



# Multi-Objective Design Optimization of a Hybrid Electric Vehicle

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# Multi-Objective Design Optimization of a Hybrid Electric Vehicle

1. Setup and HEV Model
2. Automation and Process Integration
3. Selection of Parameters
4. Optimization Strategy
5. Conclusion and Results

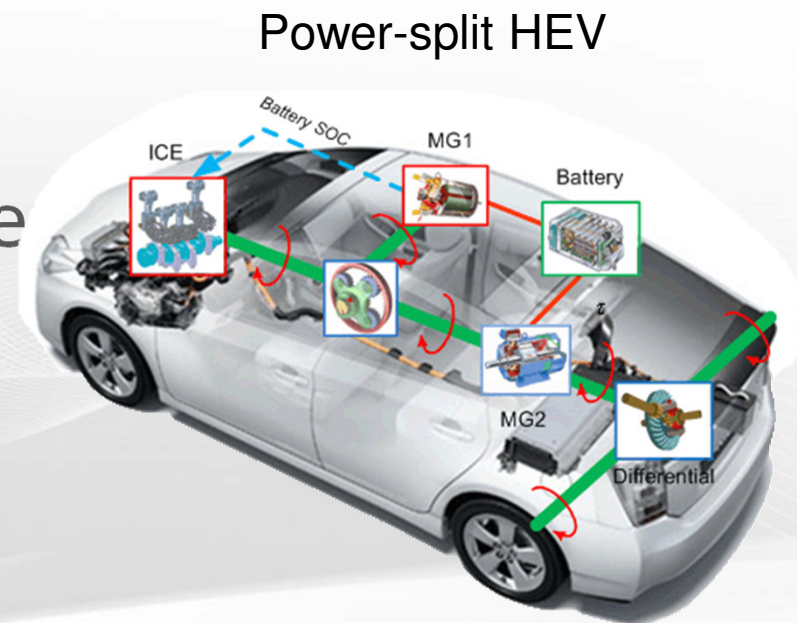
# Process Integration and Optimization of a Mathematics-based HEV

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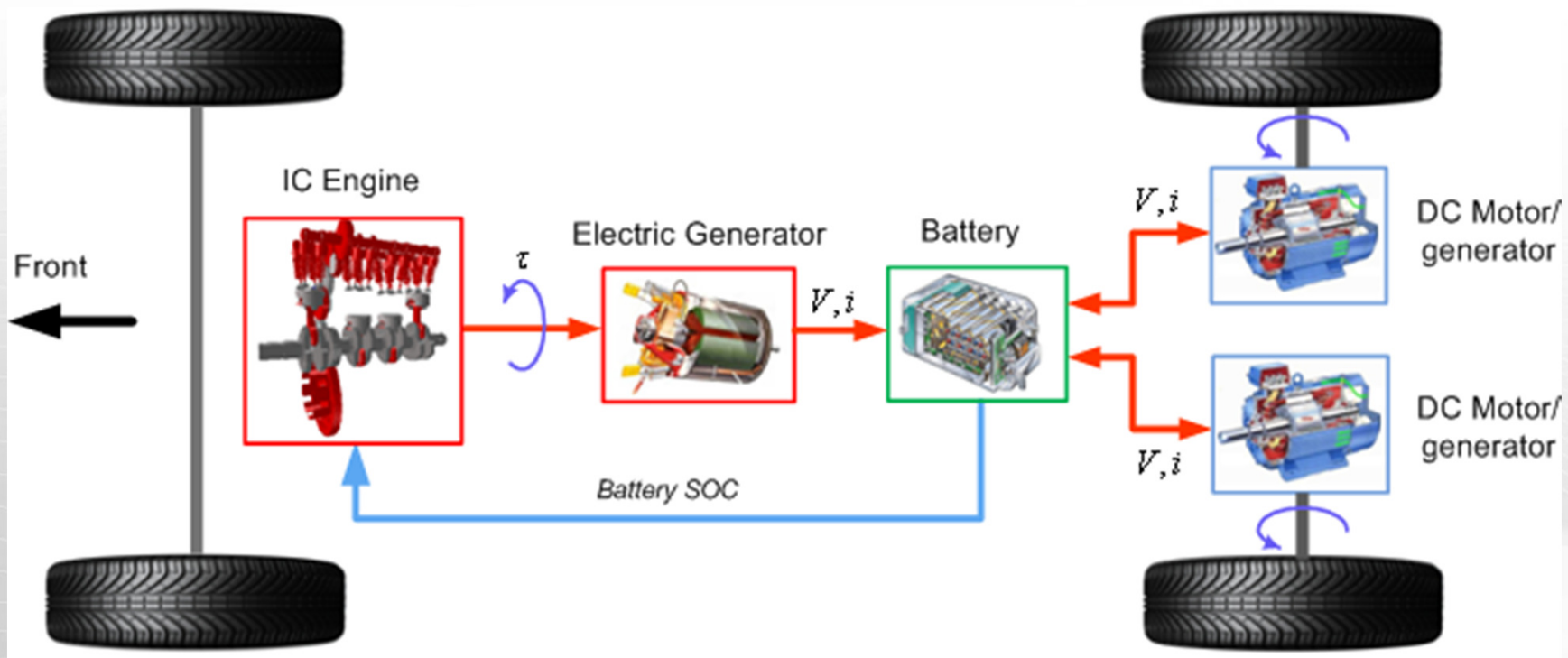
# 1. HEV Challenges

- Many new components
- Less experience
- More interaction of more components
- New thermal effects
- More need for sharing

→ More need for optimization



# 1. Hybrid and E-Vehicle Components

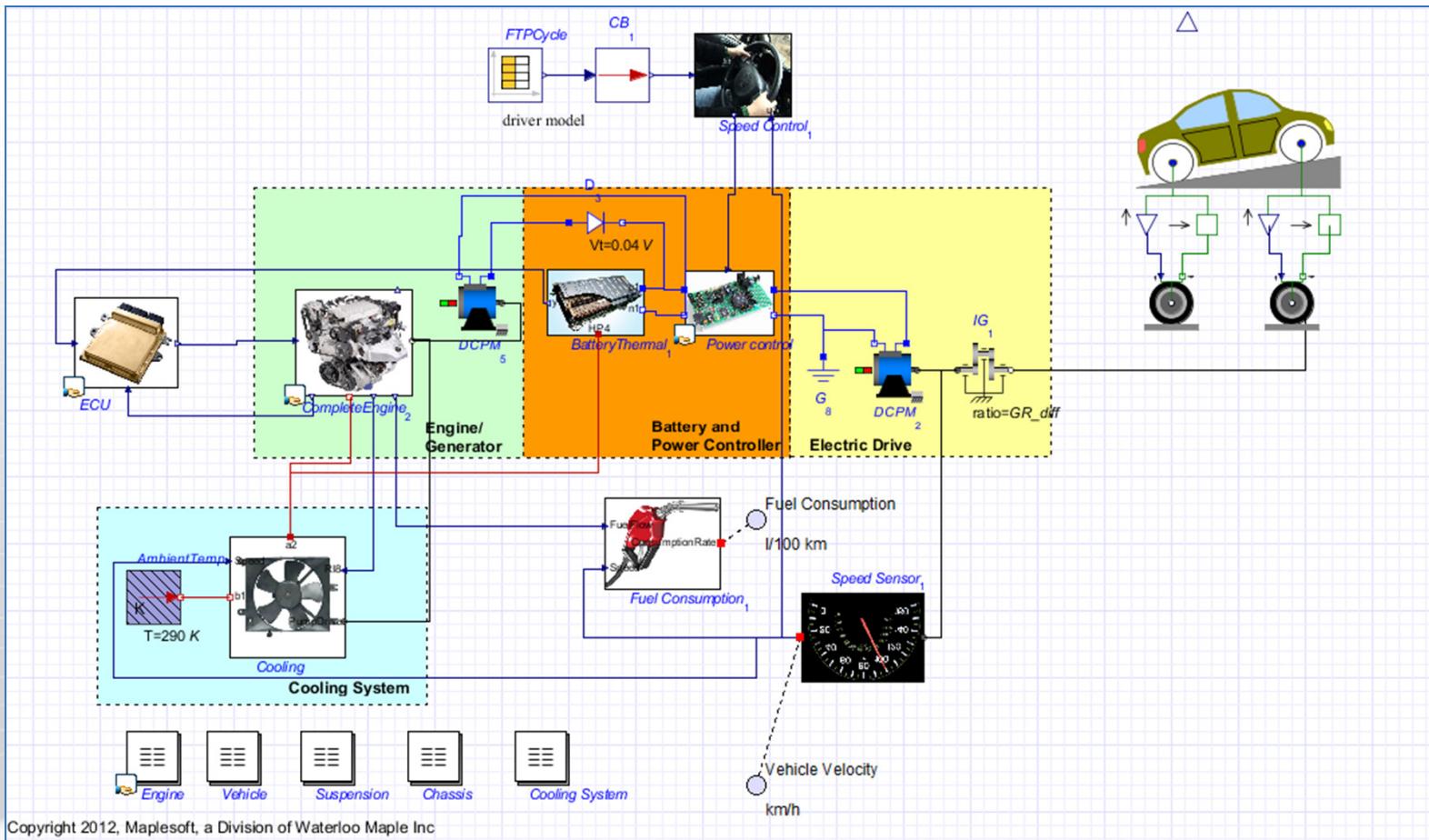




# 1. Variable Parameters

Parameter	Description [unit]	Range
<b>MinSoCThreshold</b>	Minimum SOC Threshold	0.3 – 0.6
<b>ncell</b>	Number of battery cells	100 – 200
<b>Vmanifold</b>	Engine manifold volume [m <sup>3</sup> ]	0.003 – 0.005
<b>Bore</b>	Engine bore [m]	0.0855 – 0.1
<b>Stroke</b>	Engine stroke [m]	0.0814 – 0.19
<b>Apos</b>	Area of battery pos.electrode [cm <sup>2</sup> ]	100 – 500
<b>Aneg</b>	Area of battery neg.electrode [cm <sup>2</sup> ]	100 – 500
<b>apos</b>	Specific surface of Apos [cm <sup>2</sup> /cm <sup>3</sup> ]	3000 – 5000
<b>aneg</b>	Specific surface of Aneg [cm <sup>2</sup> /cm <sup>3</sup> ]	2000 – 4000
<b>Va</b>	Nominal voltage of e-motor [V]	400 – 1000
<b>Ia</b>	Nominal current of e-motor [mA]	50 – 100
<b>Vas</b>	Phase voltage of e-motor [V]	50 – 150
<b>Ias</b>	Phase current of e-motor [mA]	50 – 150

