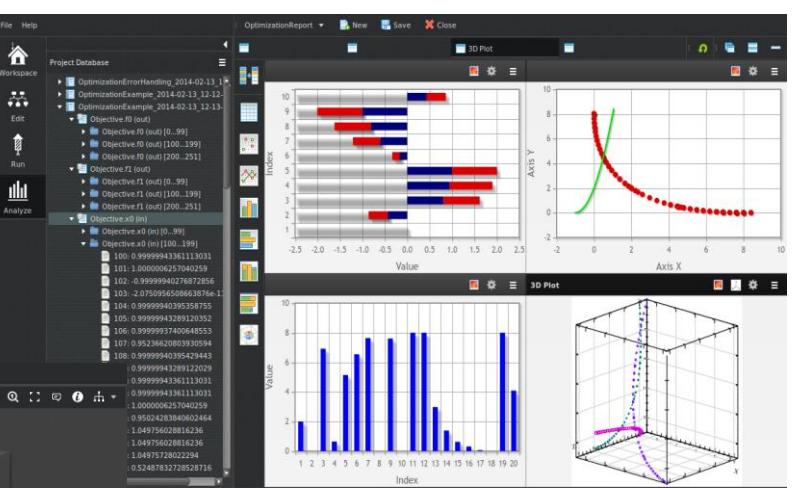
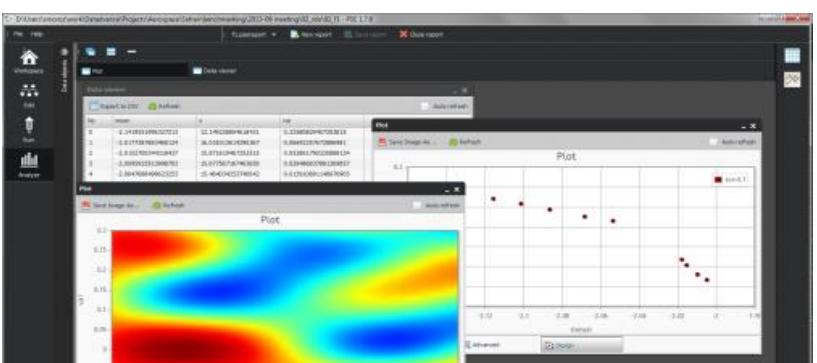
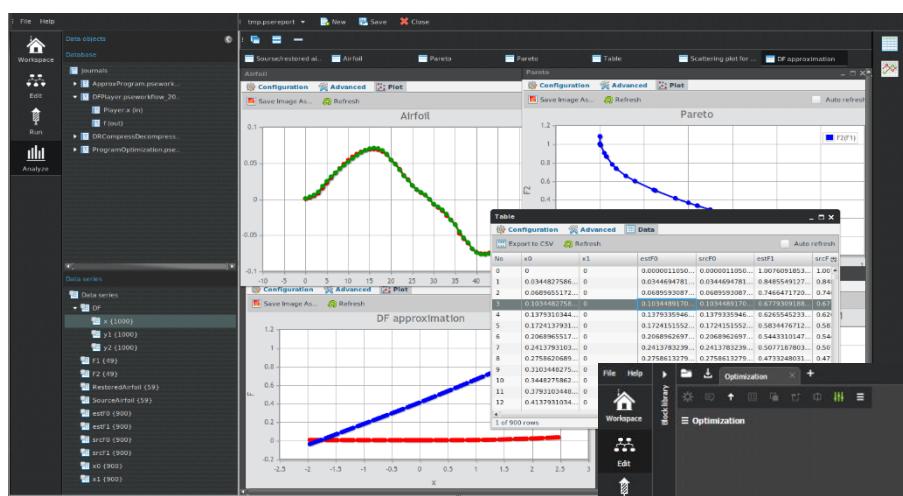
Server-based solution for predictions and optimization with approximation models.



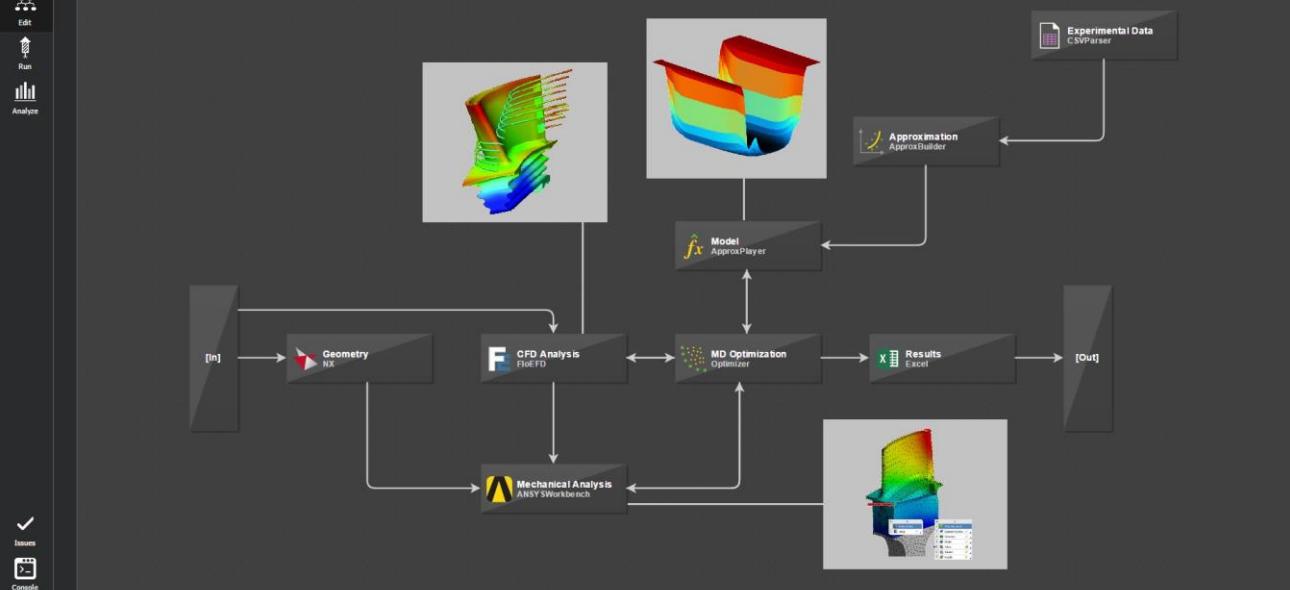
[https://youtu.be/kbaUR\\_4wZql](https://youtu.be/kbaUR_4wZql)



