

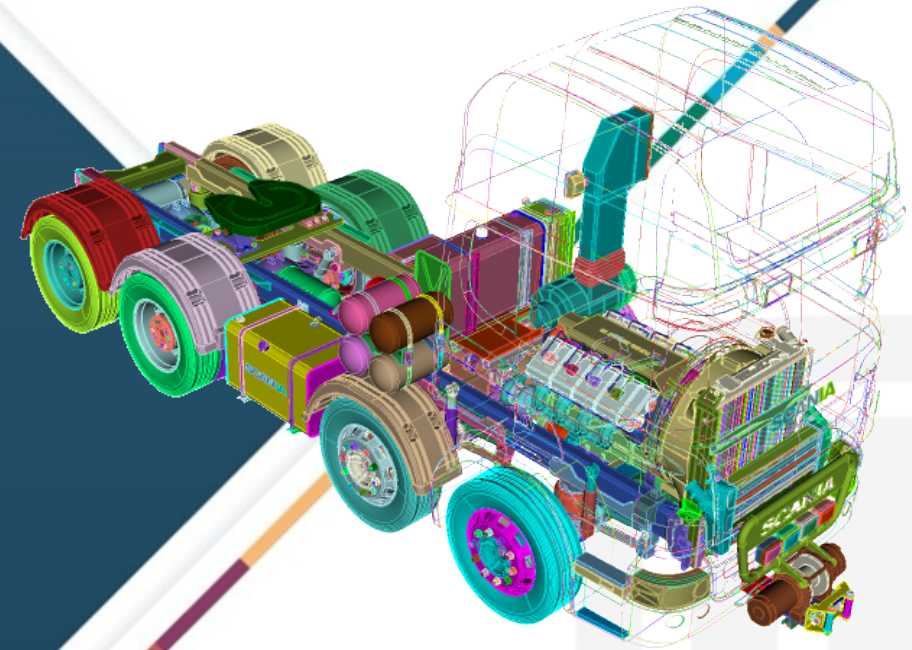


Exceptional service in the national interest

# Machine Learning Applications for Defeaturing and Model Preparation

Steven Owen, Corey Ernst,  
Armida Carbajal, Matthew Peterson

July 25, 2021



16<sup>th</sup> US Congress on  
Computational Mechanics

July 25-29, 2021  
A Virtual Event

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# Rapid CAD to Simulation with Machine Learning

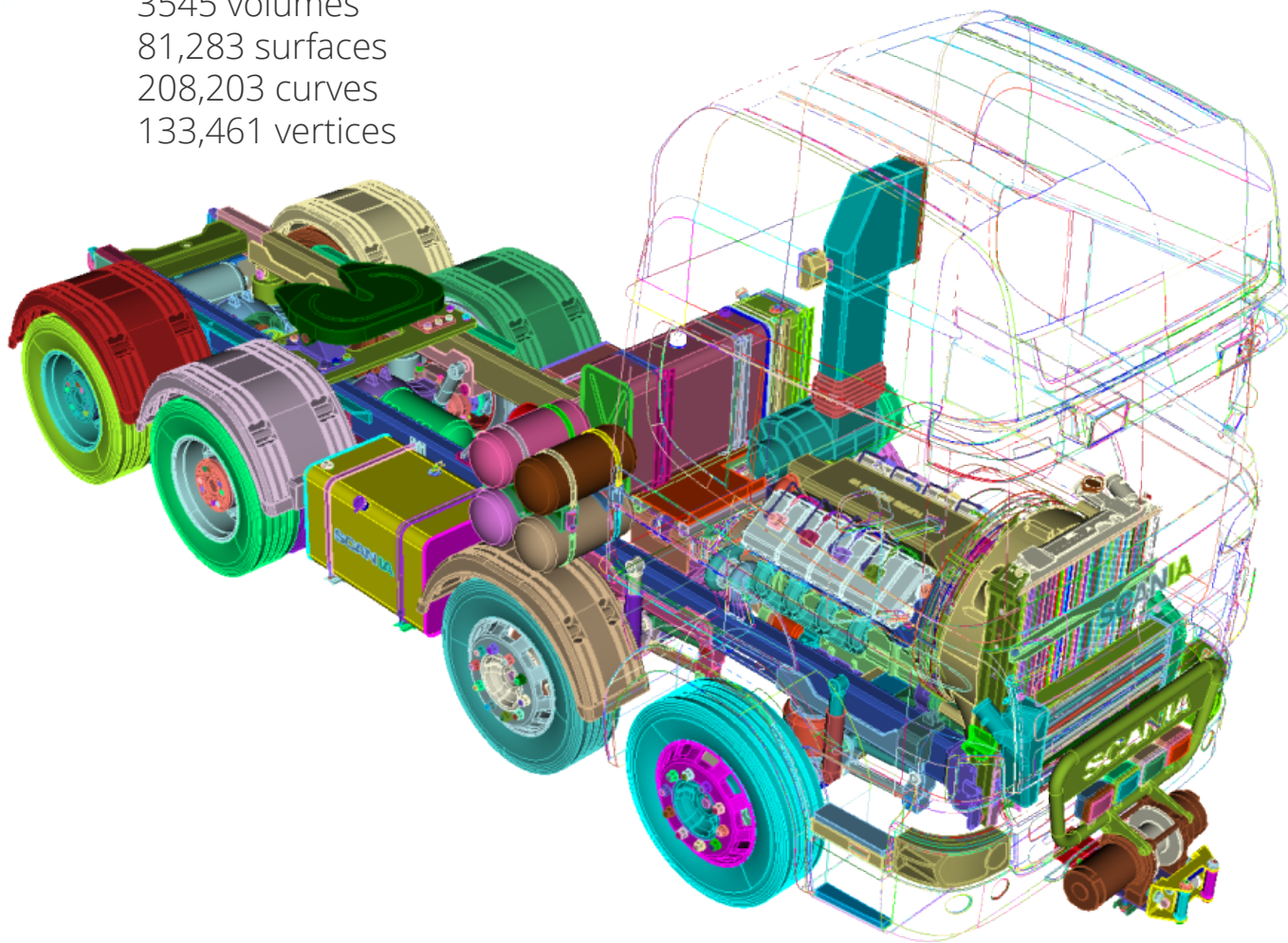
Commercial Truck CAD Model

3545 volumes

81,283 surfaces

208,203 curves

133,461 vertices



## Example workflow

- Break into subsystems for teaming and/or avoid large model size
- Remove parts that are irrelevant and not normally modeled for a specific physics
- Fix bad geometry
- Remove small features
- Remove gaps, overlaps, misalignments
- Tet mesh
- Identify and prepare specific categories of mechanisms (Bolts, Springs, Bearings, etc.)
- Compute preloads on bolts, springs, etc.
- Prepare input deck