# 1. Setup and HEV Model



# 1. Setup and HEV Model

Goal: the optimization goal is to minimize fuel consumption, while simultaneously keeping the maximum torque at a maximum

- Two objective functions have to be considered
- Multi-objective optimization strategies have to be applied



# 1. Setup and HEV Model

Constraints: Two target requirements have to be considered during the optimization process

- Maximum temperature  $< 315$  K
- Battery current  $< 500$  A

→ Violation of these constraints leads to a non feasible design and it will be marked as a failed experiment

# 1. Output Parameters

<b>Parameter</b>	
<b>Power loss</b>	Output parameter

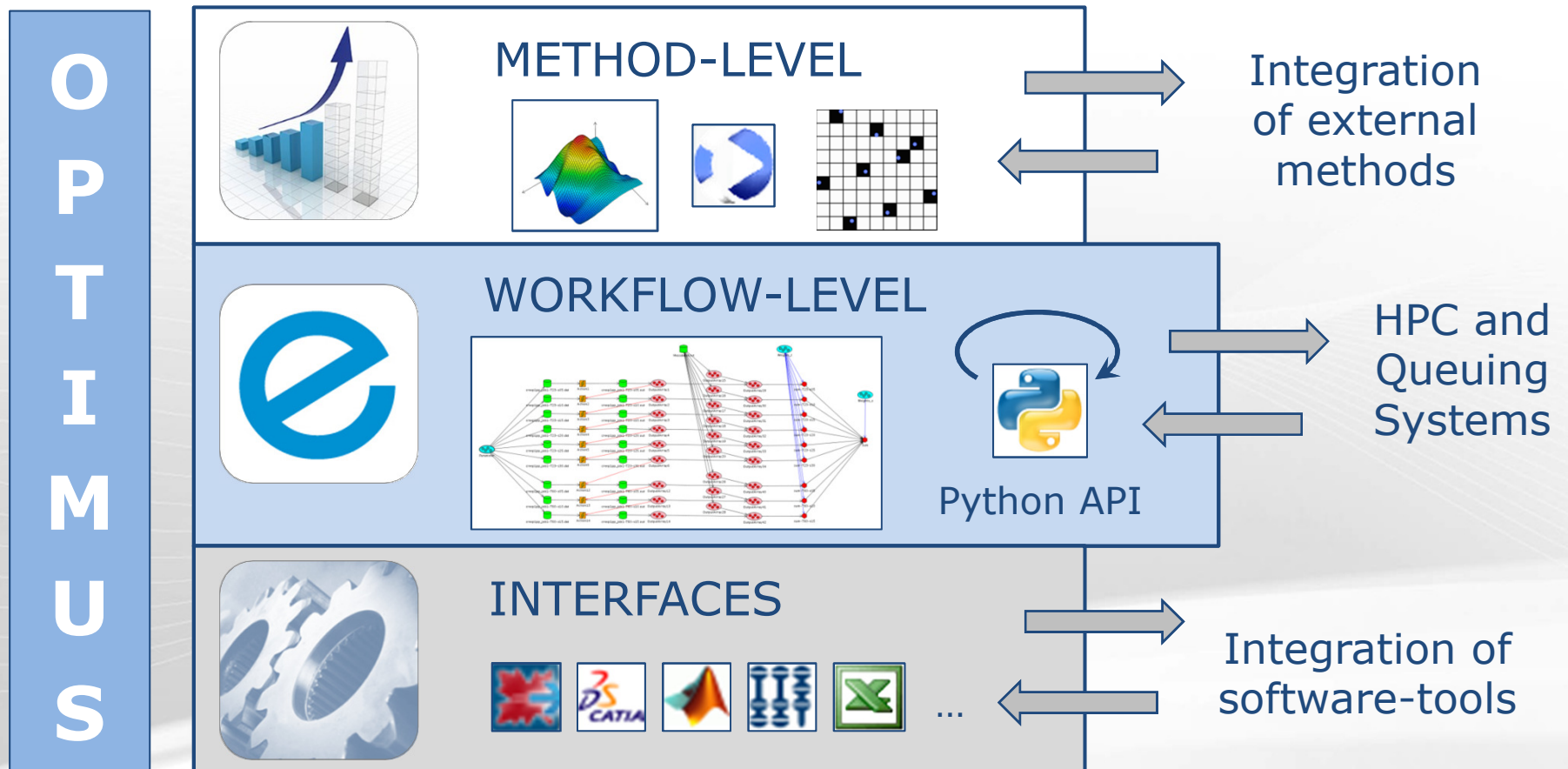
<b>Objective</b>	
<b>Fuel consumption</b>	Minimization
<b>Engine torque</b>	Maximization

<b>Constraint</b>	
<b>Battery temperature</b>	Smaller than 315 K, because higher values can damage the durability of the battery ( $< 315$ )
<b>Battery current</b>	Smaller than 500 A, because higher values can damage the durability of the battery ( $< 500$ )

# Process Integration and Optimization of a Mathematics-based HEV

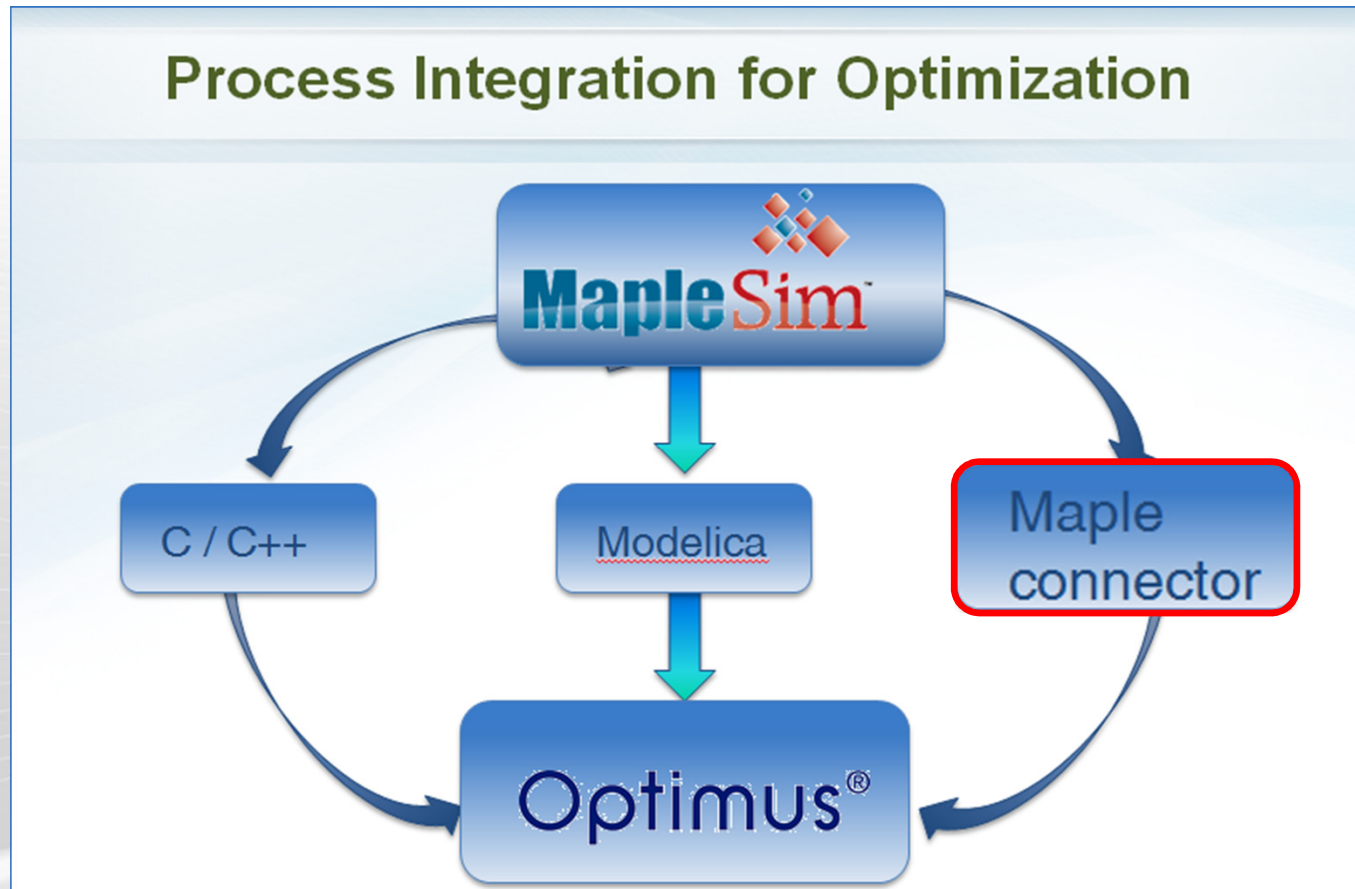
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# 2. Automation and Process Integration