# Look and Feel



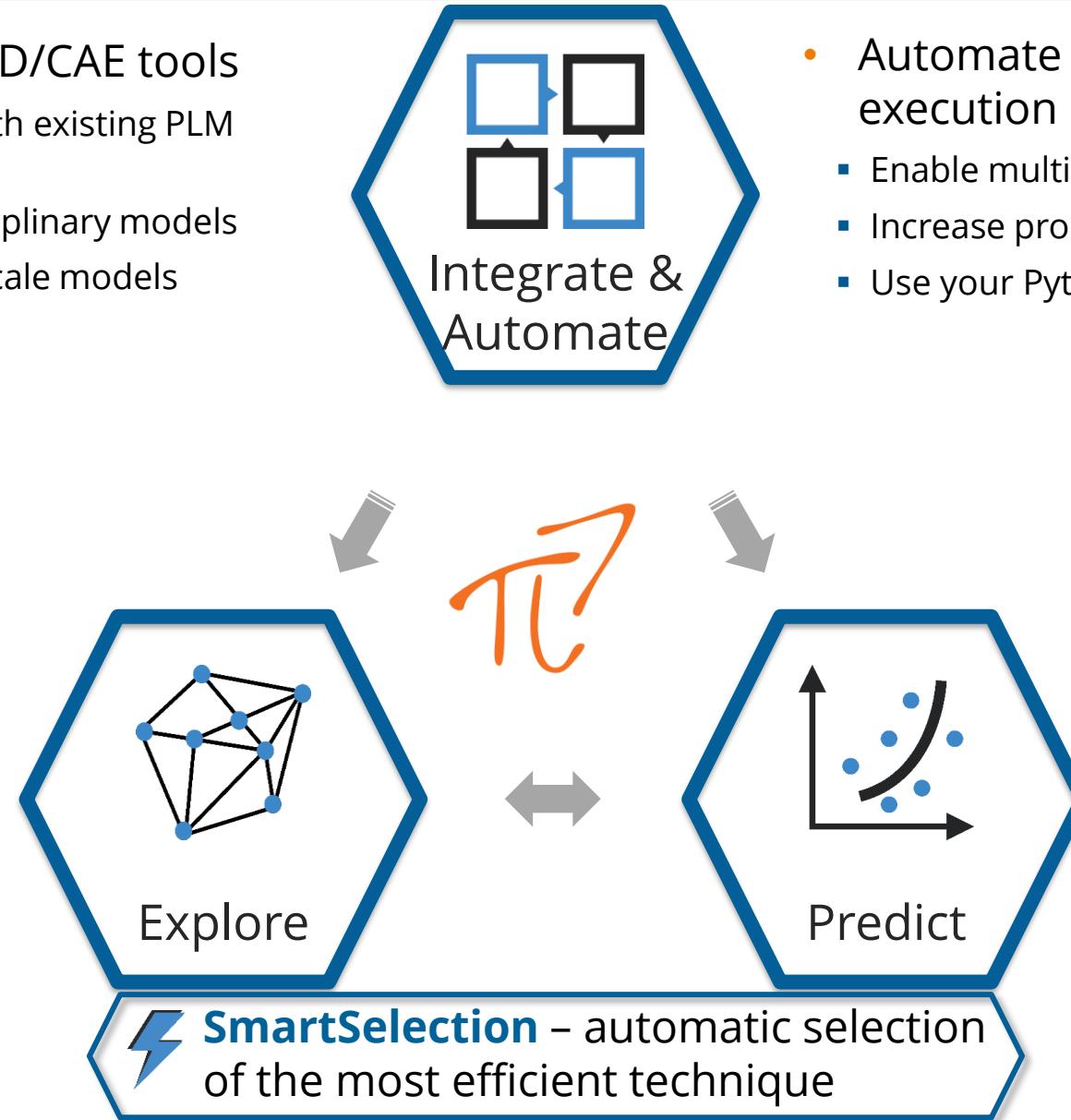
**pSeven**



**DATADVANCE**



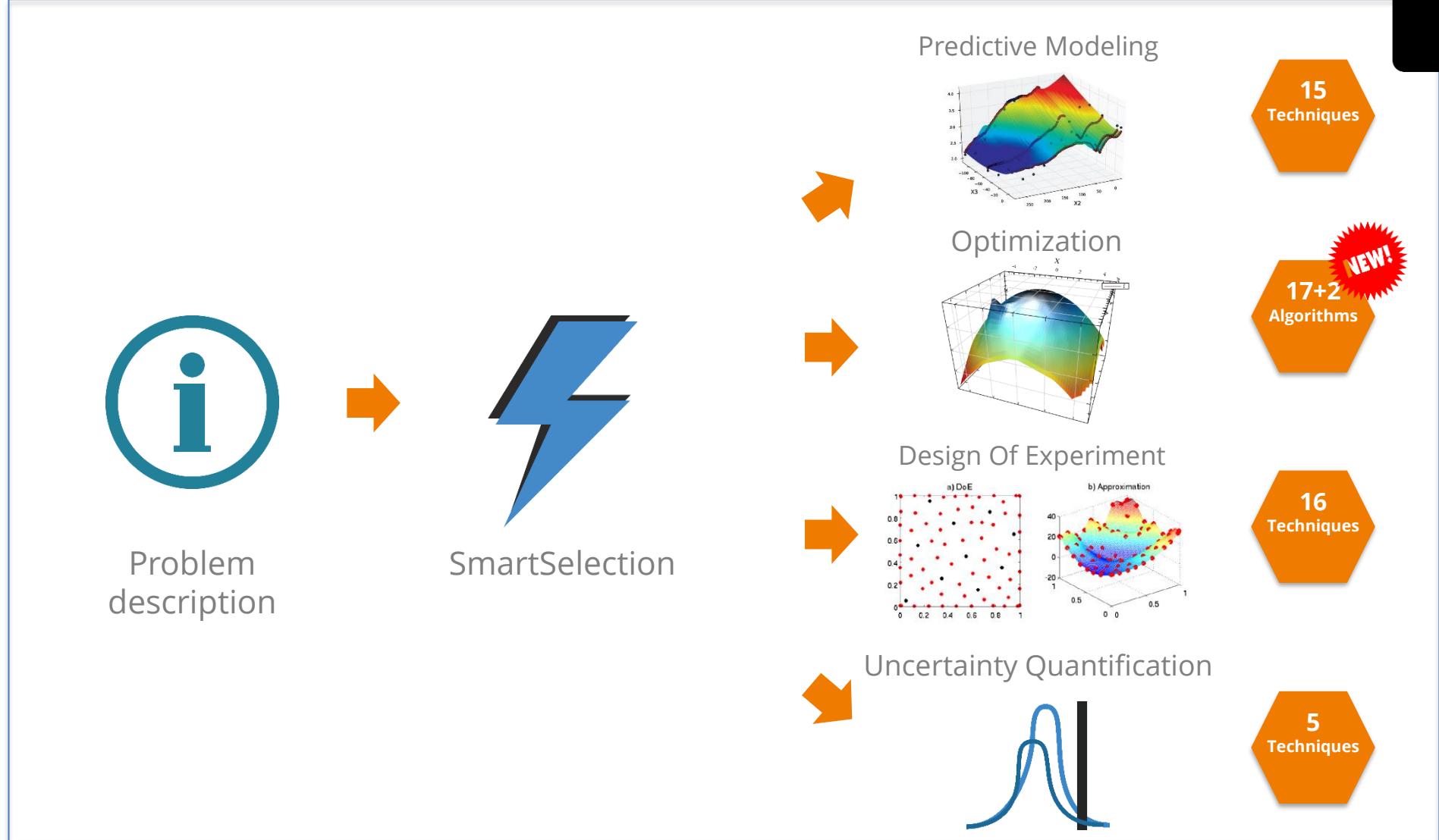
- Integrate CAD/CAE tools
  - Seamless fit with existing PLM environment
  - Build multidisciplinary models
  - Master multi-scale models
- Select the best design alternative
  - Perform trade-off studies
  - Identify dominant design parameters
  - Find optimal design based on best in class algorithms
- Make smart design and operation decisions
  - Unique Adaptive Design of Experiments analysis
- Calibrate numerical models with test data

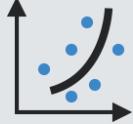


- Automate workflow execution
  - Enable multi-run strategies
  - Increase productivity
  - Use your Python-scripts

- Create predictive models
  - From Simulation, Test and Operation data
  - Data fusion from various high and low fidelity sources
  - Unique Mixture of Approximators models

# SmartSelection: Predictive Modelling & Design Space Exploration for Every Engineer





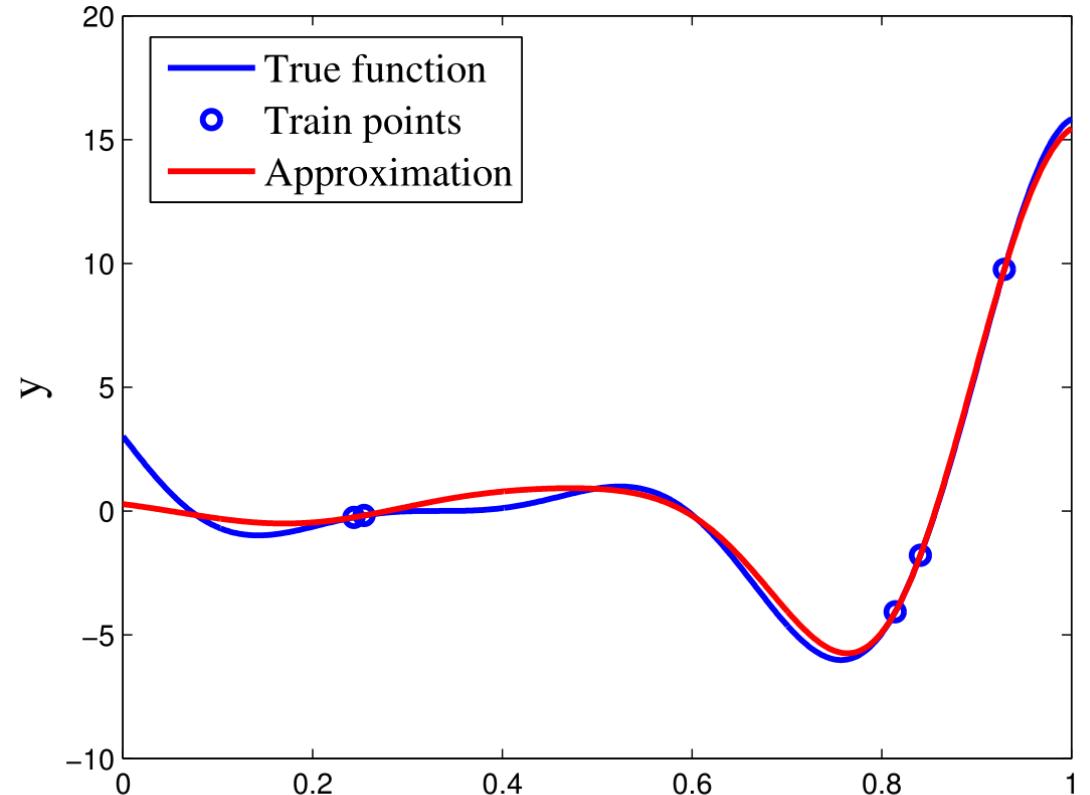
# Predictive Model - Supervised Learning Problem Formulation



- On defined set of objects  $X$  and set of known features  $Y$ , target function  $y: X \rightarrow Y$  assumed to exist
- By given training sample  $X^l = (x_i, y_i)_{i=1}^l$ , build decision function  $\hat{y}: X \rightarrow Y$  to approximate  $y$  all over  $X$

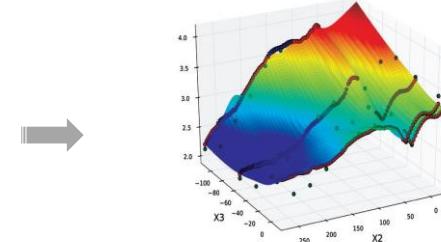
**Regression estimation** consider  $Y \in \mathbb{R}$