# Rapid CAD to Simulation with Machine Learning

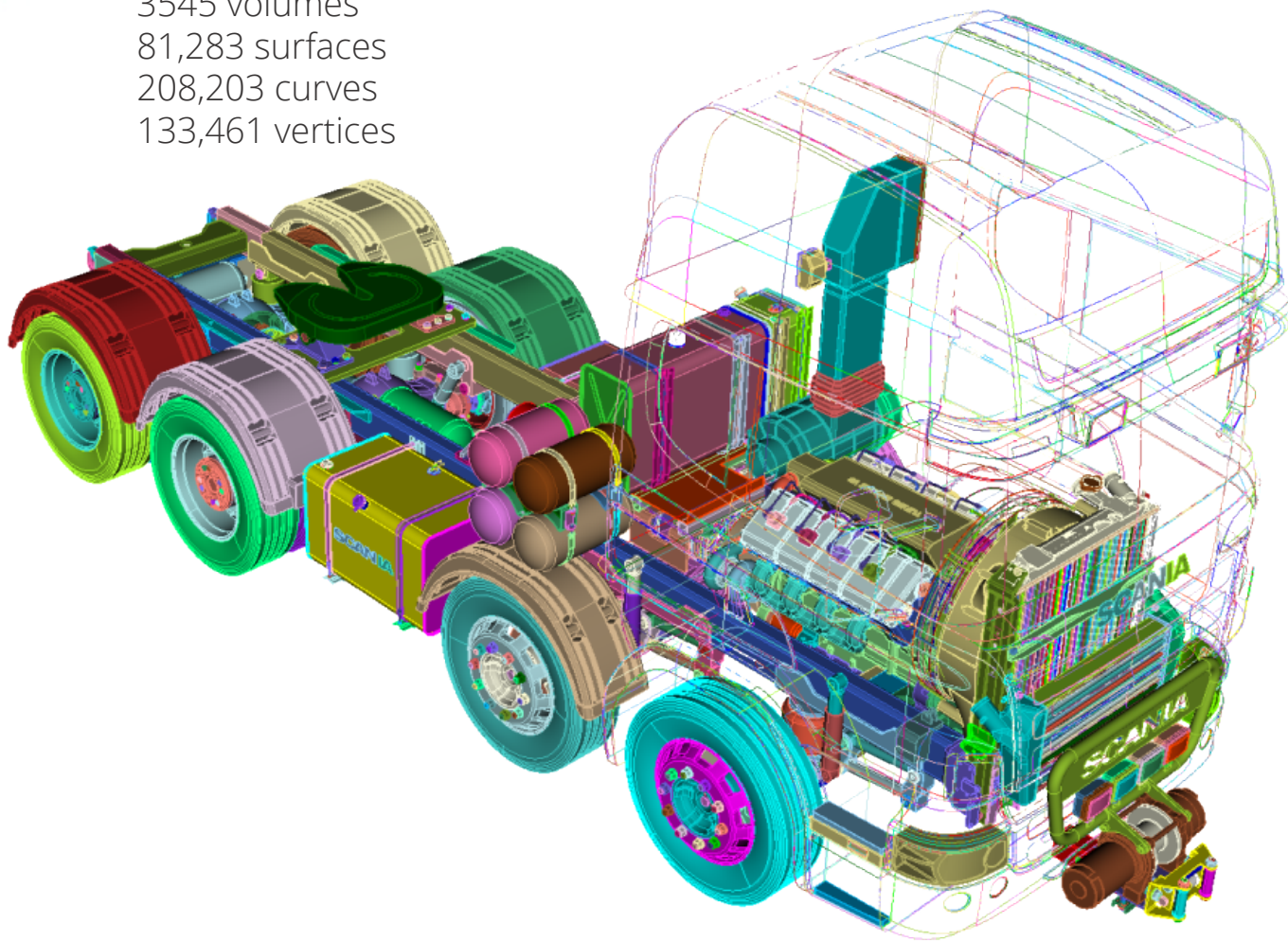
Commercial Truck CAD Model

3545 volumes

81,283 surfaces

208,203 curves

133,461 vertices



## Machine Learning for Analysis Preparation:

- Reduce/Eliminate most tedious tasks
- Guide user to next step
- Present user with smart options
- User validates ML-solutions
- Standardization between analysts