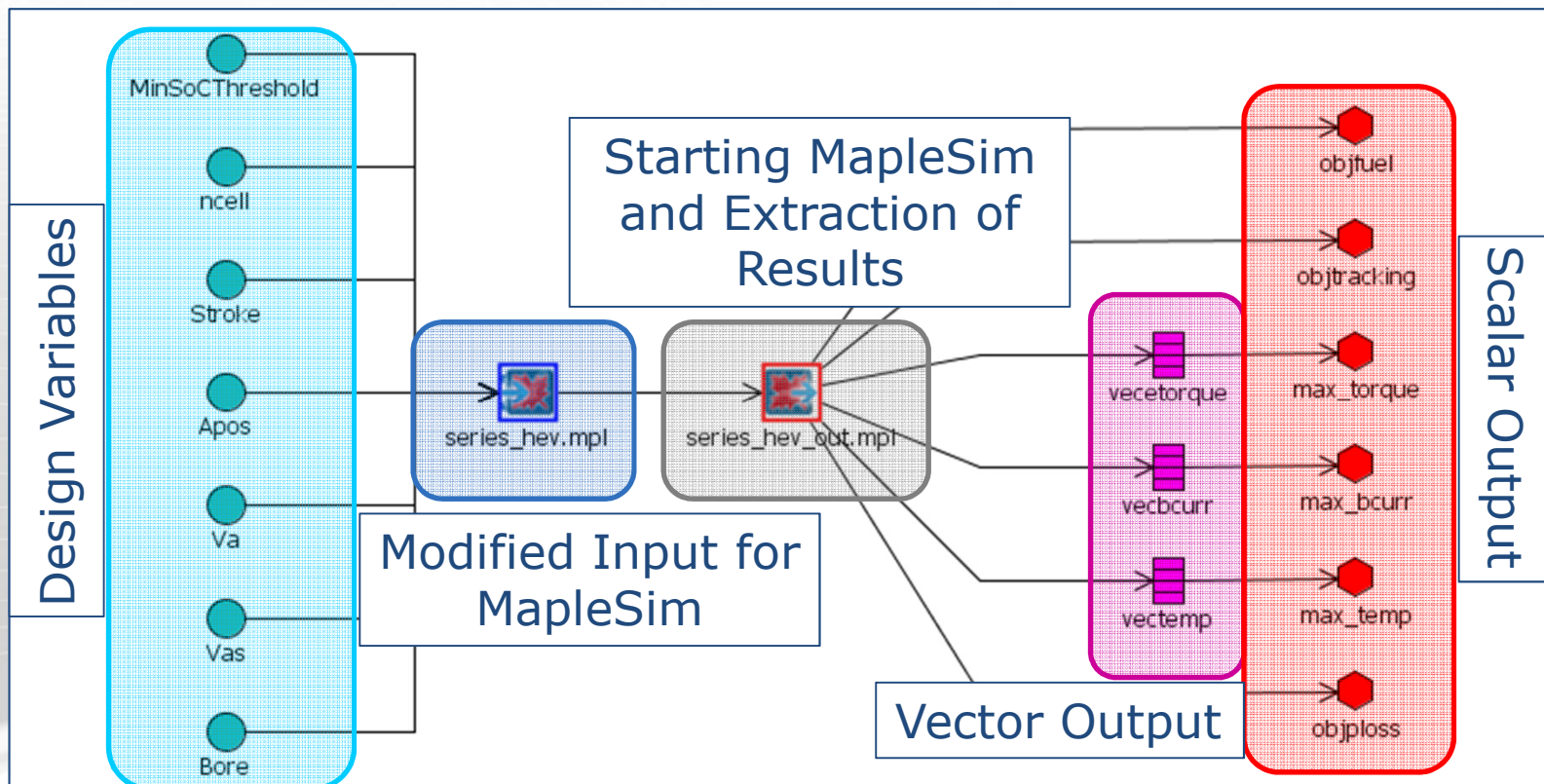


## 2. Automation and Process Integration



# 2. Automation and Process Integration

## Workflow in OPTIMUS



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# 3. Selection of Parameters

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apos	Specific surface area of battery positive electrode [cm <sup>2</sup> /cm <sup>3</sup> ]	3000 – 5000
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Va	Nominal voltage of electric motor [V]	400 – 1000
Ia	Nominal current of electric motor [V]	50 – 100
Vas	Nominal voltage of electric motor [V]	50 – 150
Ias	Nominal current of electric motor [V]	50 – 150



# 3. Selection of Parameters

## Linear Correlation Factors

$\geq 0.8$

Pearson (Spearman)	MinSoThreshold	ncell	Vmanifold	Boté	Stroke	Apos	Aneg	apos	aneg	Va	Ia	Vas	Ias
objfuel	<b>0.377</b> (0.429)	-0.087 (-0.239)	0.013 (-0.006)	0.023 (0.002)	0.029 (0.011)	<b>-0.521</b> (-0.611)	-0.015 (0.010)	0.000 (-0.028)	-0.021 (-0.020)	0.014 (0.017)	0.017 (0.009)	<b>-0.406</b> (-0.422)	0.042 (0.003)
objtracking	-0.120 (-0.074)	-0.090 (-0.037)	0.004 (-0.004)	0.009 (0.022)	-0.065 (-0.025)	-0.076 (-0.049)	0.003 (-0.005)	0.046 (0.012)	0.001 (0.002)	<b>-0.861</b> (-0.961)	-0.013 (-0.006)	0.013 (0.013)	0.011 (0.006)
objploss	0.027 (0.087)	0.241 (0.264)	-0.010 (-0.010)	0.005 (-0.018)	-0.028 (-0.069)	-0.087 (-0.144)	0.000 (0.019)	-0.002 (-0.008)	-0.044 (-0.010)	0.005 (0.001)	0.047 (0.001)	<b>-0.613</b> (-0.865)	0.065 (0.043)
objcurr	-0.127 (-0.165)	-0.252 (-0.245)	-0.027 (-0.035)	0.156 (0.164)	0.482 (0.563)	0.126 (0.174)	0.003 (0.003)	-0.026 (-0.024)	-0.032 (-0.022)	0.016 (0.006)	0.006 (0.013)	-0.453 (-0.458)	0.000 (-0.003)
objtemp	0.011 (0.070)	0.273 (0.273)	-0.017 (0.003)	-0.019 (-0.038)	-0.065 (-0.082)	0.019 (0.046)	0.017 (0.016)	-0.010 (-0.006)	-0.028 (0.008)	0.012 (0.026)	0.046 (0.024)	<b>-0.698</b> (-0.797)	0.047 (0.033)
objtorque	-0.117 (-0.124)	-0.232 (-0.224)	-0.025 (-0.027)	0.173 (0.159)	0.553 (0.555)	0.153 (0.156)	-0.002 (-0.008)	-0.021 (-0.017)	-0.029 (-0.032)	0.015 (0.014)	0.009 (0.013)	0.521 (0.533)	-0.015 (-0.018)

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# 4. Optimization Strategy

1. Decide for an optimization strategy

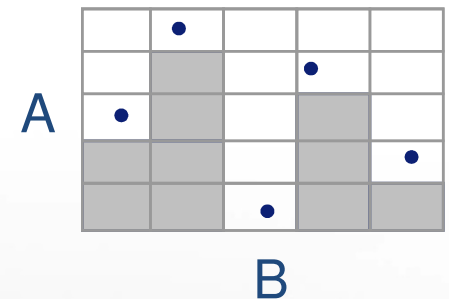
2. Study: Investigate the behavior of the system and the Pareto points

3. Study: Investigate the influence of swarm size on mPSO

# 4. Optimization Strategy

## STEP1 Design of Experiments

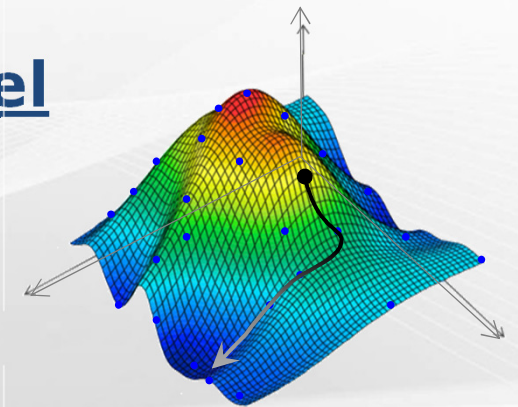
Exploration of the design space  
Latin Hypercube Sampling with n experiments



QUALITY

## STEP2 Response Surface Model

Describing the system behavior using  
mathematical models (Kriging)



VERIFY

## STEP3 Multiobjective Optimization

Determine an optimal parameter combination on  
the model using evolution strategies  
(NSEA+/mPSO algorithm)