**Classification** consider  $Y = \{1, \dots, M\}$  or  $Y = \{0, 1\}^M$



**X**

Provide input parameters



Predictive Model

**y**

Predict output parameters

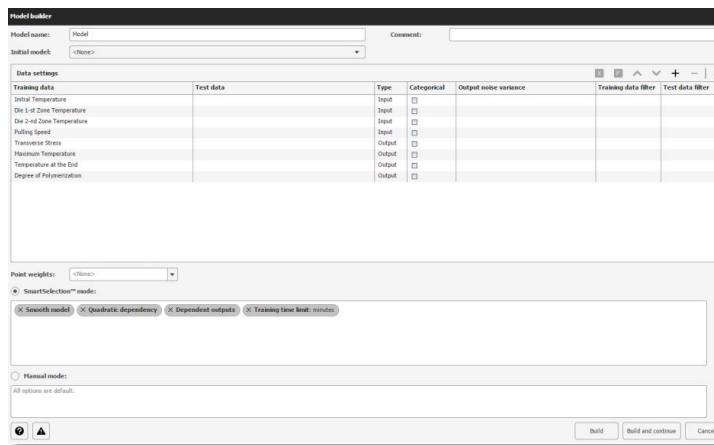


# Tools for Predictive Modeling



**pSeven includes a dedicated set of tools for building and managing predictive models that allows to:**

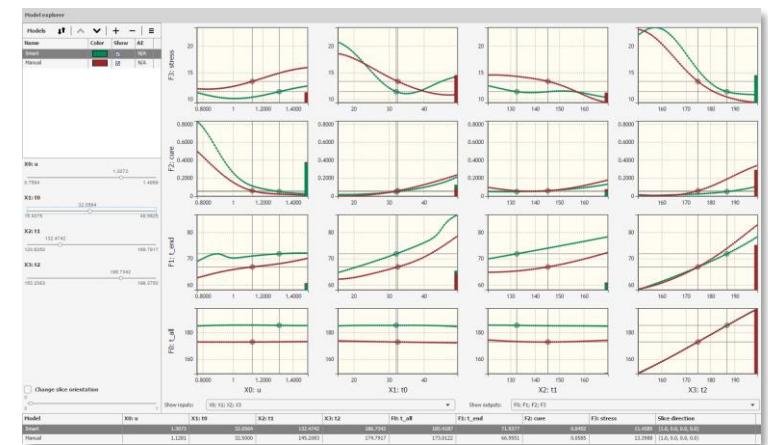
- Build fast and robust **predictive** models with **automatic** selection of technique
- **Validate**, test against reference data and compare models to find the **best** approximation quality
- **Explore** behavior of multidimensional models with studying input-output **dependencies**
- Export models to external **file**, including C source code, executable, Matlab/Octave, Excel and FMI



Model Builder



Model Validator



Model Explorer



# How to build Predictive Model in a few clicks

The screenshot shows the pSeven web-based application interface. On the left, there is a sidebar with icons for Workspace, Edit, Run, Analyze, Issues, and Console. The main area has tabs for WELCOME and Airbus. In the WELCOME tab, it says "This project does not have description yet. To add it, edit README.md file located in the project directory." There is a "Edit description" button. Below this, the "Project description" section has a heading "Edit" and a list of ways to edit: "Click on this link: [README.md](#). Double-click README.md file in the Project pane. Select **Edit description** item from the Project pane context menu." The "Syntax" section explains CommonMark syntax for headers and asterisks. A callout box in the bottom right corner says "As a **web-based application** pSeven is available in Chrome".

This project does not have description yet. To add it, edit README.md file located in the project directory.

For more details, open the project and see the **Project description** tab in **Workspace**.

## Project description

### Edit

There are several ways to edit project description:

- Click on this link: [README.md](#).
- Double-click README.md file in the **Project** pane.
- Select **Edit description** item from the **Project** pane context menu.

### Syntax

Project description uses CommonMark syntax.

You can make headers using hash characters:

```
# H1  
## H2  
### H3  
#### H4  
##### H5  
##### H6
```

Asterisk (\*)

As a **web-based application**  
**pSeven** is available in Chrome

\*italics\* or italics  
\*\*bold\*\* or bold



# Building Predictive Models

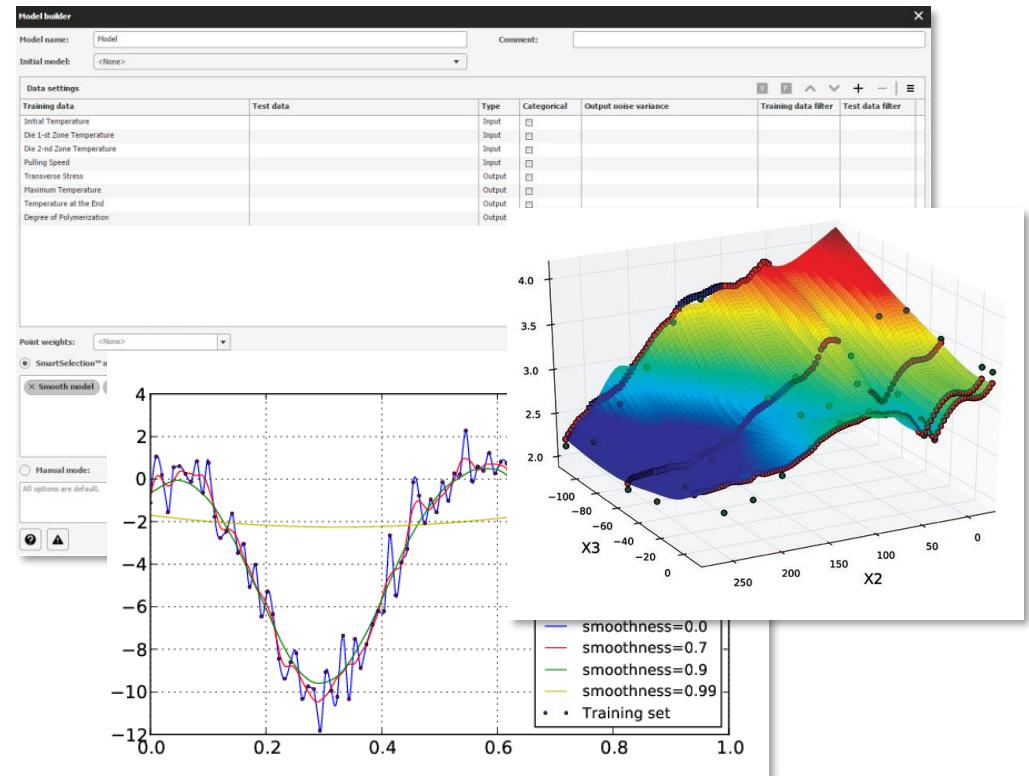


## pSeven provides capabilities and flexibility for building predictive models:

- Variety of industry-proven approximation techniques
- Full control of the model building time
- Accuracy and error estimation
- Dealing with oscillations and model smoothing
- Logarithm of outputs and exact fit
- Handling anisotropic data, discontinuities and inhomogeneous data, multi-fidelity data
- Updating existing models with new data & combining of the models



SmartSelection selects the most efficient technique for a given problem and data automatically!





# Predictive Modeling Techniques

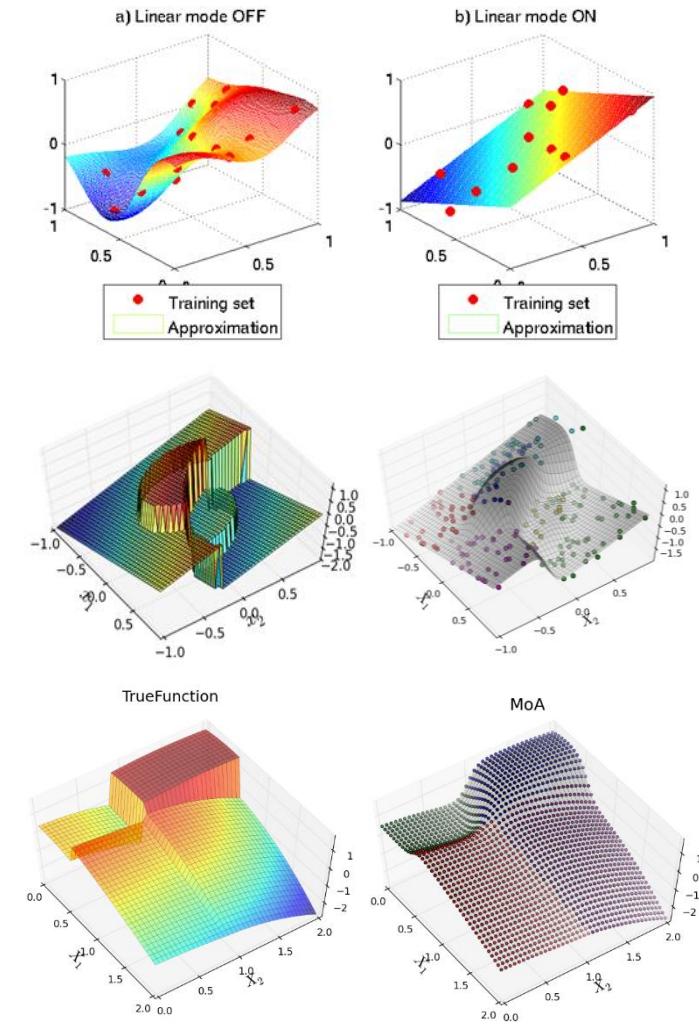


**pSeven provides industry proven techniques for any type of approximation and data:**

- Piecewise Linear Approximation (PLA)
- 1D Splines with Tension (SPLT)
- Response Surface Model (RSM)
- Tensor Approximation (TA)
- Incomplete Tensor Approximation (iTA)
- Gaussian Processes (GP)
- Sparse Gaussian Process (SGP)
- Tensored Gaussian Processes (TGP)
- Gradient Boosted Regression Trees (GBRT)
- High-Dimensional Approximation (HDA)
  - 2 Layer Artificial Neural Network
- High-Dimensional Approximation Combined with Gaussian Processes (HDAGP)
- Mixture of Approximators (MoA)
- Table Function (TBL)