# Rapid CAD to Simulation with Machine Learning

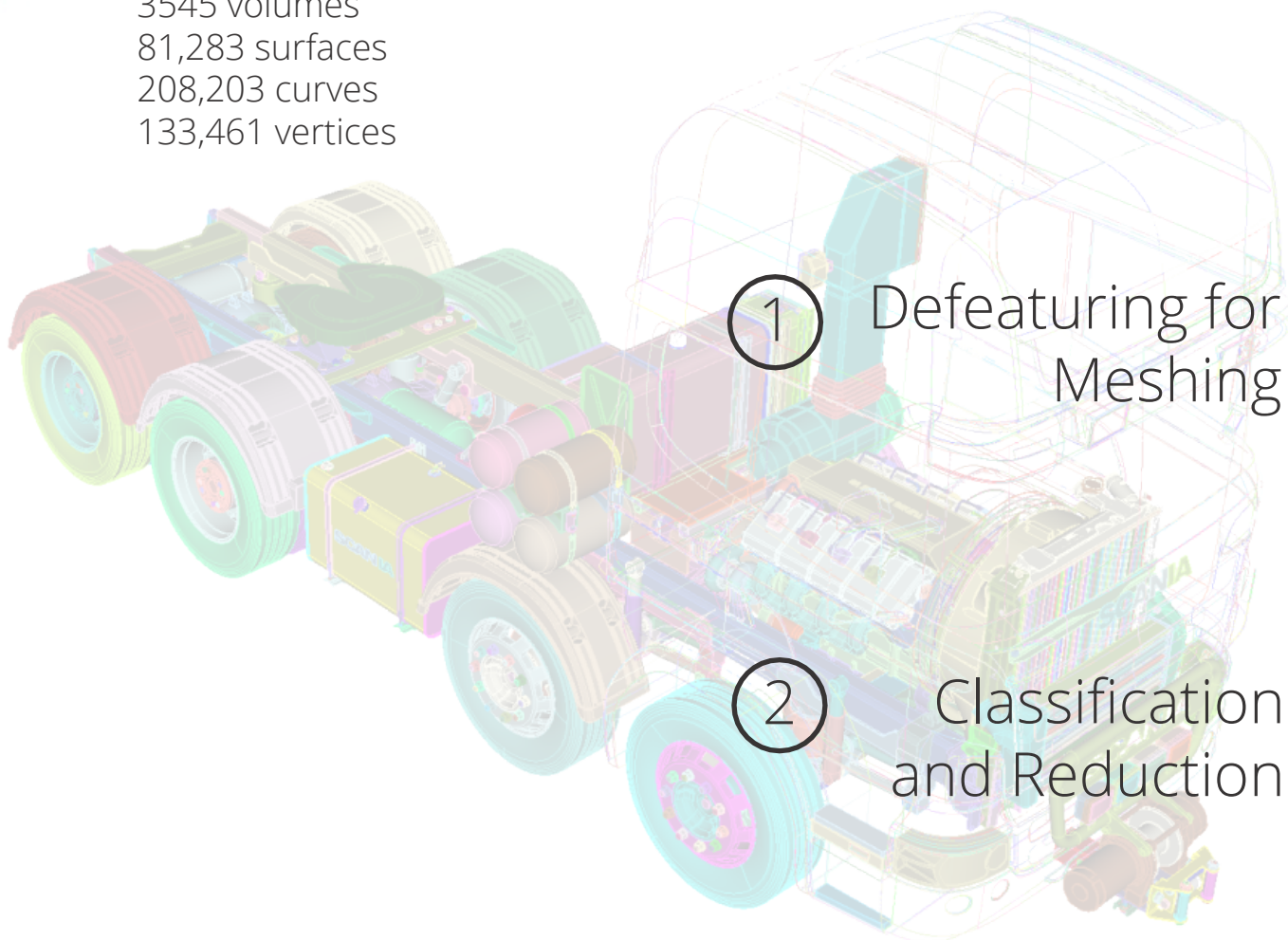
Commercial Truck CAD Model

3545 volumes

81,283 surfaces

208,203 curves

133,461 vertices



① Defeaturing for Meshing

② Classification and Reduction

## Example workflow

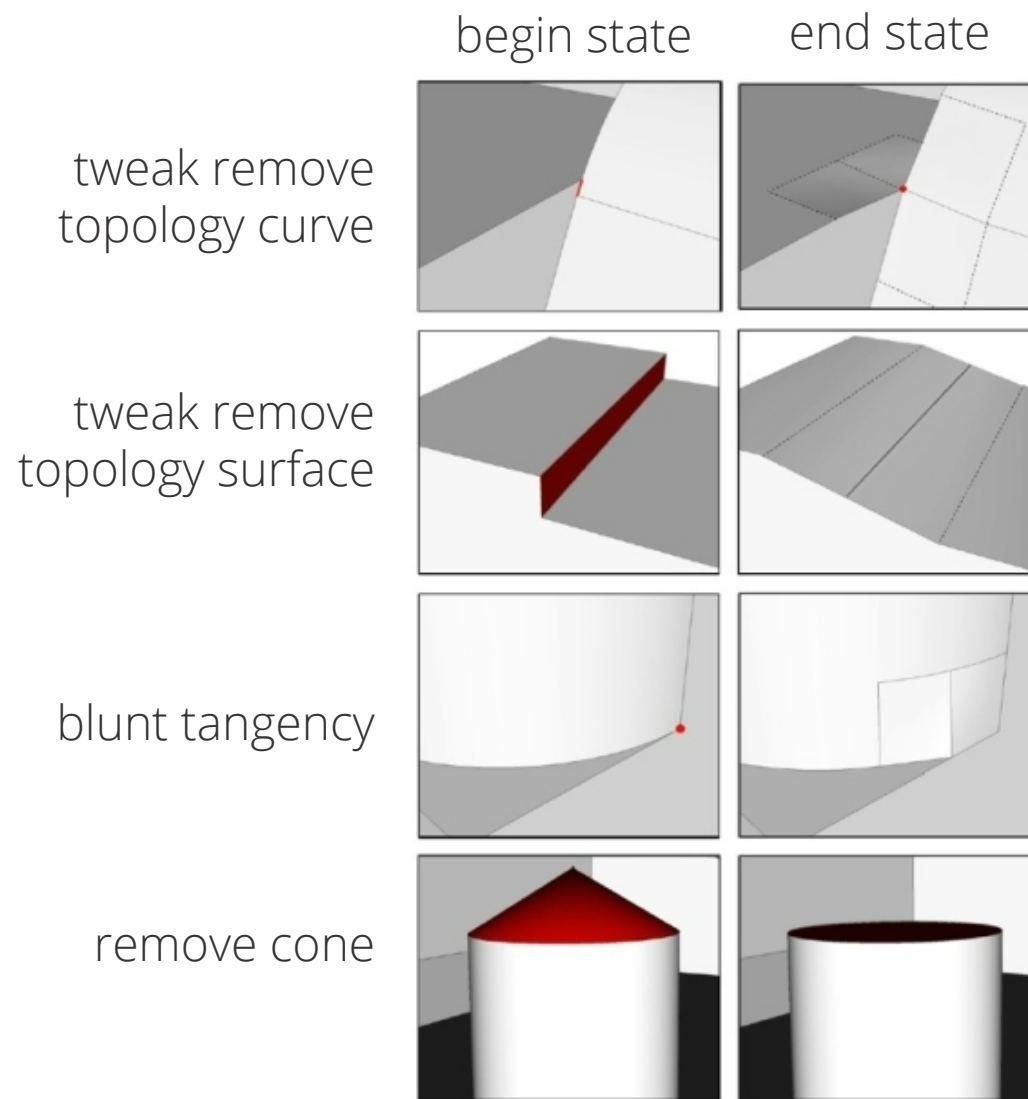
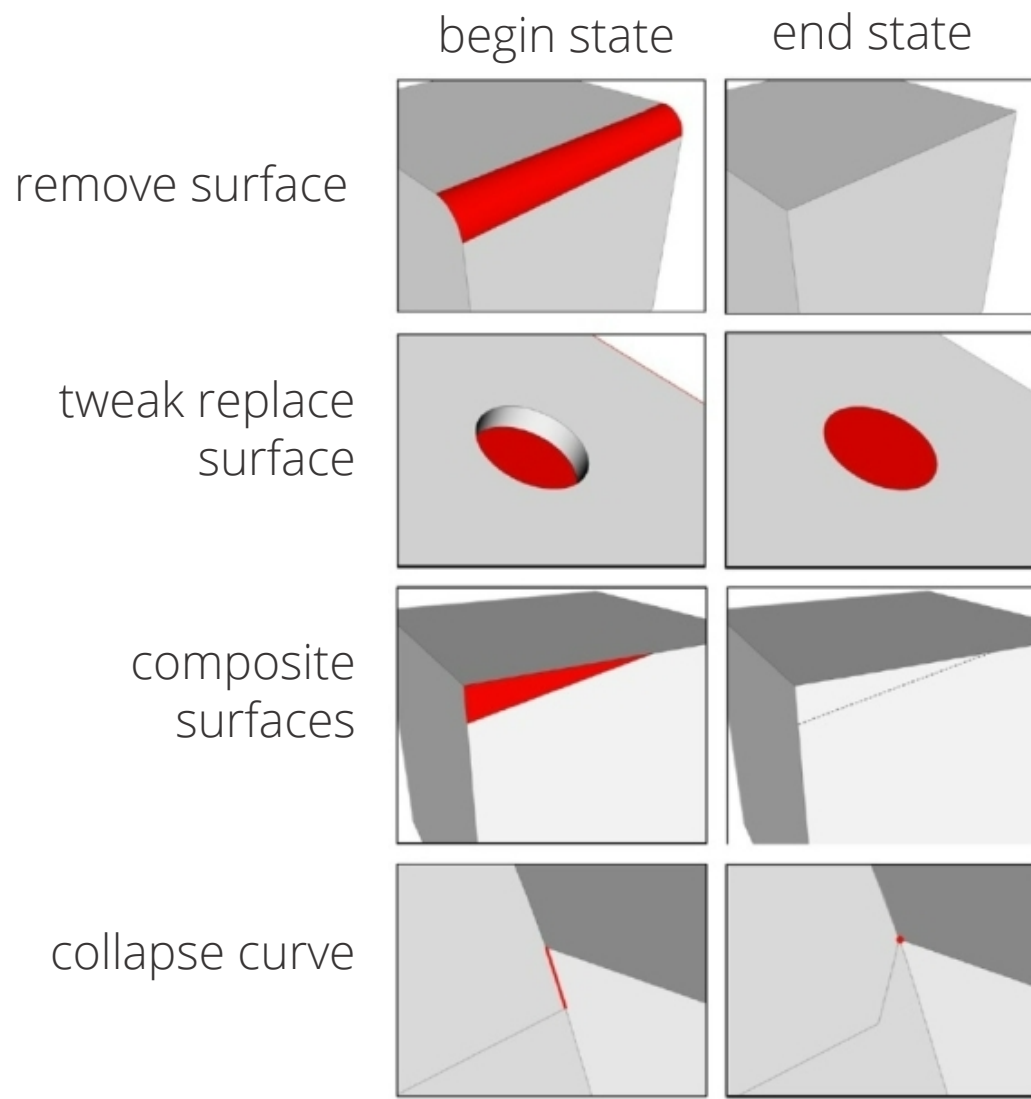
- Break into subsystems for teaming and/or avoid large model size
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- Remove small features
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- Identify and prepare specific categories of mechanisms (Bolts, Springs, Bearings, etc.)
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- ① Find the most problematic features in the CAD Model
  - For a given local topology
    - Small Curve
    - Small Surface
    - Sharp Angle at Vertex
  - Predict local mesh quality
    - Scaled Jacobian
    - In-Radius
    - Deviation
- ② Find solution that yields best mesh quality
  - For a given local geometry operation
    - Collapse Curve
    - Remove Surface
    - Blunt Tangency
    - Etc..
  - Predict local mesh quality after operation
    - Scaled Jacobian
    - In-Radius
    - Deviation