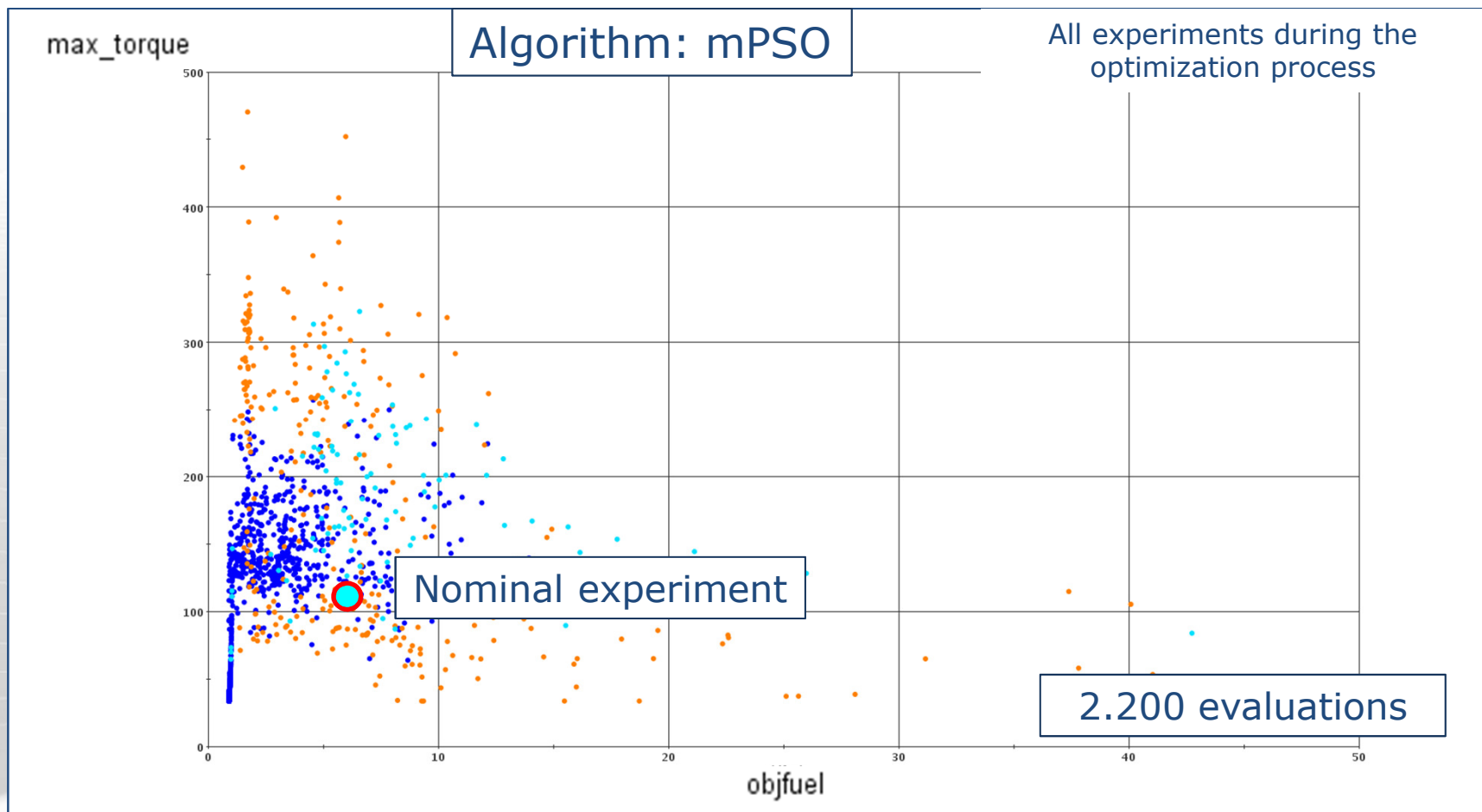
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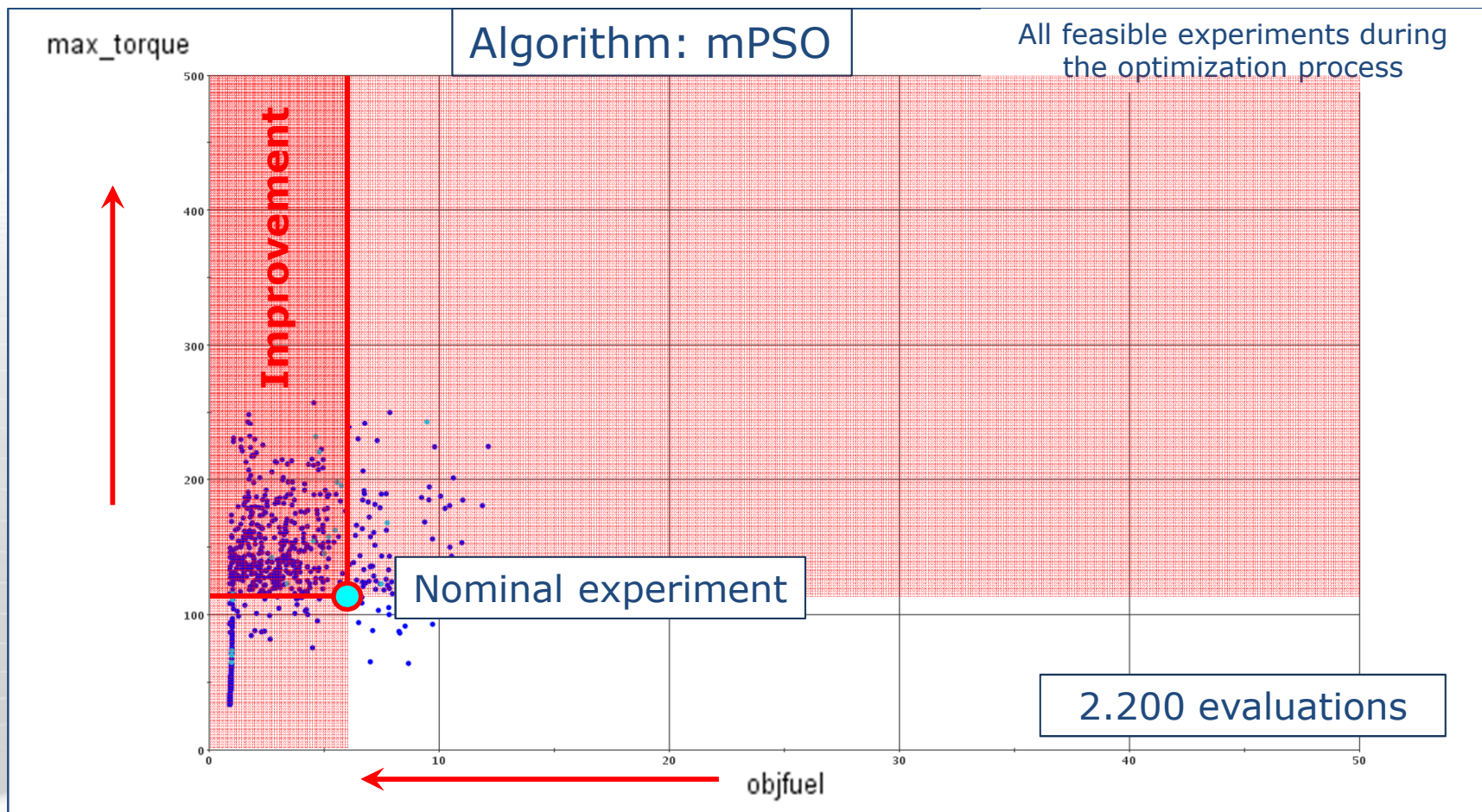
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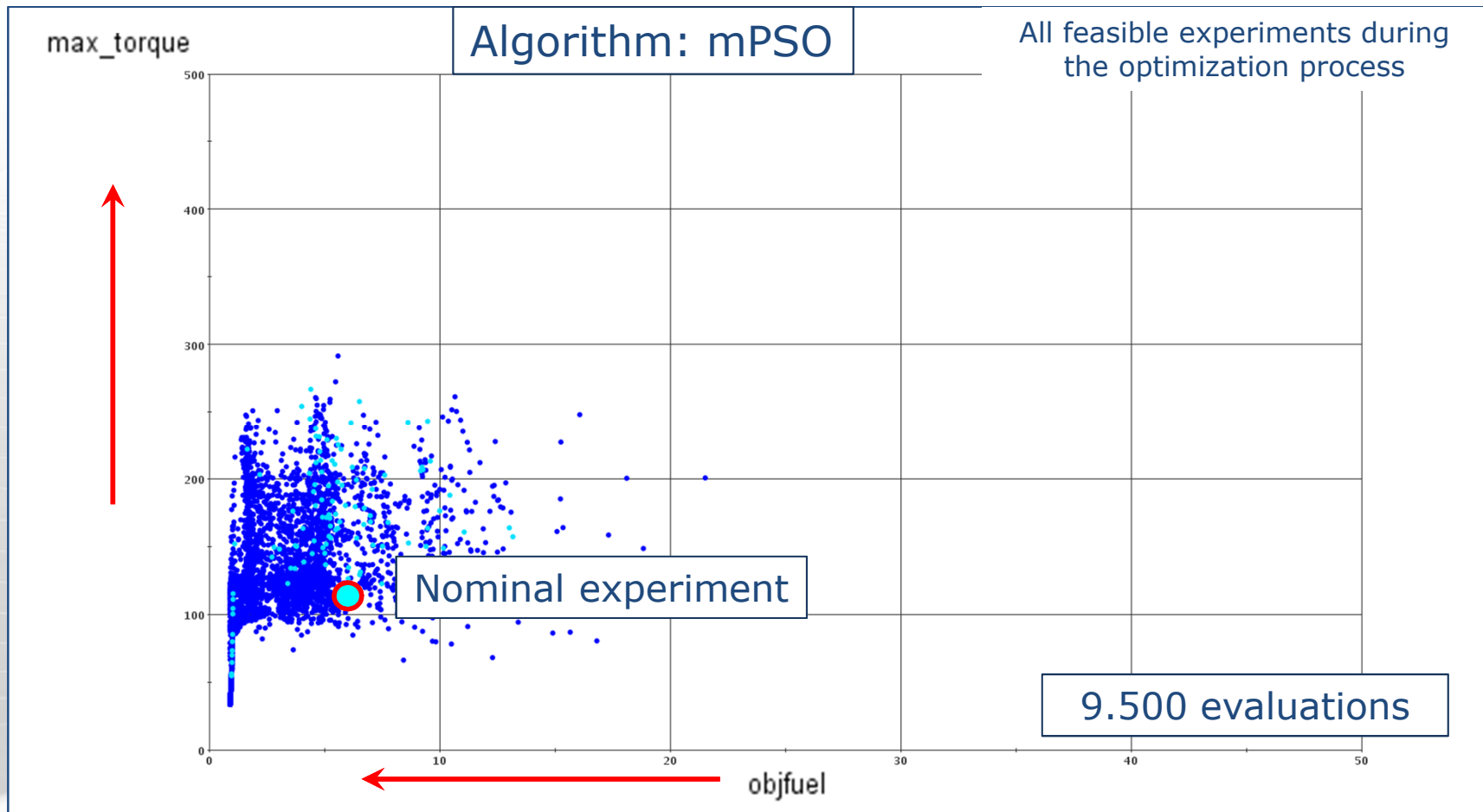
# 4. Optimization Strategy