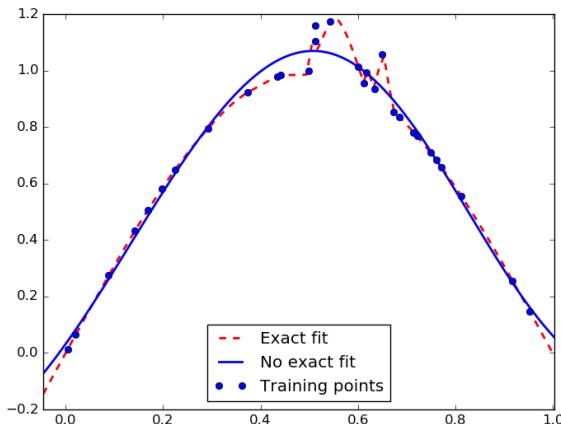
**SmartSelection** selects the most efficient **technique** for a given problem and data **automatically!**

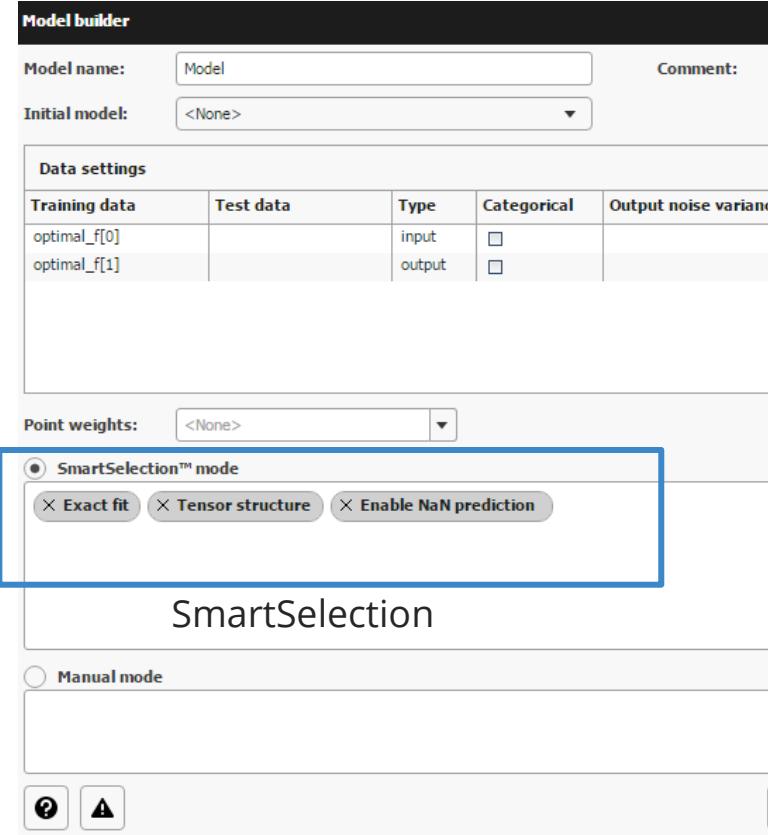




## SmartSelection grants automatic and adaptive choosing of approximation technique(s):

- Set of **hints** helps the user to describe problem and desired solution:
  - Hints about the **data**: dependencies, structure, etc.
  - Desired **model** properties: exact fit, smooth, etc.
  - **Building** properties: time, quality, etc.
- For better approximation quality, different parts of the model can be built using **different techniques**
- Meta-learning inside
- Hyper-Parameter Optimization inside





**SmartSelection**

Hints

<b>linear dependency</b>	
<b>quadratic dependency</b>	
<b>discontinuous dependency</b>	
<b>dependent outputs</b>	
<b>has tensor structure</b>	
<b>smooth model</b>	
<b>exact fit</b>	
<b>accuracy evaluation</b>	
<b>do not store training sample</b>	
<b>randomized training</b>	
<b>seed</b>	
<b>enable NaN prediction</b>	
<b>acceptable quality</b>	
<b>training time limit</b>	
<b>validation type</b>	
<b>internal validation</b>	

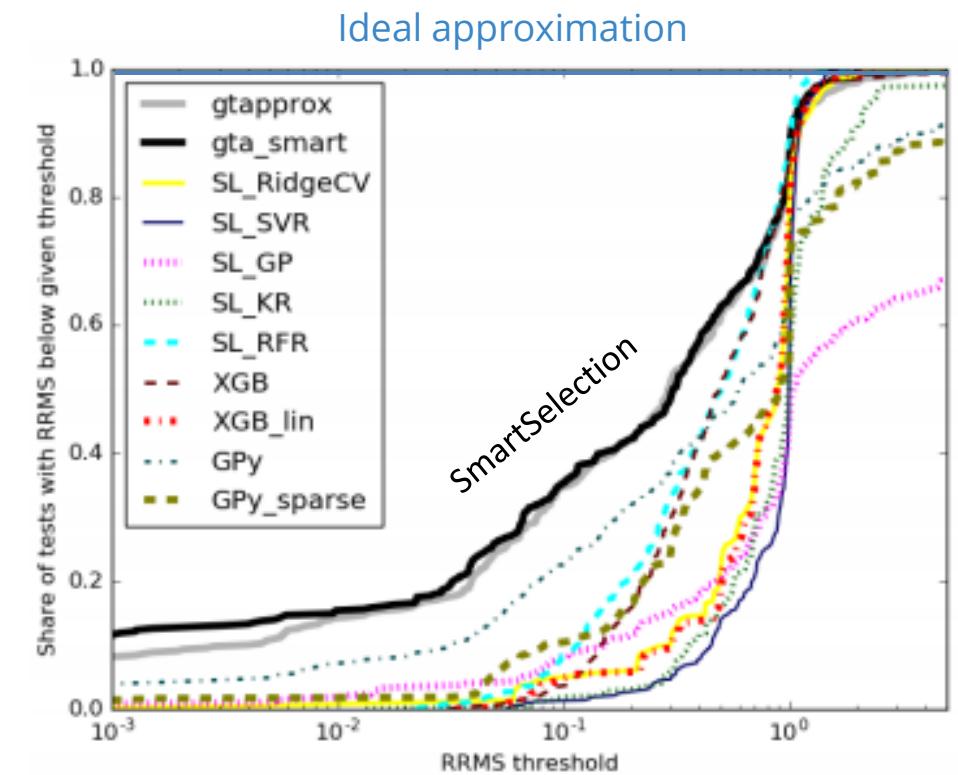


# Benchmark: SmartSelection for Predictive Modeling



## Quality of predictive model built with SmartSelection:

- Even with default settings SmartSelection builds predictive models of better approximation quality than famous open algorithms, like Scikit-learn, XGBoost and GPy.



pSeven vs. open algorithms

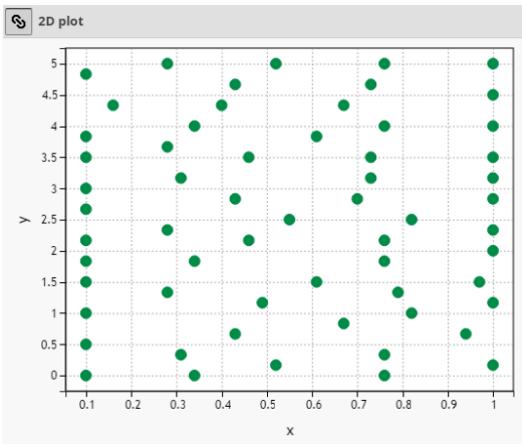
Full benchmark can be found here: <https://arxiv.org/abs/1609.01088>