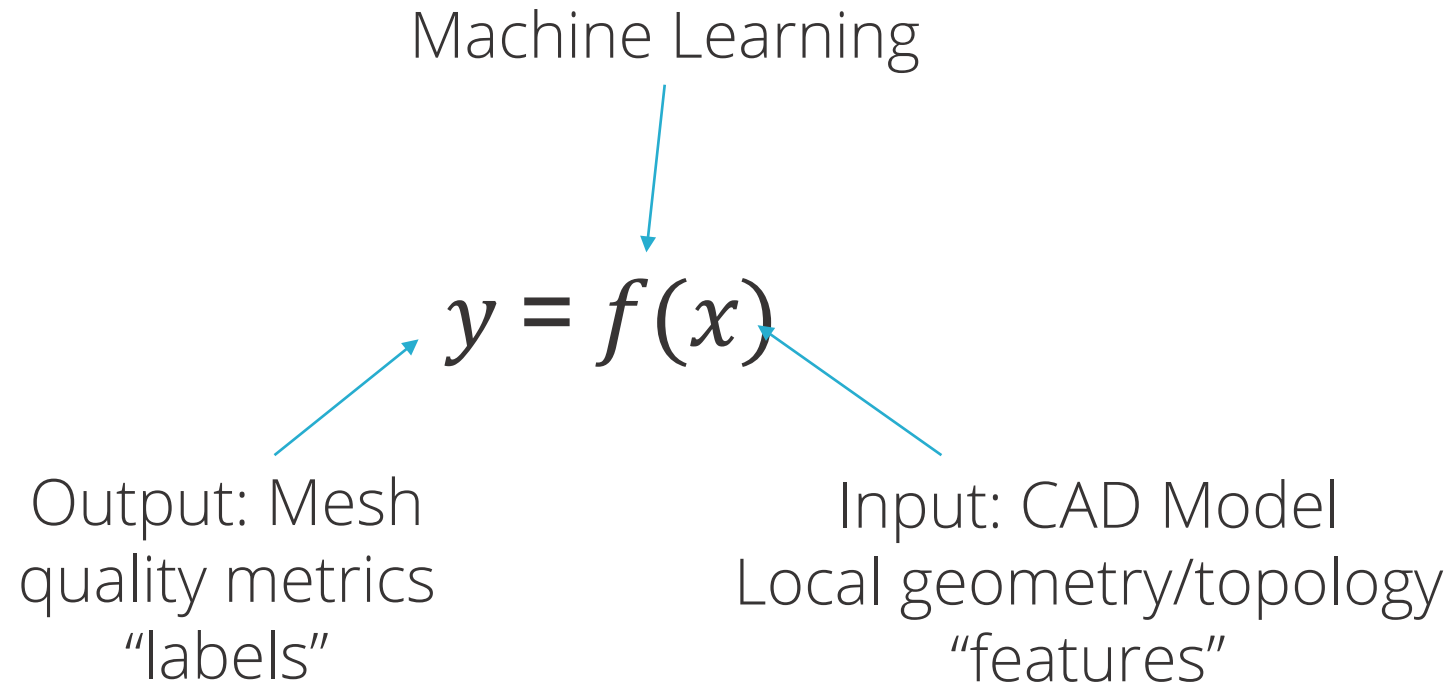


## Example defeaturing operations



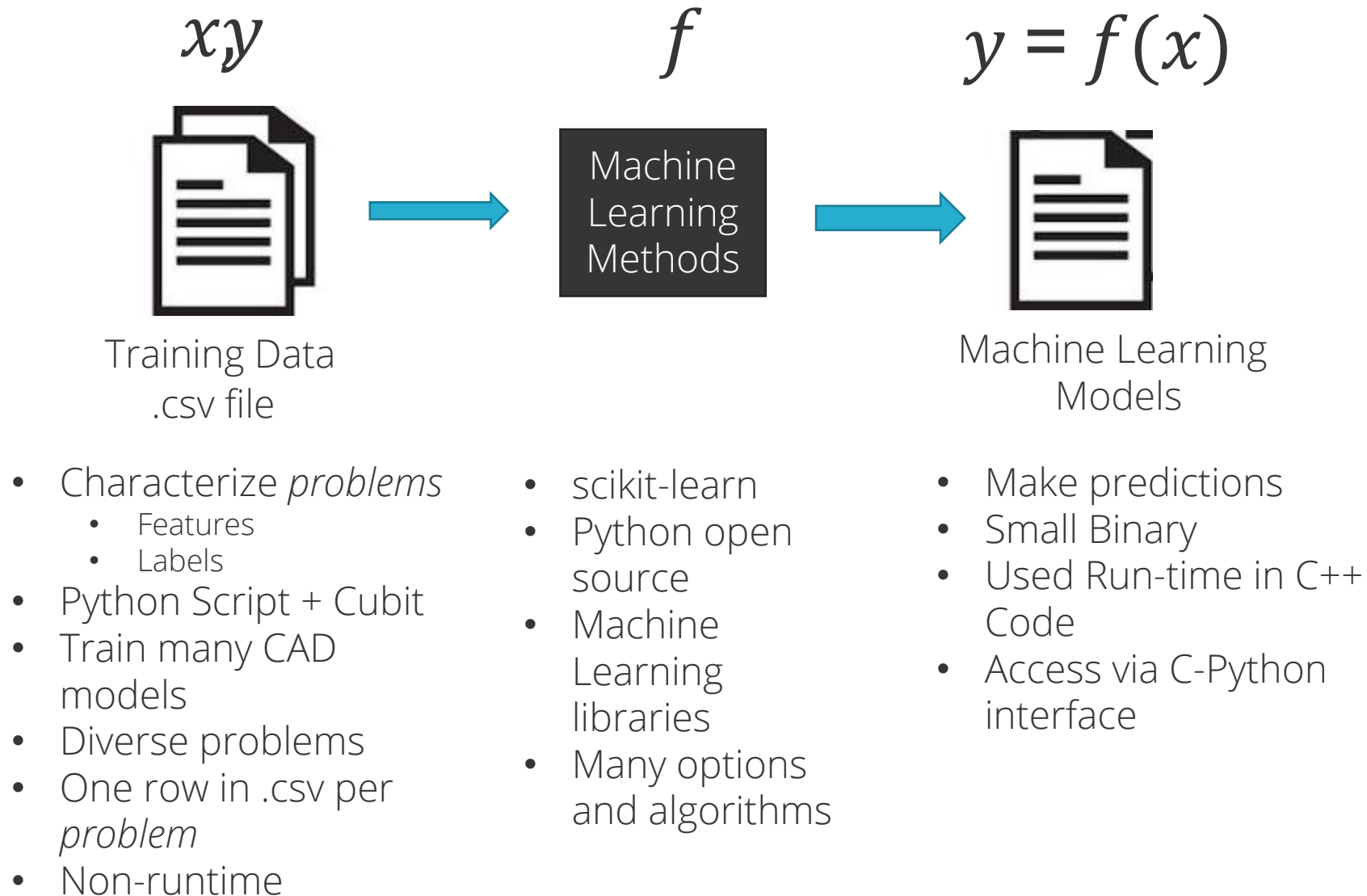


## Machine Learning for Defeaturing





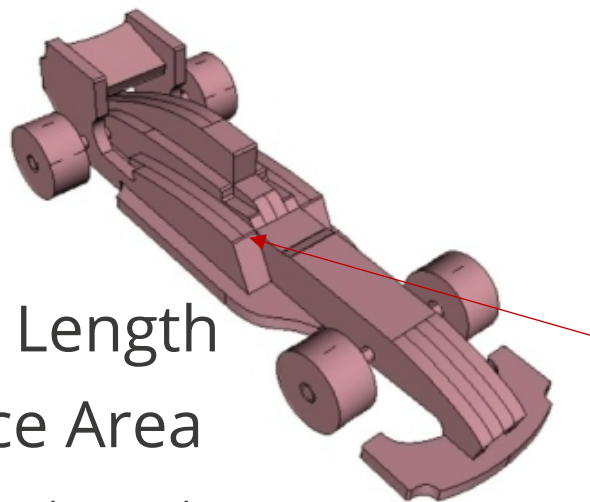
# Machine Learning for Defeaturing



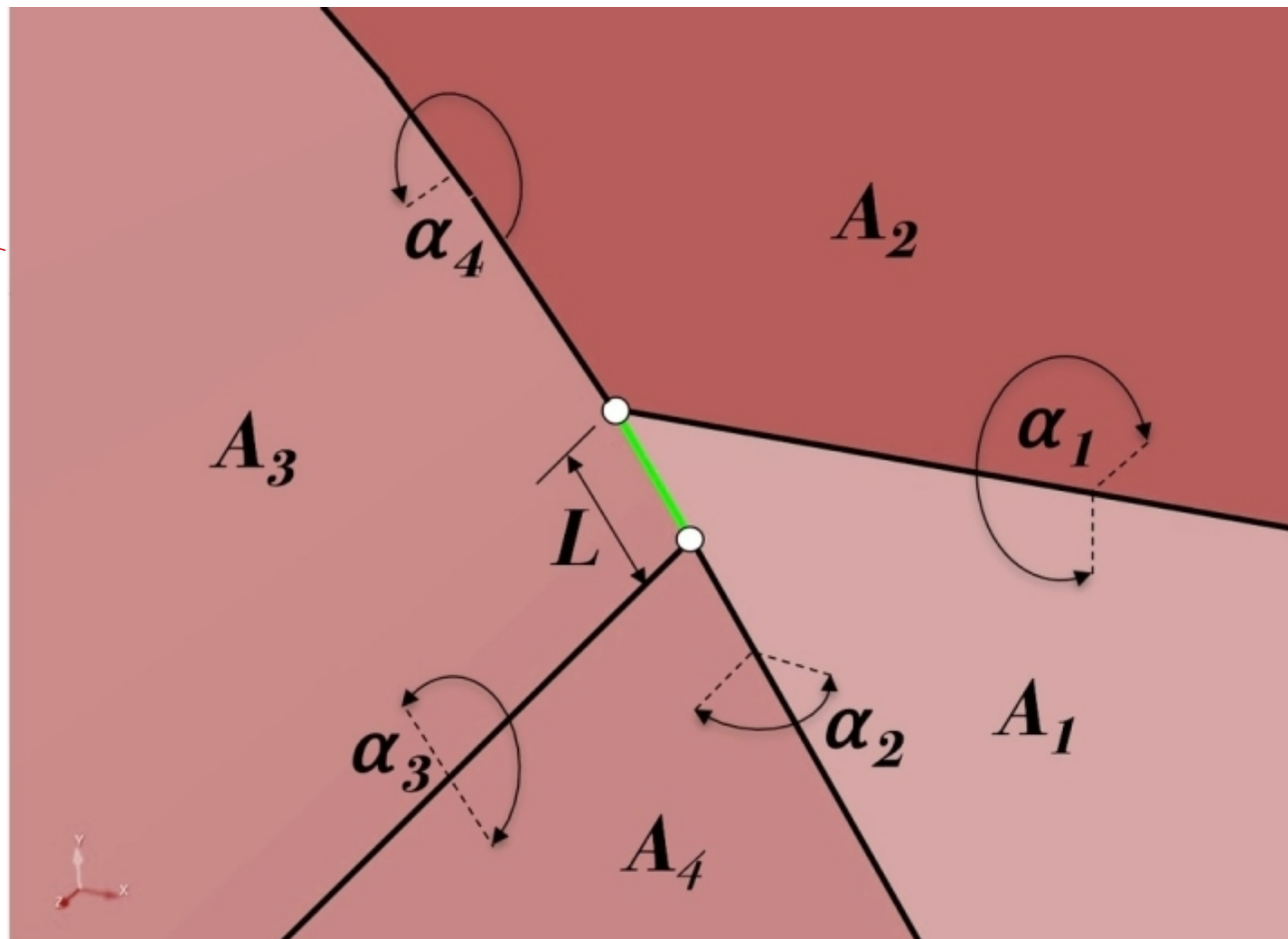




## Training Data Features



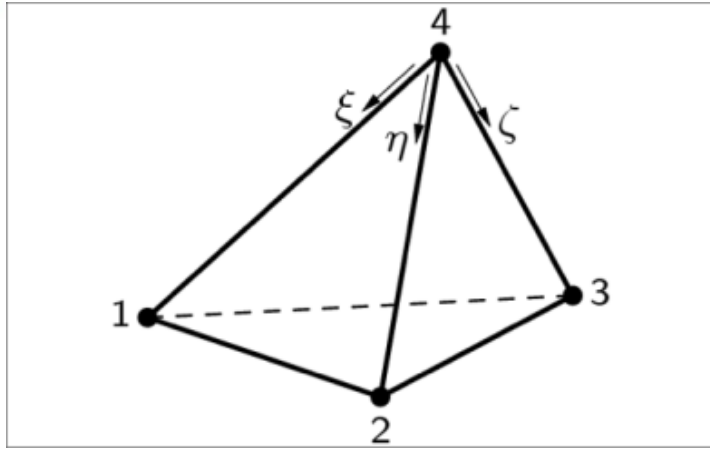
Curve Length  
Surface Area  
Angle at vertex  
Angle at curves  