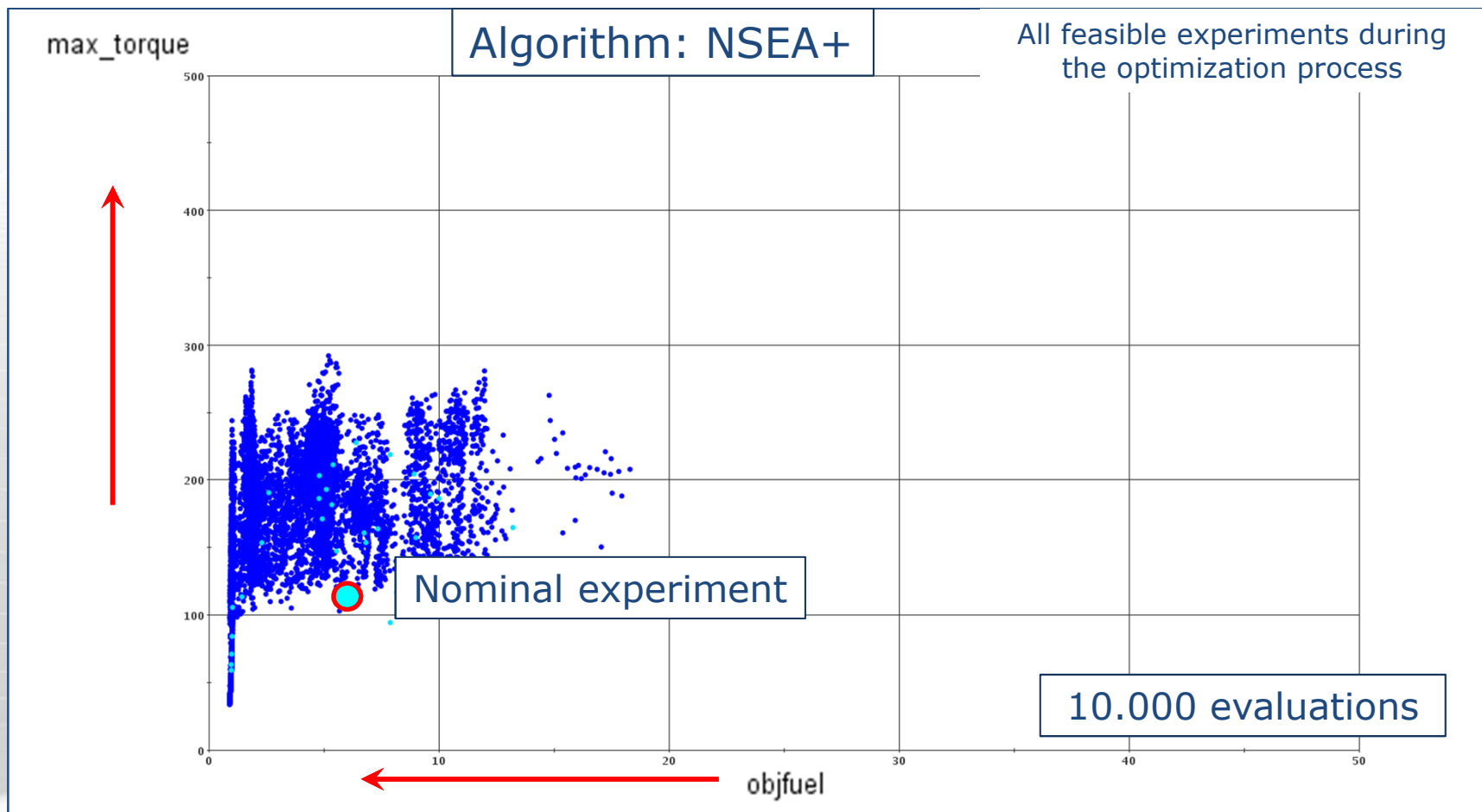
# 4. Optimization Strategy



# 4. Optimization Strategy

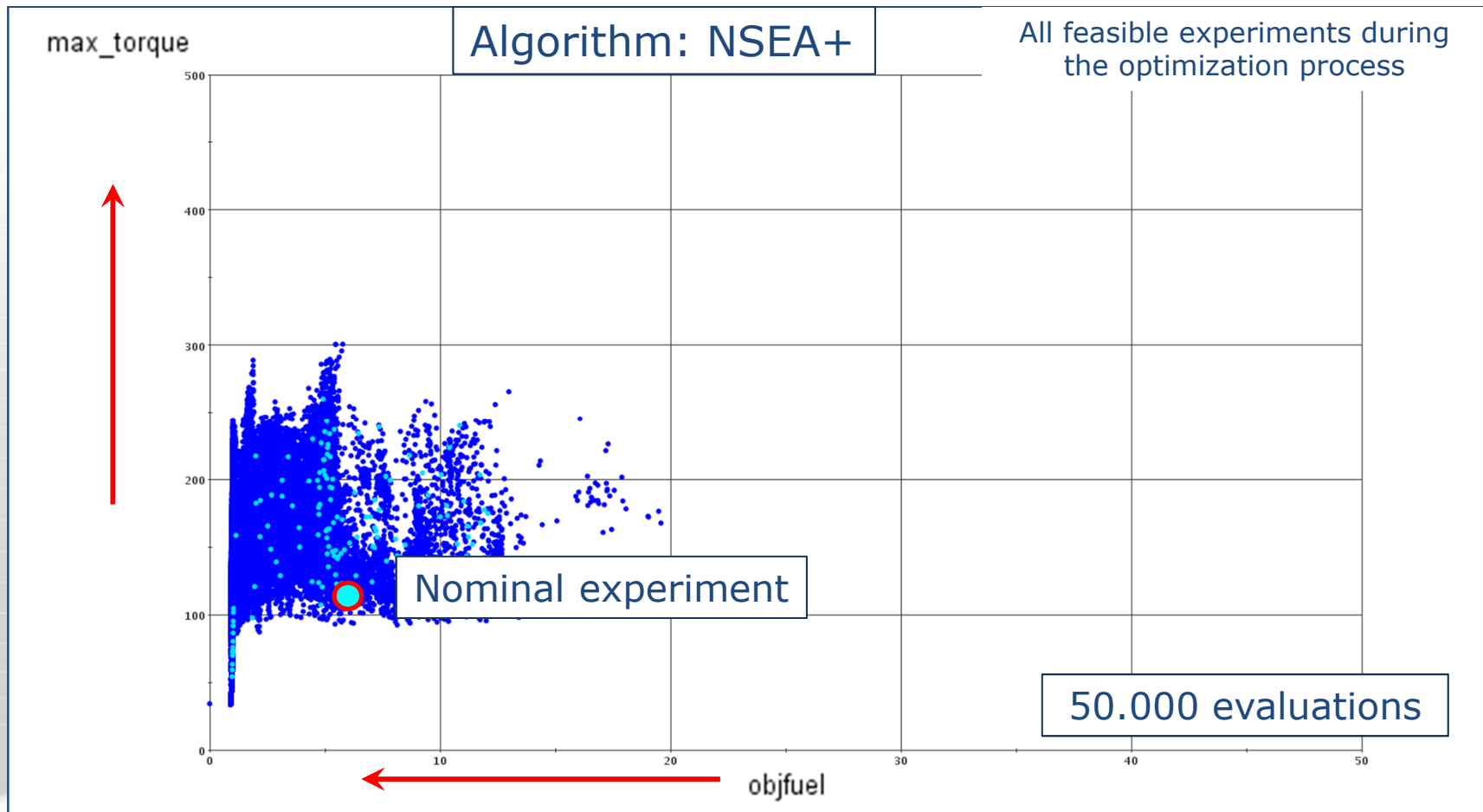


# 4. Optimization Strategy