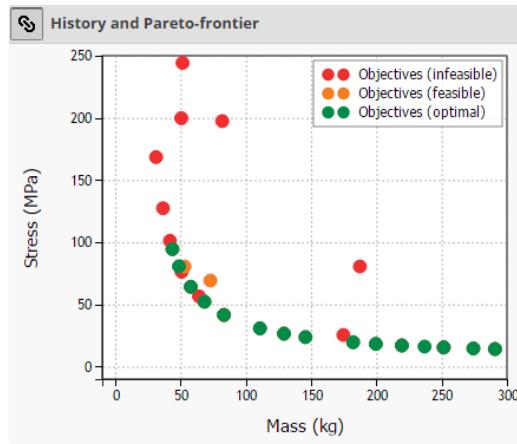
# Design Exploration Tools in pSeven



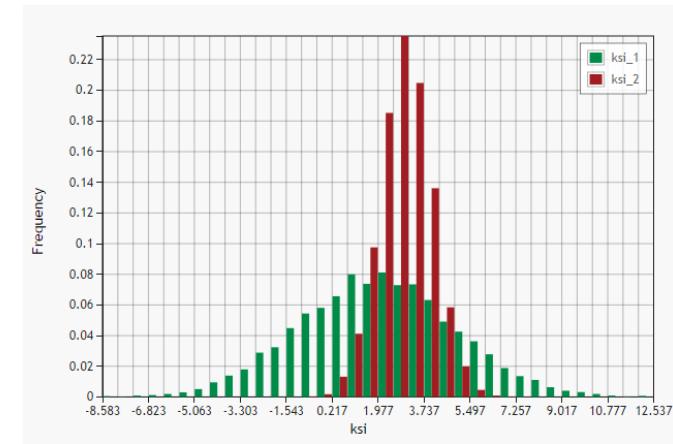
- Design of Experiments



- Design Optimization

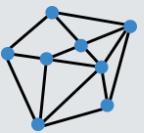


- Uncertainty Quantification



Design space  
exploration

UQ



# Design Optimization in pSeven

**pSeven provides easy and effective solution for most of industry optimization problems:**

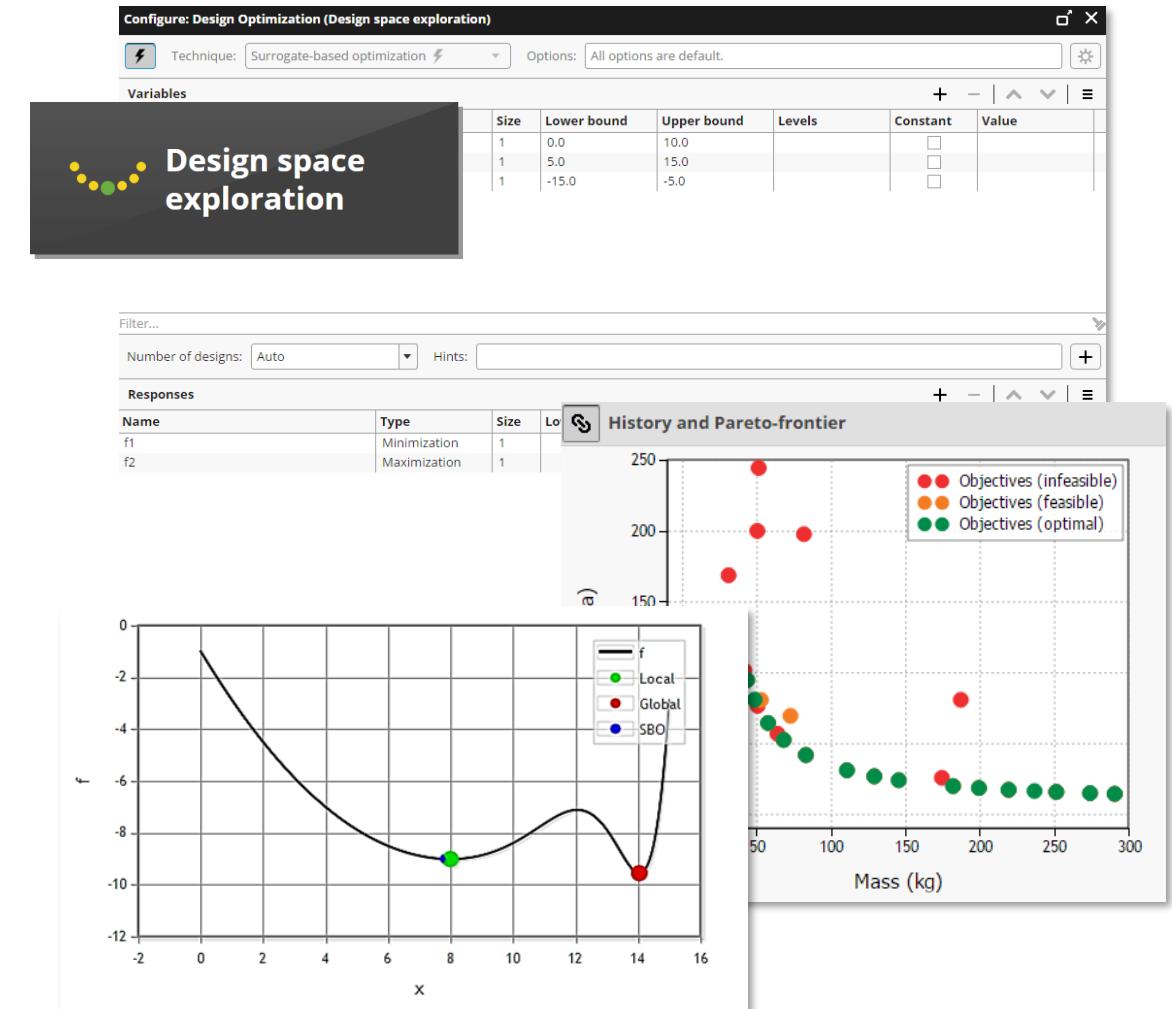
- Single- or multi-objective, robust optimization
- Large **dimensionality\***
- Long model evaluation time\*\*
- Continuous and **discrete** input variables
- Nonlinear, multimodal or **noisy** objective functions and constraints
- Presence of **implicit constraints** and domains of **undefined behavior**
- Presence of uncertainties

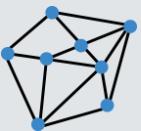


**SmartSelection chooses the optimization algorithms automatically and adaptively!**

\* - Up to 100 design variables for nonlinear time-consuming models

\*\* - For example, any CAE model





# Optimization Algorithms in pSeven



**pSeven includes full set of algorithms for solving single- and multi-objective design optimization problems:**

- Method classes\*:

- Multi-Objective
- Derivative-Free Optimization
- Gradient-Based Optimization
- Surrogate-Based Optimization (SBO)

- Globalization algorithms:

- Random Linkage (RL)
- Plain MultiStart (PM)
- MultiStart Pi (MS)

- Local algorithms:

- Quadratic Programming (QP)
- Sequential QP (SQP)
- Sequential Quadratic SQP (SQ2P)
- Feasible Direction (FD)
- Mixed-Integer Linear Programming (MILP)
- Method of Multipliers (MoM)
- Non-Linear Conjugated Gradients (NCG)
- Non-Linear Simplex (NLS)
- Powell's Conjugate Direction Method (Powell)

SmartSelection chooses the optimization algorithms automatically and adaptively!



\* - Multiple method classes are also supported, for example:

Multi-Objective Gradient-Based Optimization



# SmartSelection for Design Exploration



## Design Exploration technique is chosen automatically with SmartSelection:

- Set of **options** and **hints** helps the user to describe a problem and desired solution:
  - Types of **variables**: continuous, discrete, categorical
  - Types of **responses**: evaluation, minimization, constraint etc.
  - Function of responses: generic, linear, quadratic etc.
  - Problem **hints**: noisy, expensive

## Design Optimization algorithms' parameters are adaptively tuned during the solution:

- In the case of performance degradation solution process interrupts and restarts with the next suitable algorithm

Configure: Design space exploration (Design space exploration)

Technique: Latin hypercube sampling Options: All options are default.

Name	Type	Size	Lower bound	Upper bound	Levels	Constant	Value
X1	Continuous	1	0.0	10.0		<input type="checkbox"/>	
X2	Continuous	1	0.0	5.0		<input type="checkbox"/>	
X3	Continuous	1	5.0	15.0		<input type="checkbox"/>	

Filter...

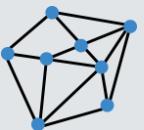
Number of designs: Auto Hints: +

Name	Type	Size	Lower bound	Upper bound	Value	Function	Blackbox
F1	Evaluation	1				Generic	<input checked="" type="checkbox"/>
F2	Evaluation	1				Generic	<input checked="" type="checkbox"/>

Filter...

Run options Ports and parameters OK Cancel Apply

The screenshot shows the 'Configure: Design space exploration' dialog. In the 'Variables' section, three continuous variables X1, X2, and X3 are defined with sizes 1, lower bounds of 0.0, and upper bounds of 10.0, 5.0, and 15.0 respectively. In the 'Responses' section, two evaluations F1 and F2 are defined with sizes 1, and function types Generic. The 'Technique' dropdown is set to 'Latin hypercube sampling'. The 'Run options' button is highlighted with a red box.



## Benchmark: Multi-Objective Gradient-Based Optimization



- Test problem: Well-known TP7
- Objective functions evaluations:
  - pSeven - 187
  - NSGA2 - 200