Valence at vertex  
Number of loops  
Hydraulic Radius  
Etc...



Topology-Based Features

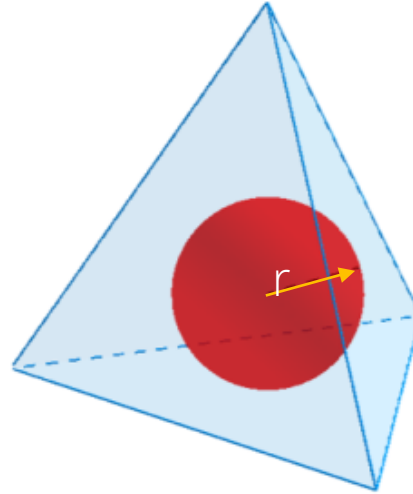


## Training Data Labels



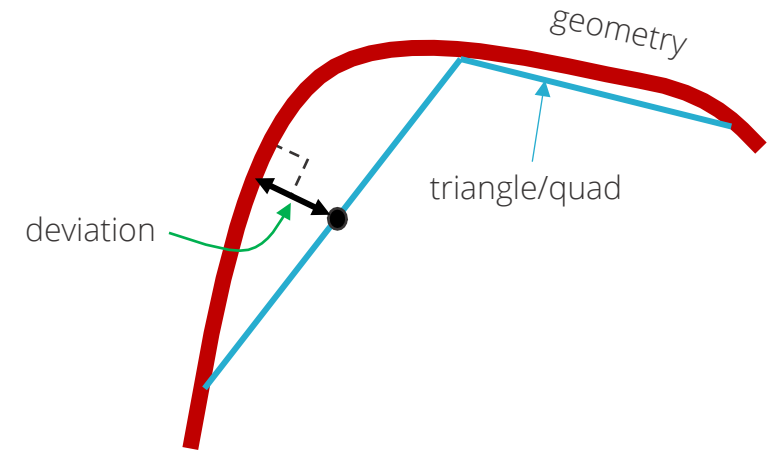
### Scaled Jacobian

- Function of angles at vertices
- Scaled -1.0 to 1.0
- Independent of mesh size



### In-Radius Ratio

- Ratio with equilateral tet with target edge length
- Scaled 0.0 to 1.0
- Sensitive to mesh size