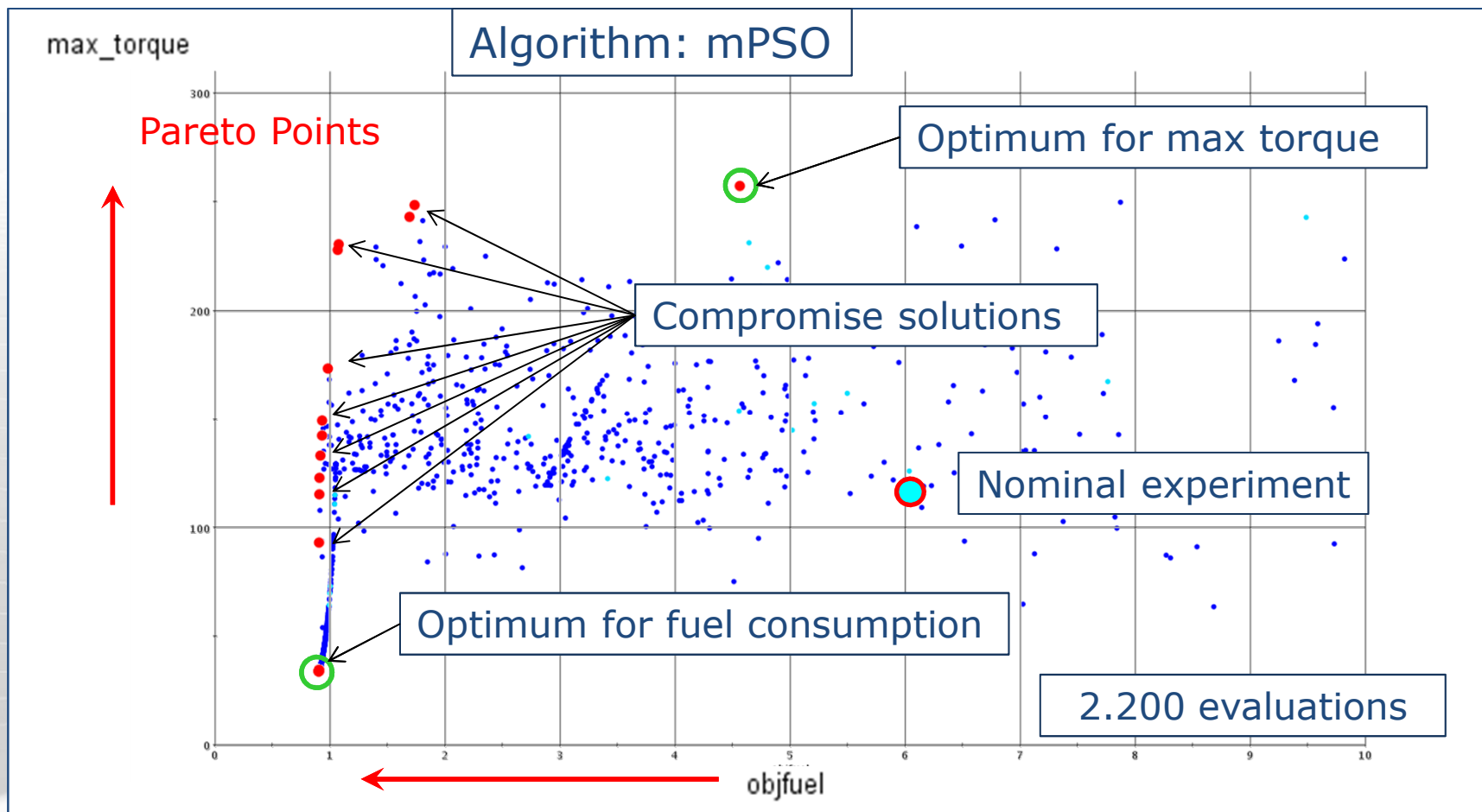




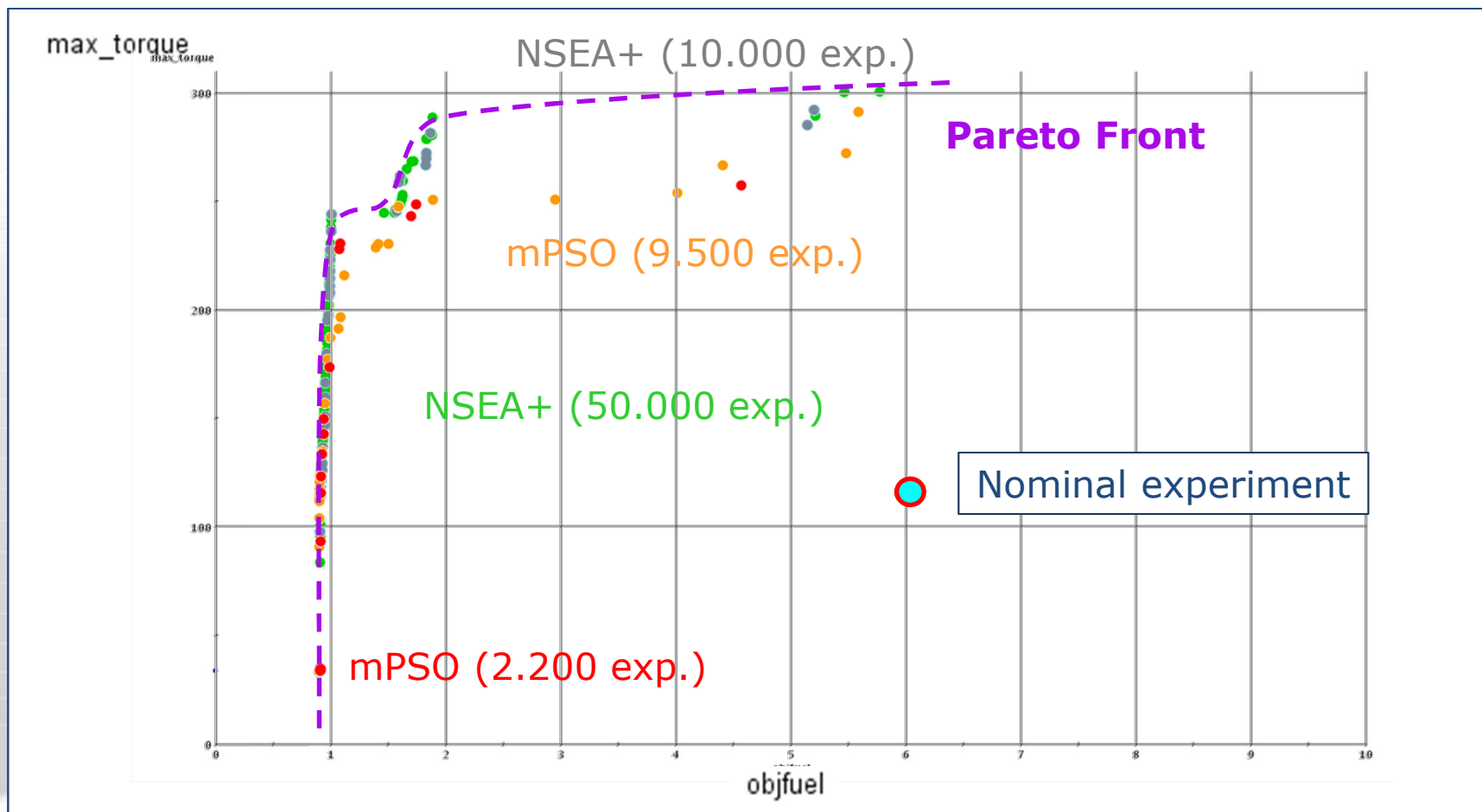
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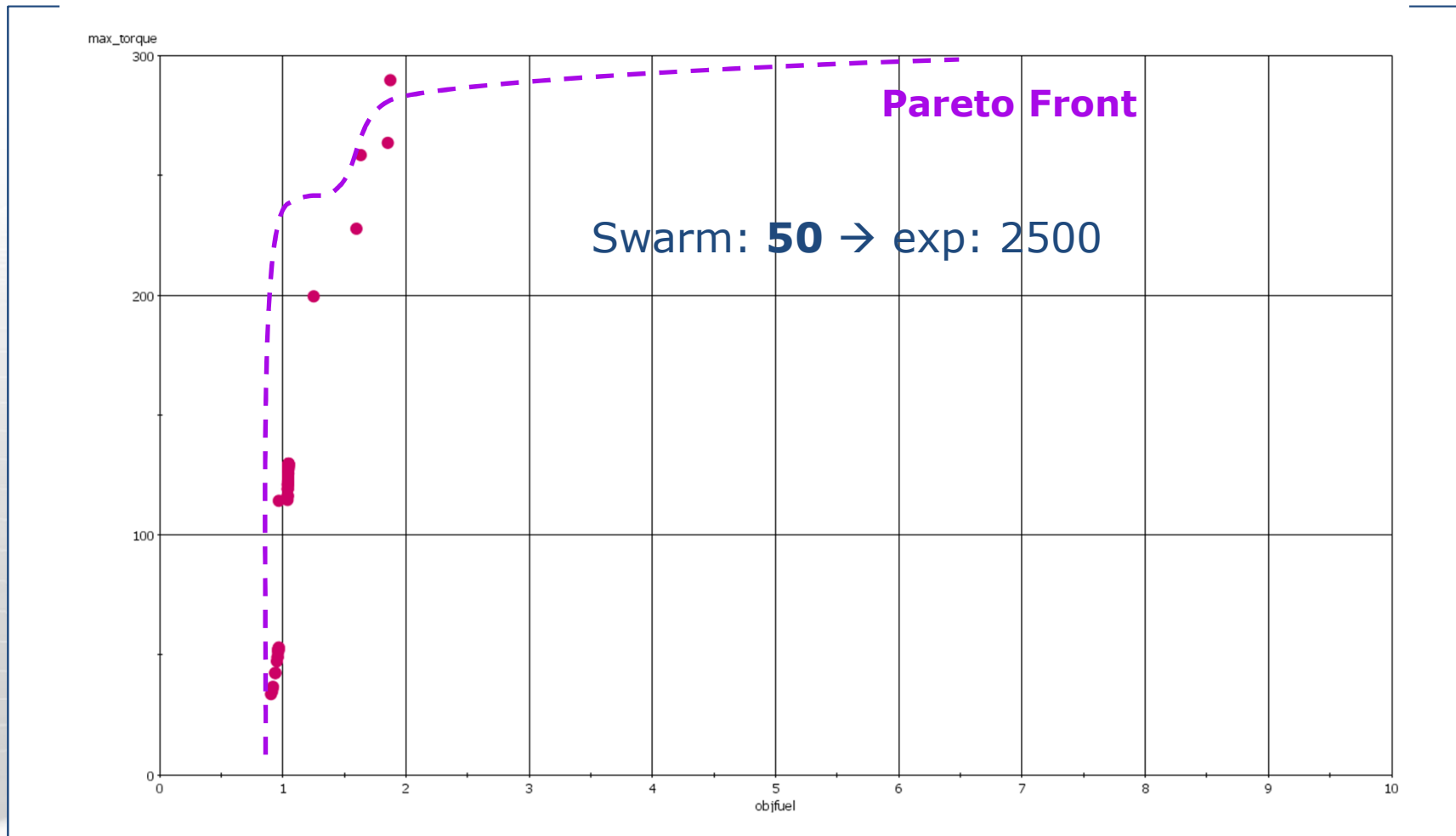