$$\min_{\vec{x}} \left( x_0, \frac{1+x_1}{x_0} \right),$$

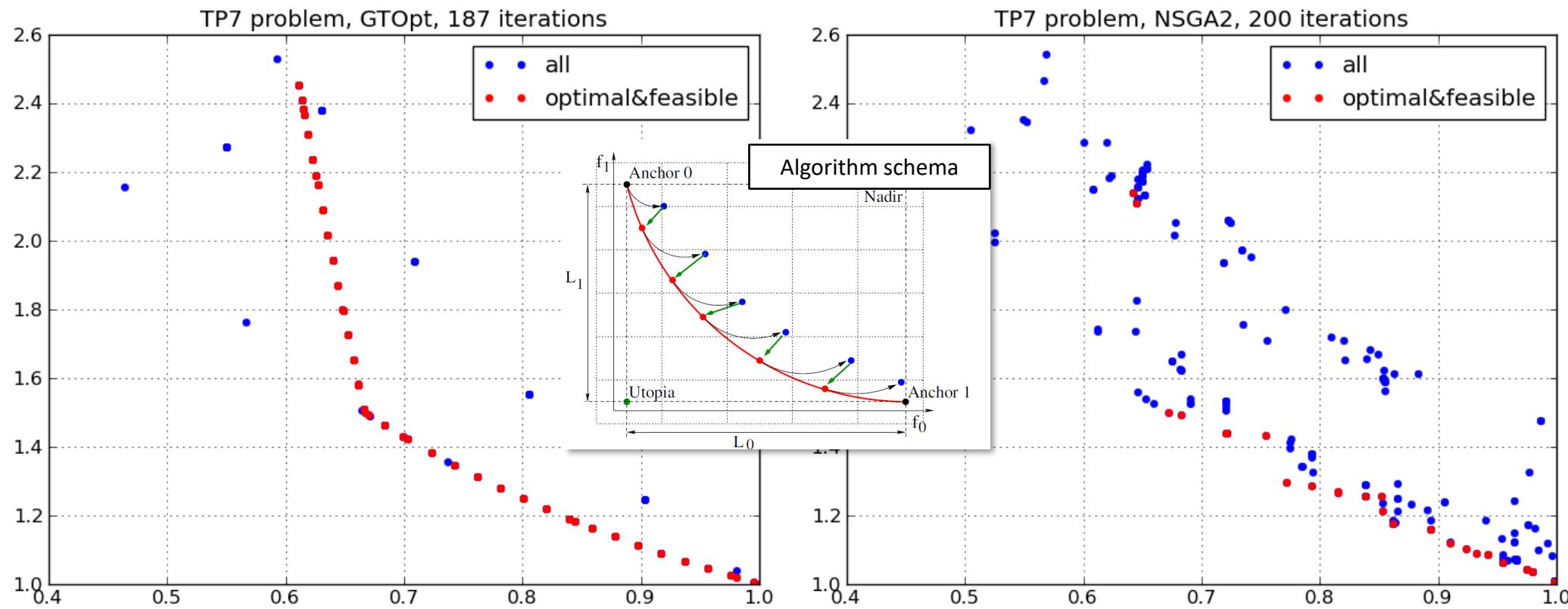
where

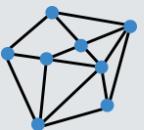
$$0.1 \leq x_0 \leq 1$$
$$0 \leq x_1 \leq 0.5$$

subject to

$$x_1 + 9x_0 \geq 6$$

$$-x_1 + 9x_0 \geq 1$$

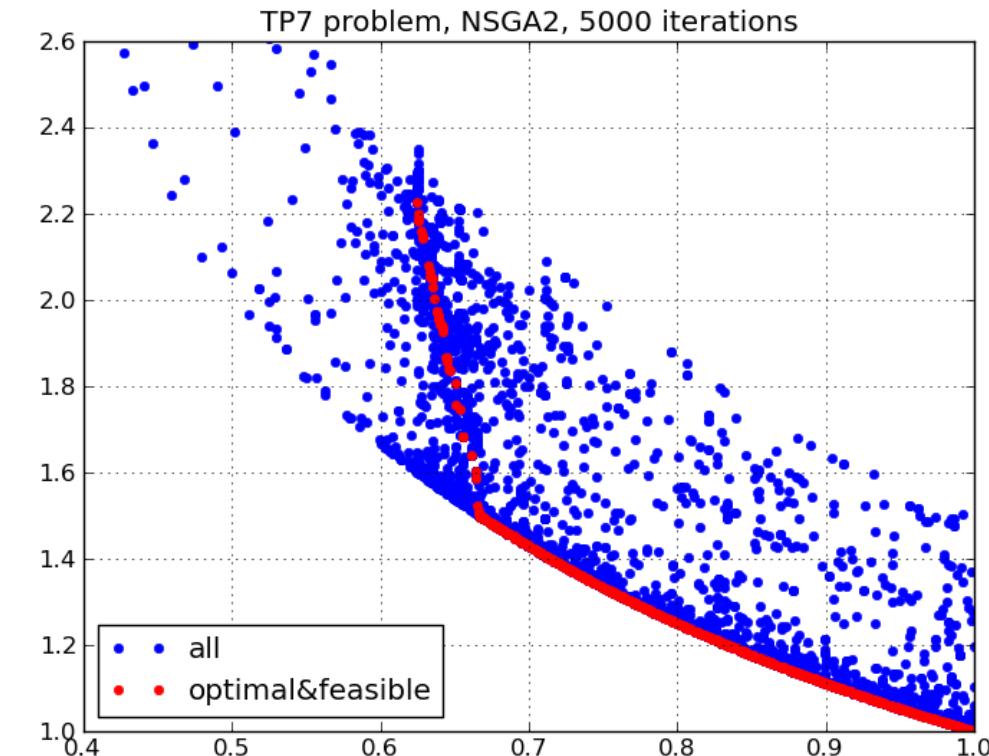
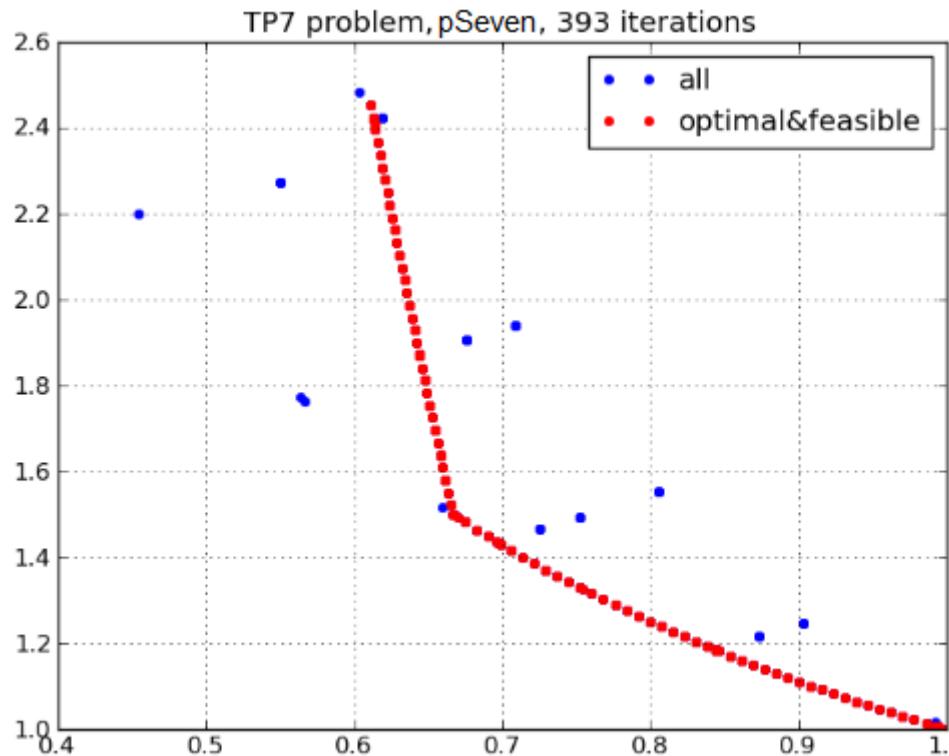


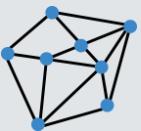


## Benchmark: Multi-Objective Gradient-Based Optimization



- Test problem: Well-known TP7
- Objective functions evaluations:
  - pSeven - 393
  - NSGA2 - 5000
- NSGA2 needs **13x times** more evaluations to have the same Pareto Frontier density





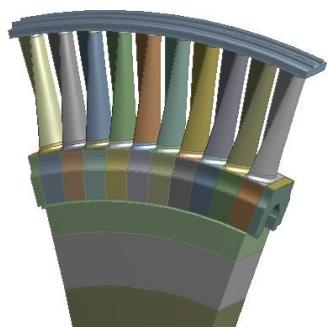
# Surrogate-Based Optimization in pSeven



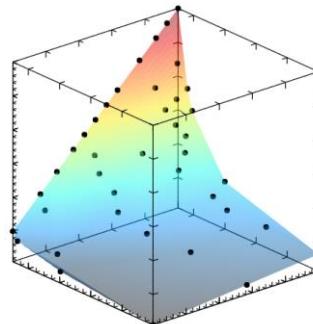
Surrogate-Based Optimization (**SBO**) is a class of optimization methods that involves **building** and optimizing **surrogate** models as an intermediate step.

## SBO in pSeven:

- Implements in-house developed **DoE** strategy, which respects as much **feasibility** domain of the problem as possible
- Intelligently spends model evaluations **budget** and **speeds** up the solution multiple times
- Allows solving **large scale** problems (up to 1000 design variables) effectively using in-house implementation of **multi-resolution GP**



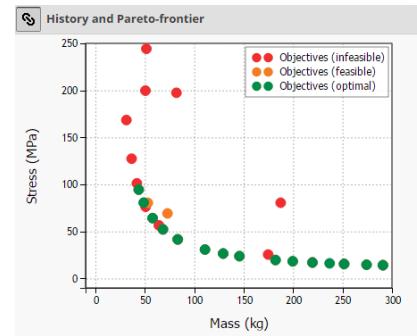
Create model



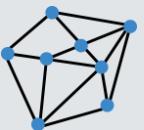
SBO runs a DoE to build  
an initial surrogate