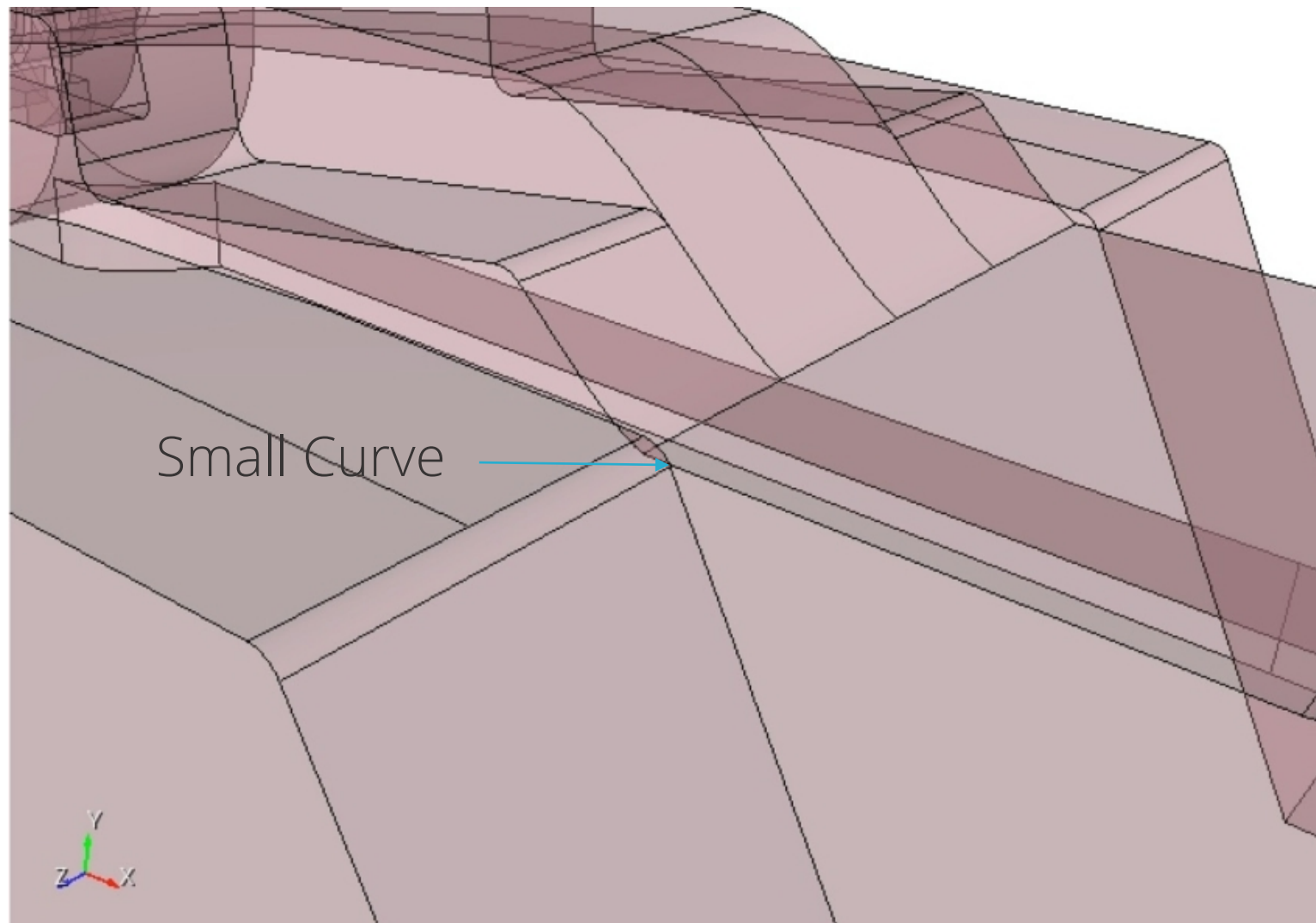


### Deviation

- Distance from triangle centroid to closest point on geometry
- Scaled by target mesh size
- Ensure mesh "matches" geometry