# 4. Optimization Strategy

1. Decide for an optimization strategy
2. Study: Investigate the behavior of the system and the Pareto points

3. Study: Investigate the influence of swarm size on mPSO

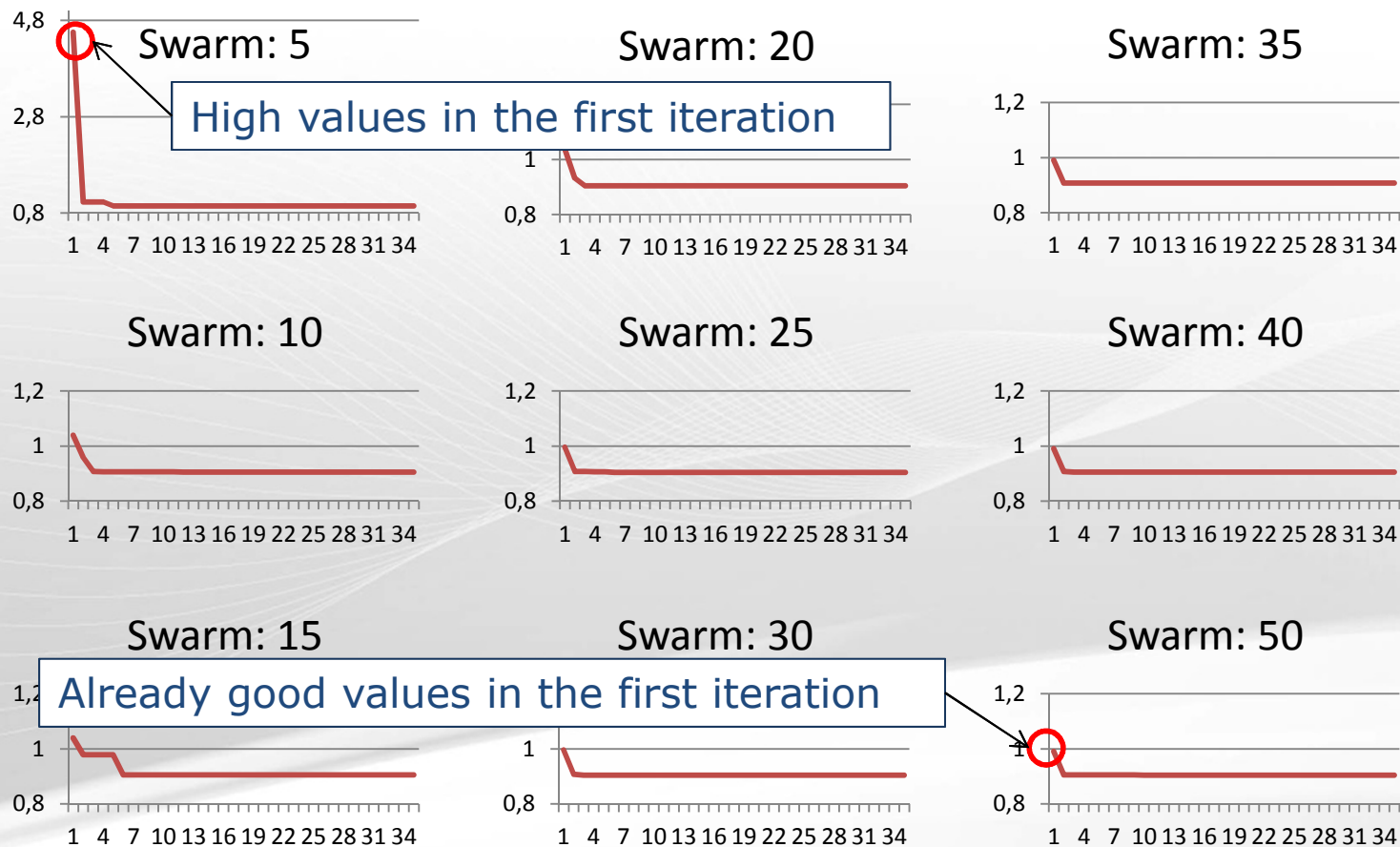
# 4. Optimization Strategy





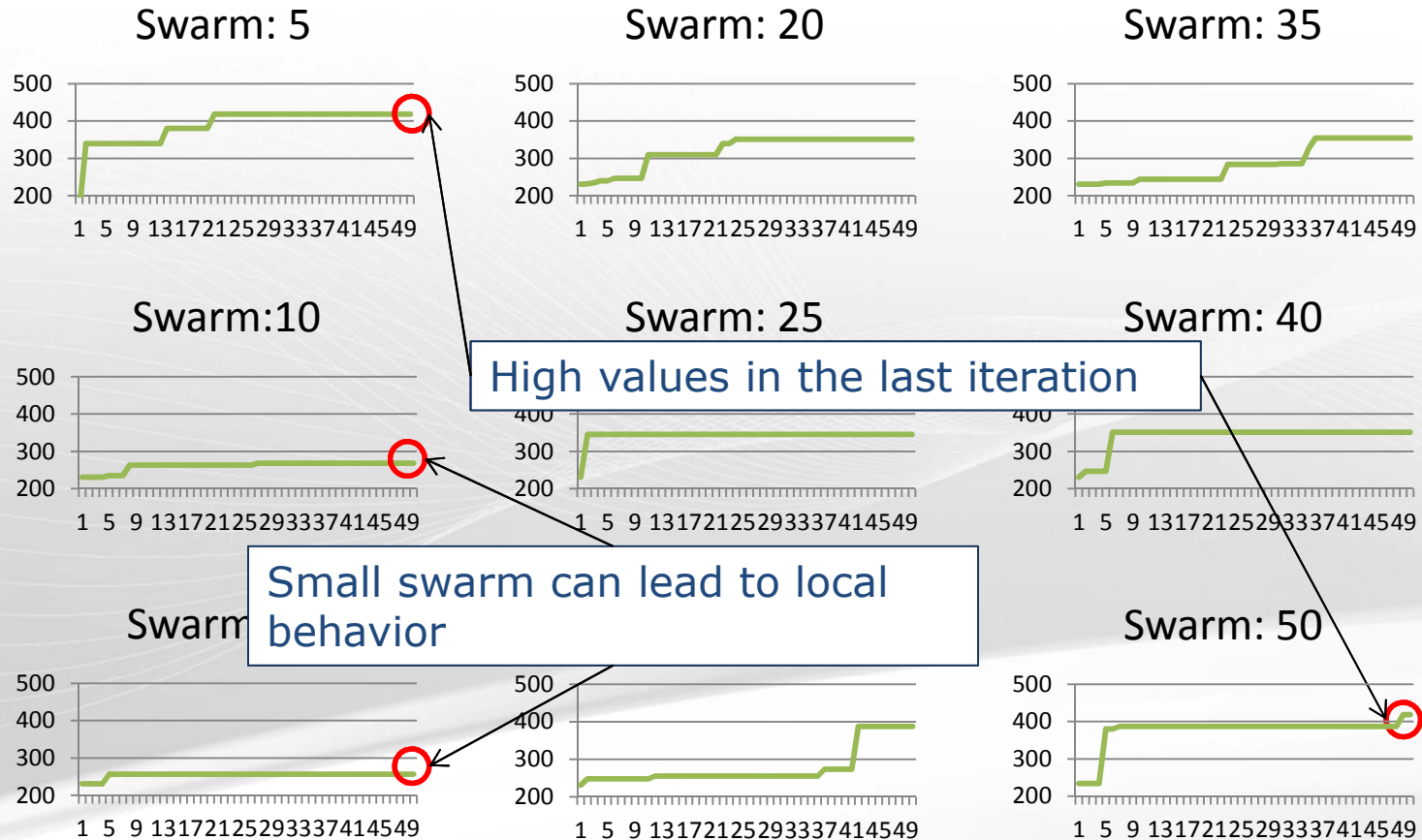
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Best objfuel value