SBO predicts next  
design point

Surrogate is  
rebuilt



Optimization step

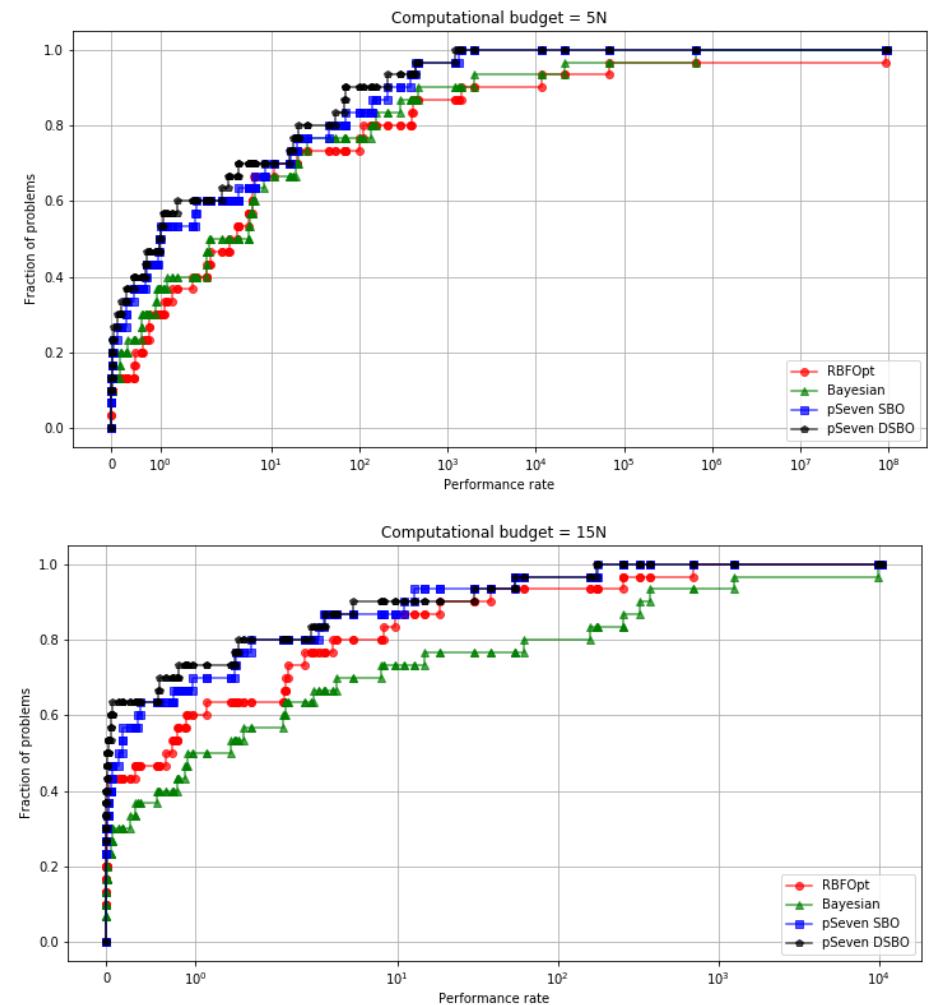


# Benchmark: Surrogate-Based Optimization Algorithms for Expensive Problems

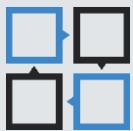


- Test problem: combination of 30 different benchmark functions
- pSeven, RBFOpt and Bayesian surrogate-based optimization algorithm were used
- pSeven is well-suited for expensive optimization problems in cases of the restricted computational resources
  - Especially the Direct SBO (DSBO) algorithm that fundamentally relies only on the exploitation meaning that it takes into account only the values of a surrogate model

Learn more →



Performance profiles of algorithms at the given number of evaluations



# Simulation Process and Data Processing Automation in pSeven



All Simulation, Data Analytics and Optimization tasks can be routinely automated!

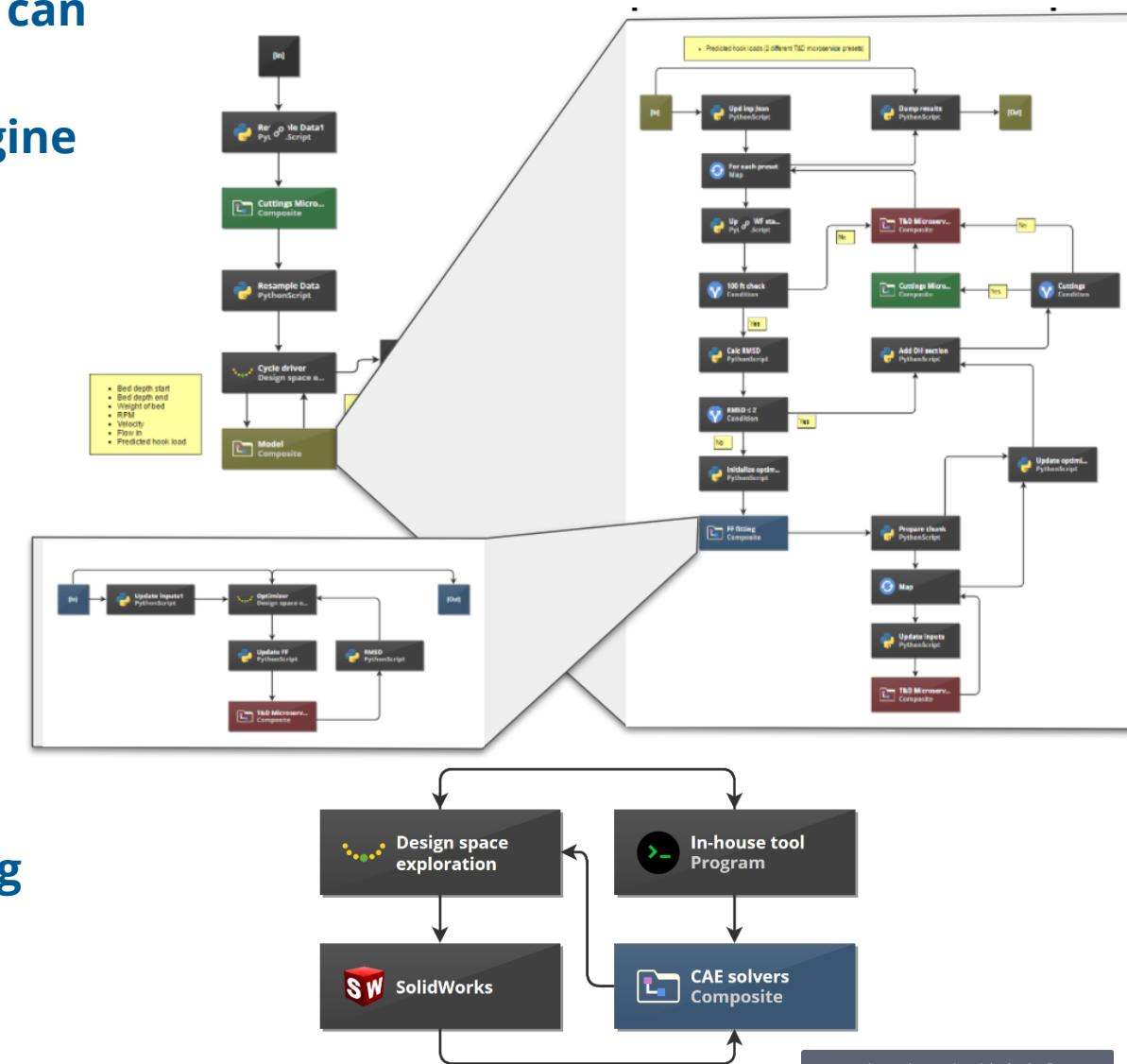
Powerful workflow automation and orchestration engine

Full-fledge visual programming:

- Intuitive and visual definition of complex processes
- Implementation of logic operations and nested loops
- Parallel and remote execution
- Data reuse and caching

Rich library of building blocks

Missing blocks and tools can be easily embedded using Python, a first class citizen in pSeven





pSeven  
Enterprise

# Same Look and Feel, But ...

**pSeven Enterprise**

Disk optimization | SciPy Integration Report | RotatingDisk2020 | 37 | 38 | 39 | 40 | 41 | 42 | 43 | 44 | 45 | 46 | 47 | 48 | 49 | 50 | 51

Sample Data Statistics

index	x	y	objective	constraint	status
37	7.512645e-1	5.501375e-1	8.220649e-2	8.670497e-1	"feasible"
38	7.512645e-1	5.501375e-1	8.220654e-2	8.670497e-1	"feasible"
39	7.512645e-1	5.501376e-1	8.220645e-2	8.670497e-1	"feasible"
40	7.682123e-1	5.834670e-1	5.819199e-2	9.305839e-1	"feasible"
41	7.682123e-1	5.834670e-1	5.819199e-2	9.305839e-1	"feasible"
42	7.682123e-1	5.834670e-1	5.819197e-2	9.305839e-1	"feasible"
43	8.733740e-1	7.475936e-1	3.910348e-2	1.321678e+0	"feasible"
44	8.733740e-1	7.475936e-1	3.910356e-2	1.321678e+0	"feasible"
45	8.733740e-1	7.475936e-1	3.910343e-2	1.321678e+0	"feasible"
46	8.651721e-1	7.501789e-1	1.845282e-2	1.311291e+0	"feasible"
47	8.651722e-1	7.501789e-1	1.845280e-2	1.311291e+0	"feasible"
48	8.651721e-1	7.501790e-1	1.845282e-2	1.311291e+0	"feasible"
49	9.043159e-1	8.163747e-1	9.354968e-3	1.484255e+0	"feasible"
50	9.043159e-1	8.163747e-1	9.354973e-3	1.484255e+0	"feasible"
51	9.043159e-1	8.163748e-1	9.354964e-3	1.484255e+0	"feasible"

Sample size: 71 Filtered: 71 Column width:

2D plot

Parallel coordinates

**pSeven Enterprise**

Disk optimization | SciPy Integration Report | RotatingDisk2020 | 37 | 38 | 39 | 40 | 41 | 42 | 43 | 44 | 45 | 46 | 47 | 48 | 49 | 50 | 51

Workflow contents

- [In]
- Ext. param Python script
- Optimizer Python script
- Rotating disk Composite
- [Out]

```
graph TD; In[In] --> ExtParam[Ext. param Python script]; ExtParam --> Optimizer[Optimizer Python script]; Optimizer --> Out[Out]; Optimizer --> Composite[Rotating disk Composite]; Composite --> Out
```

**pSeven Enterprise | AppsHub**

Search...

Turbine Disk Optimization 34 days ago 67

Bistable MEMS 34 days ago 34

Heat Exchanger 61 days ago 53

Single Cylinder 67 days ago 86

Run setup

Input parameters

Presets: Default [3 out of 6]

Filter

Results

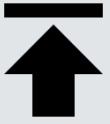
- @Step index
- @Error
- @Sandbox prototype
- @Sandbox path
- @Batch size
- @Batch index
- optimal\_mass
- optimal\_stress
- optimal\_x
- Rotating disk
- @Step index



- Facilitate **collaboration** between engineering teams **worldwide** thanks to cloud-native solution
- Improve **accessibility** of pSeven, external software and computational resources thanks to modern scalable architecture



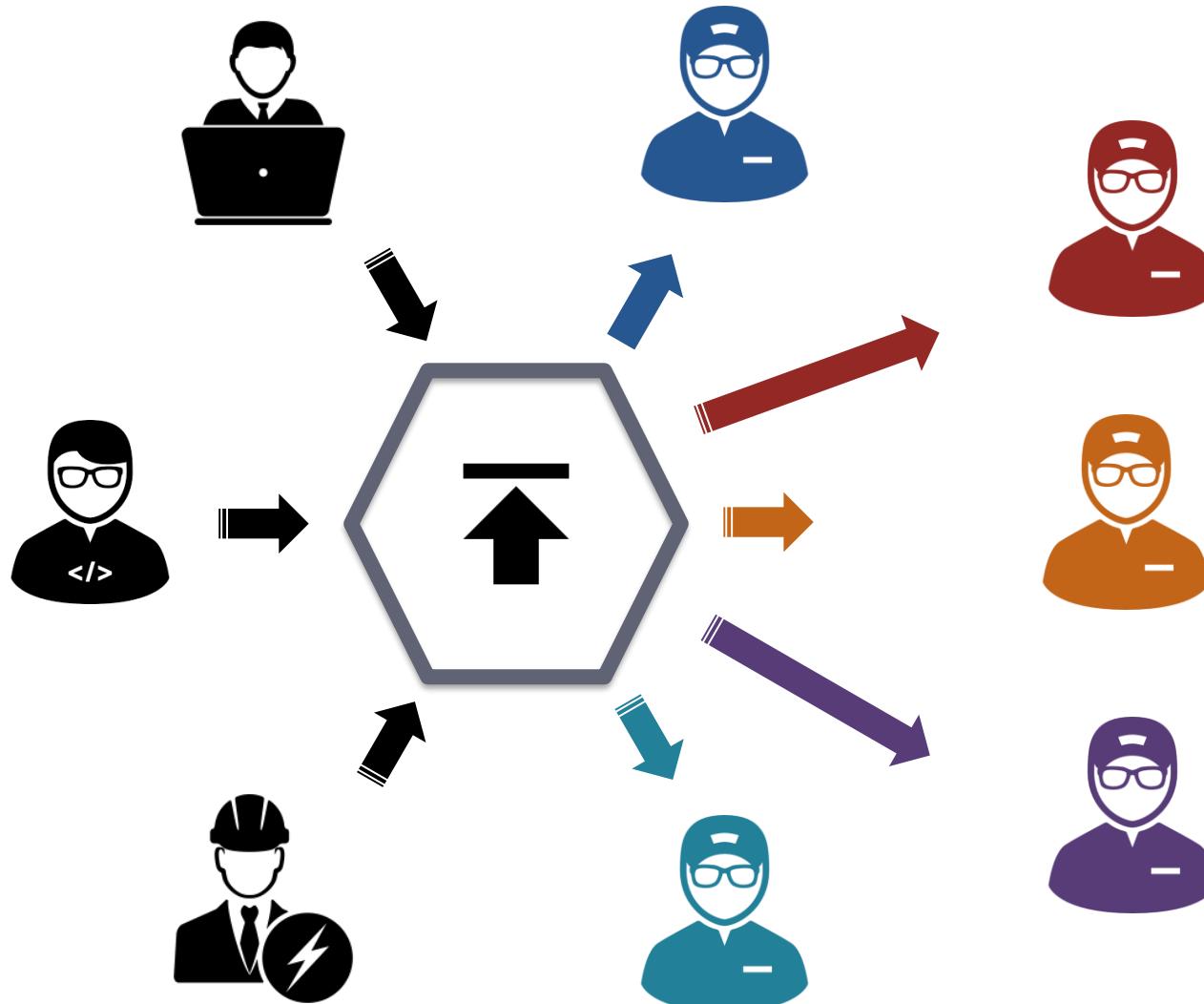
- Democratization through **publishing** of workflow-powered web applications and services
- Easy extension of platform capabilities and **integration** within existing ecosystem through **open APIs**



# Create, Publish and Run Apps and Services, and Democratize

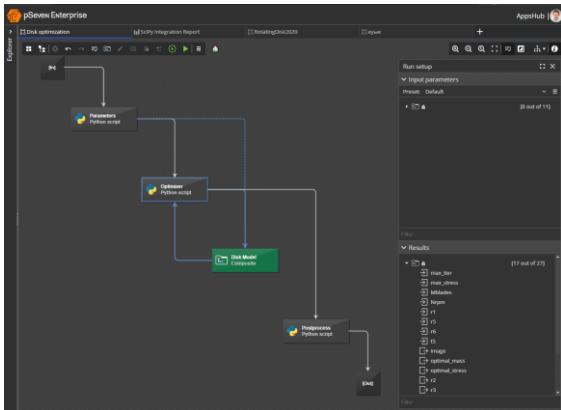


M&T Experts & Developers      Domain Experts      Suppliers/Partners



# Create and Publish Workflow-powered Web Applications using AppsHub

## M&T EXPERTS (sim experts or data scientists)



Create automated workflows

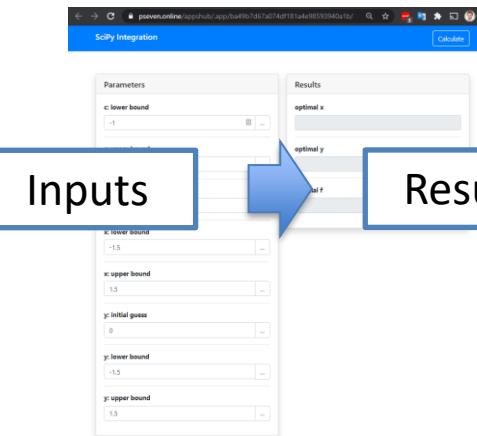
Setup simulations, load data, tune algorithms, ...



Package and Publish fully automated workflows to **AppsHub**

Experts can build Custom UI themselves, using any web UI framework

## DOMAIN EXPERTS



Inputs

Results



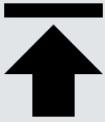
Run Workflow-Powered Apps using auto generated UI

Specify inputs, run and analyze results, ...



# pSeven Enterprise

DATADVANCE



# Run Apps – AppsHub Examples



## Co-authoring UI

The screenshot shows a web-based interface for a 'RotatingDisk2020\_04.p7wf' workflow. On the left, a sidebar lists various app blocks: Block library, Exploration, General, Integration, Modeling, and ANSYS Workbench. The main workspace displays a workflow diagram with nodes like 'Rotating Disk Demo Workflow', 'Rotating disk Composite', and 'Optimizer Python script'. A 'Run setup' panel on the right shows parameters such as 'shaft\_radius' (0.12), 'stress\_max' (2975), and 'max\_iter' (10). The interface includes a 'Results' section with 17 items, including 'Rotating disk' and 'Optimizer'. At the bottom, browser icons for Safari, Firefox, Chrome, and Edge are shown.



## AppsHub UI

The screenshot shows the AppsHub UI with a dark theme. It features a grid of cards for different engineering applications: 'L/D Modeling' (32 days ago), 'Turbine Disk Optimization' (34 days ago), 'Bistable MEMS' (34 days ago), 'Three-Section Beam' (58 days ago), 'Heat Exchanger' (63 days ago), and 'Single Cylinder' (63 days ago). Each card includes a thumbnail image, the application name, the author's email, and a brief description. At the bottom, browser icons for Safari, Firefox, Chrome, and Edge are shown. A small icon of a person with three thought bubbles is in the top right corner.

## What's next?



- You are encouraged to use pSeven in your Skoltech design and data analytics projects
- You are also encouraged to use pSeven in any other projects you just plan and already do
  - We will provide you and your team with the student licenses for free!
  - Key URLs
    - [www.datadvance.net](http://www.datadvance.net)
    - <https://www.datadvance.net/download/>
    - <https://www.datadvance.net/user/protected/license-request/>
- Welcome to contribute!
  - Prototype and develop new algorithms right within the platform
  - Embed software tools
  - Build vertical apps with low-code platform – they will become part of our publicly available AppsHub
  - Join <https://pseven.online/pseven/> to play/test/contribute.



[1 - Workflow Basics](#)

[2 - Integration capabilities basics](#)

[3 - Design Exploration. Design of Experiments](#)

[4 - Design Exploration. Optimization](#)

[5 - Analyze Basics](#)

[6 - Predictive Modeling. Model Builder](#)

[7 - Predictive modeling. Model Validator and Model Explorer](#)



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