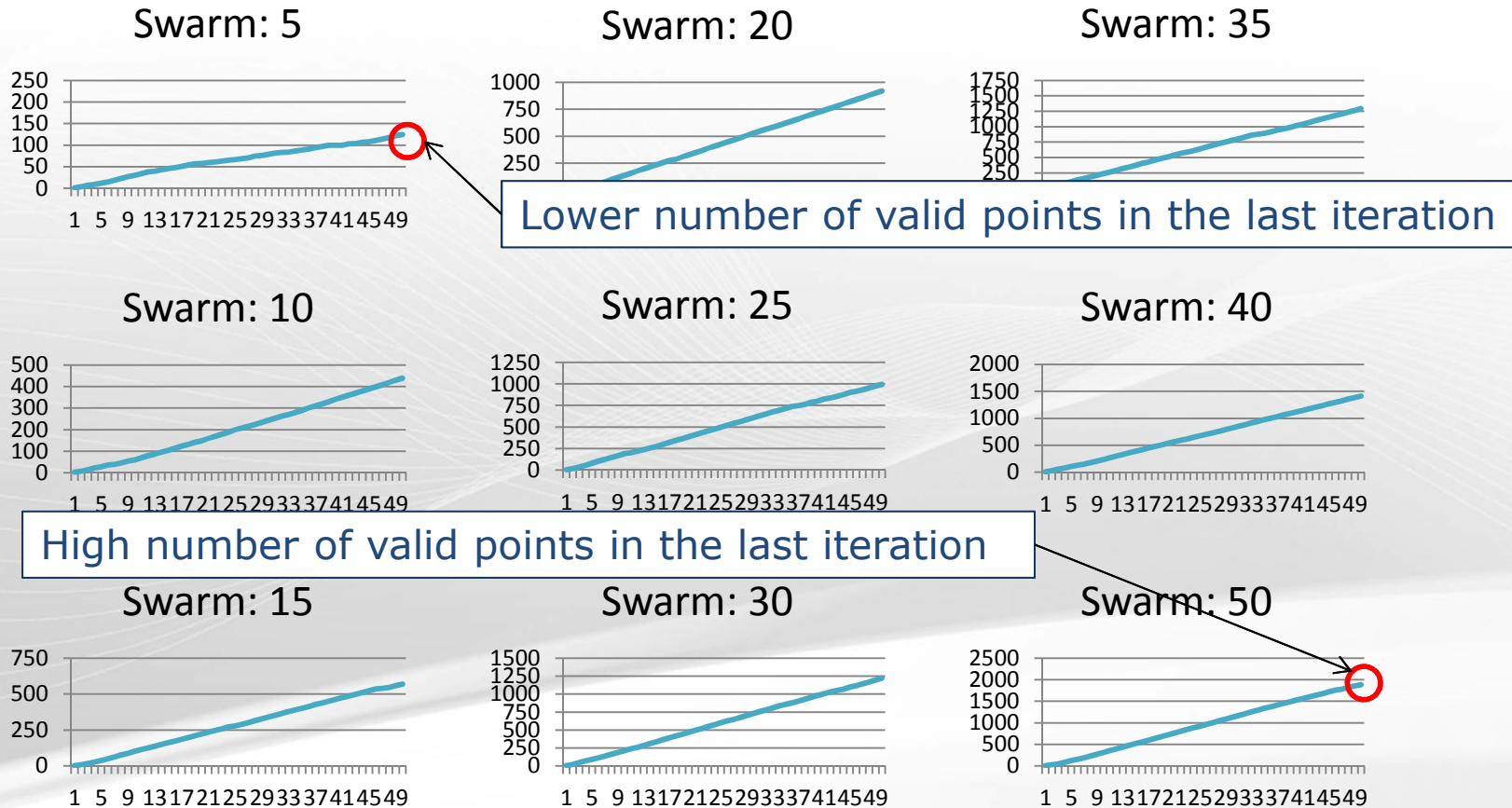
# 4. Optimization Strategy

Best max\_torque value



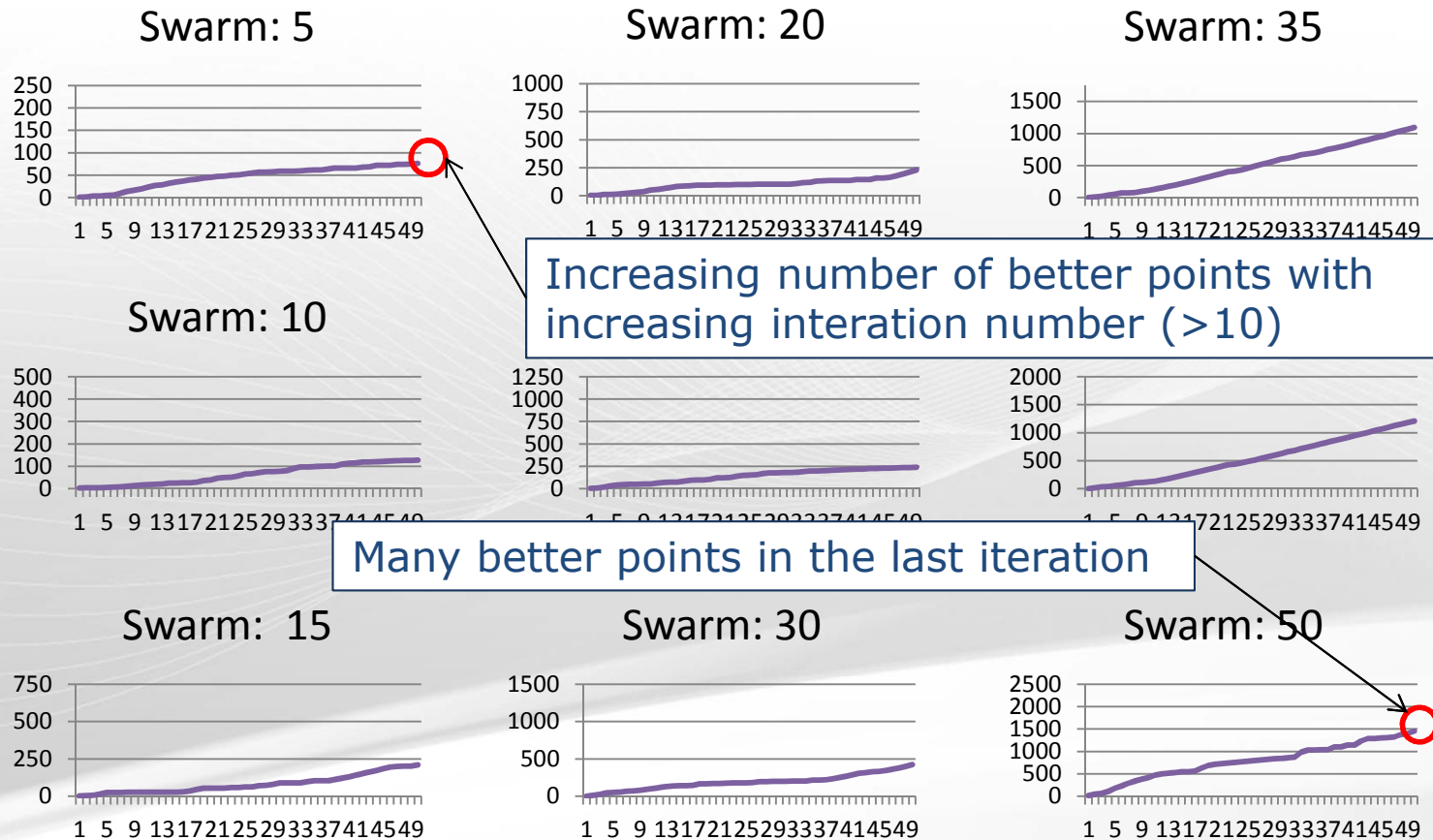
# 4. Optimization Strategy

# valid points



# 4. Optimization Strategy

# better points (compared to nominal)

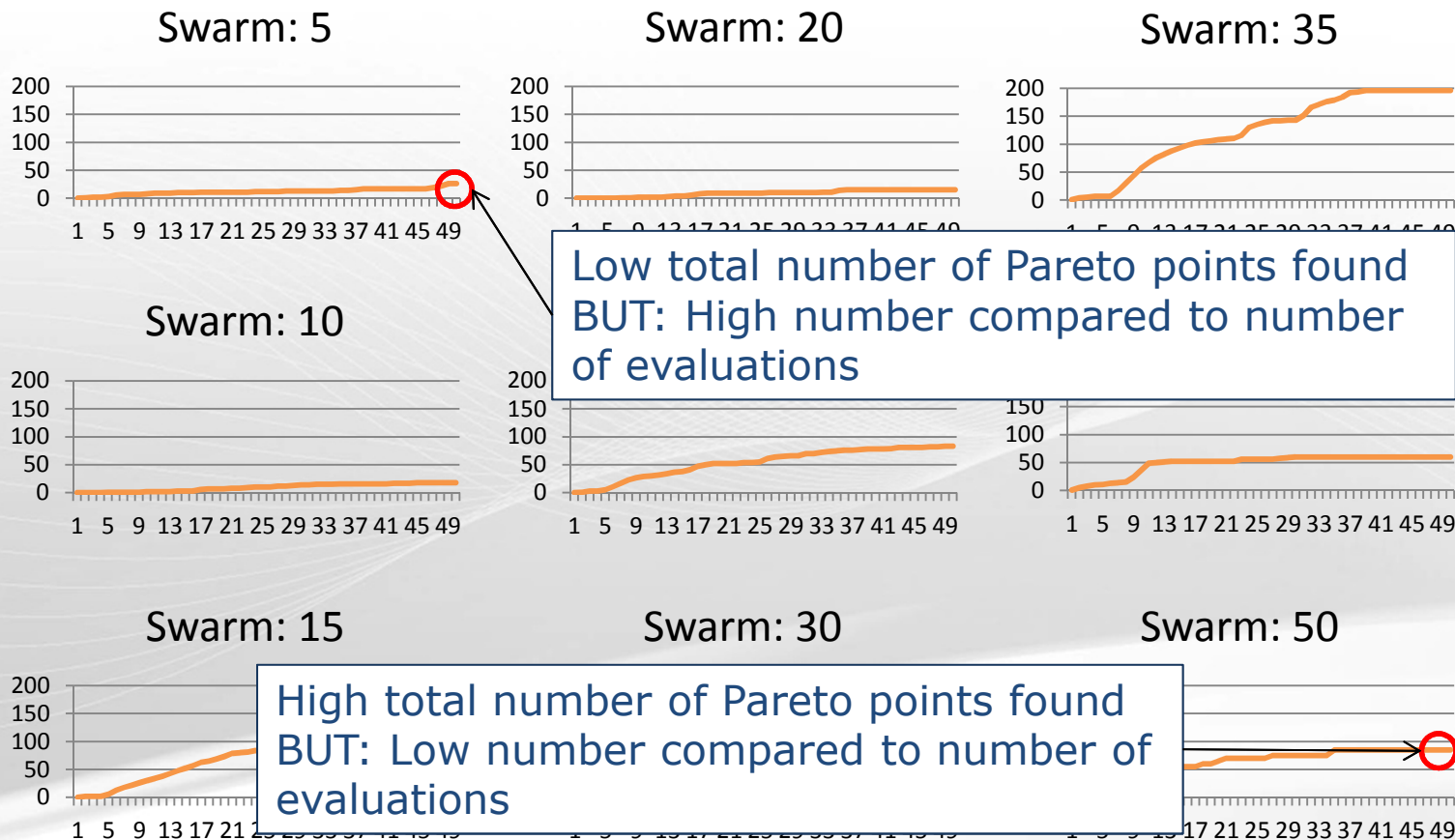


Increasing number of better points with increasing iteration number (>10)

Many better points in the last iteration

# 4. Optimization Strategy

# Pareto points





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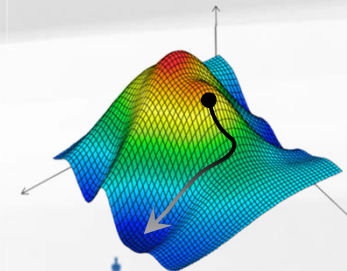
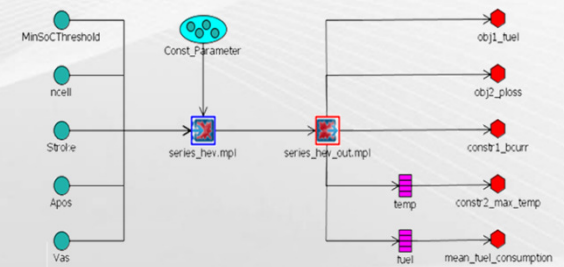
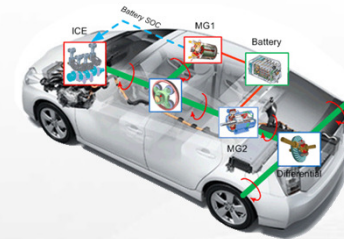
- Optimus offers the user a lot of flexibility by allowing the selection of many options for the algorithm
  - For inexperienced users good default values are predefined
  - Advanced users have the possibility to change the settings to improve the optimization behavior

# 5. Conclusion and Results

- In case of expensive simulations or a small amount of time, the user can afford less evaluations than suggested
  - The user can decrease the swarm size and therefore perform more iterations

# 5. Conclusion and Results

- ✓ Easy-to-use procedure to set up and handle a complex vehicle model in MapleSim
- ✓ Efficient automation possibility using OPTIMUS
- ✓ Application of advanced optimization algorithms without additional effort



# QUESTIONS ? ? ?

Thank you for your attention



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