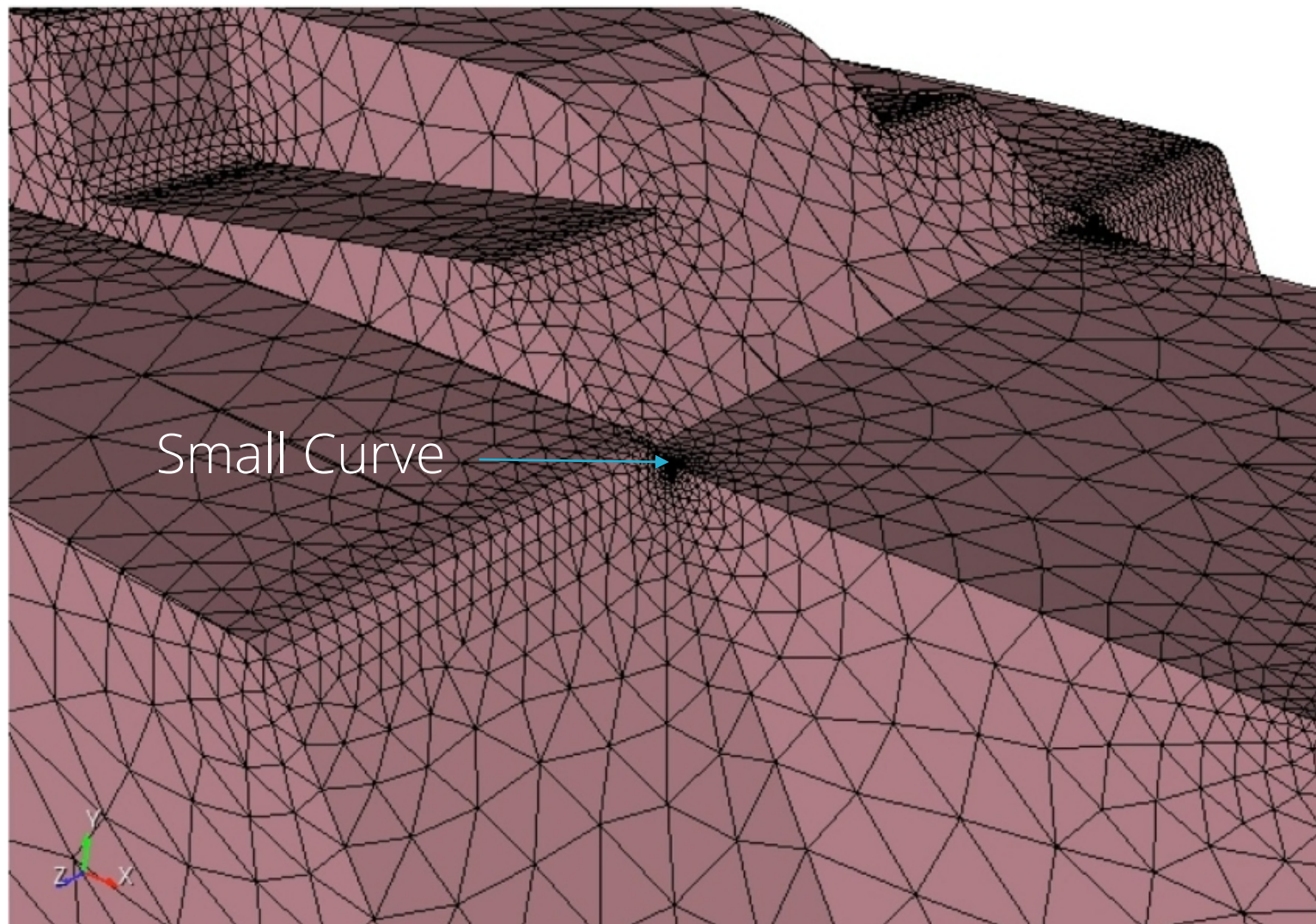


## Training Data Labels





## Training Data Labels

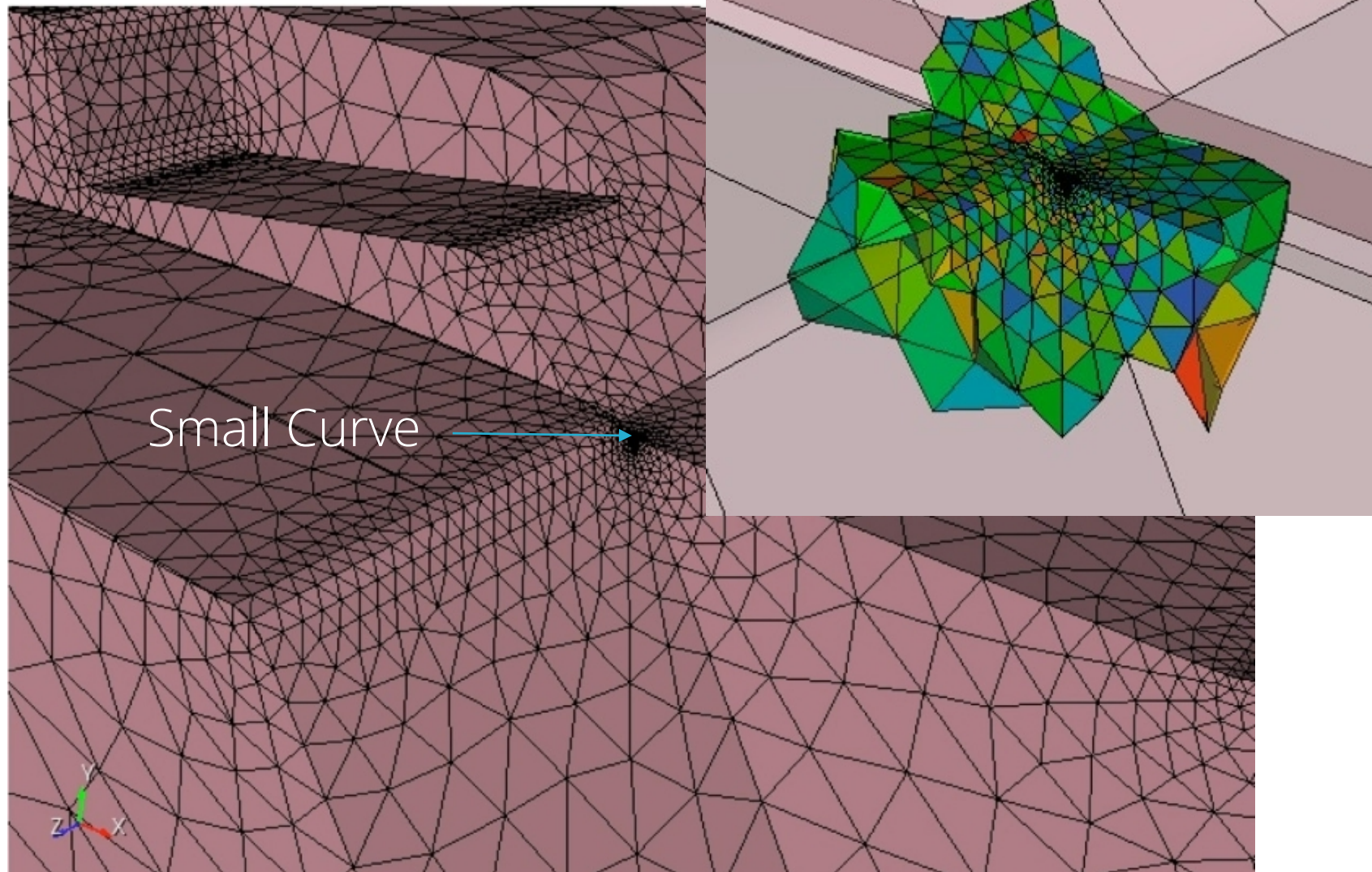






## Training Data Labels

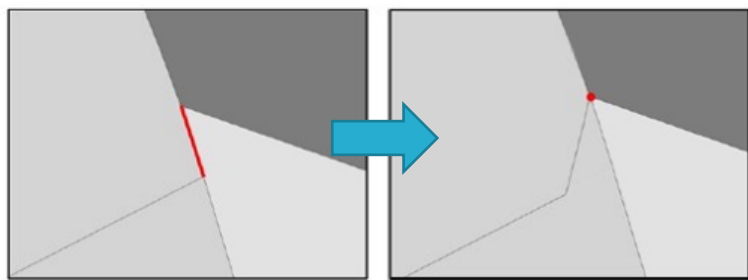
Local Tets at Geometric Curve





# Training Data Labels

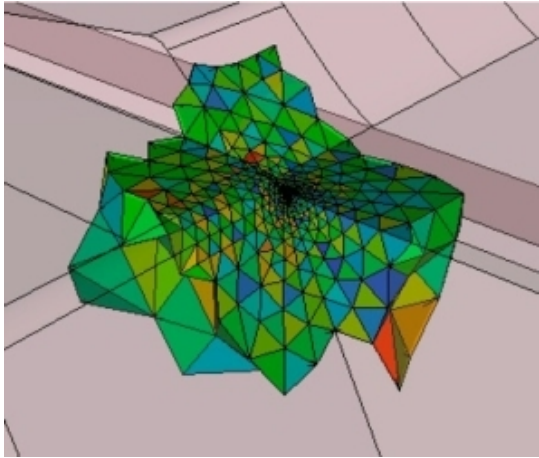
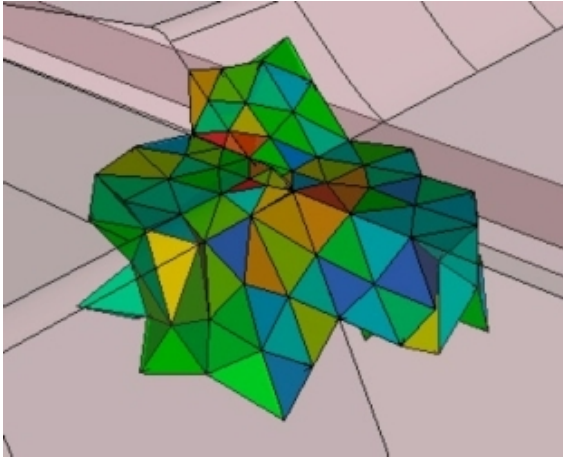
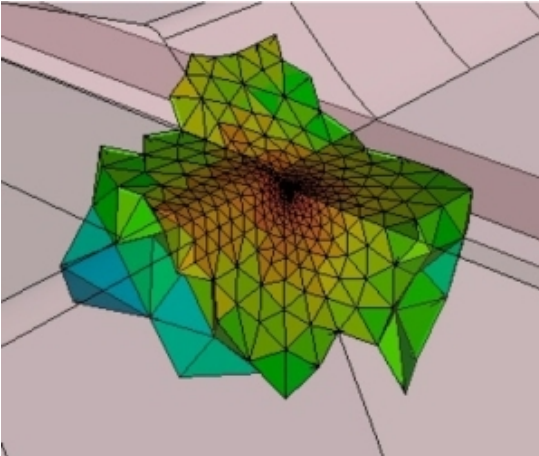
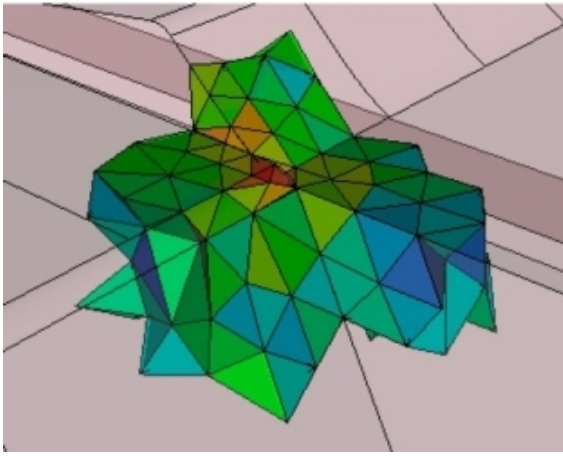
Collapse Curve



Before

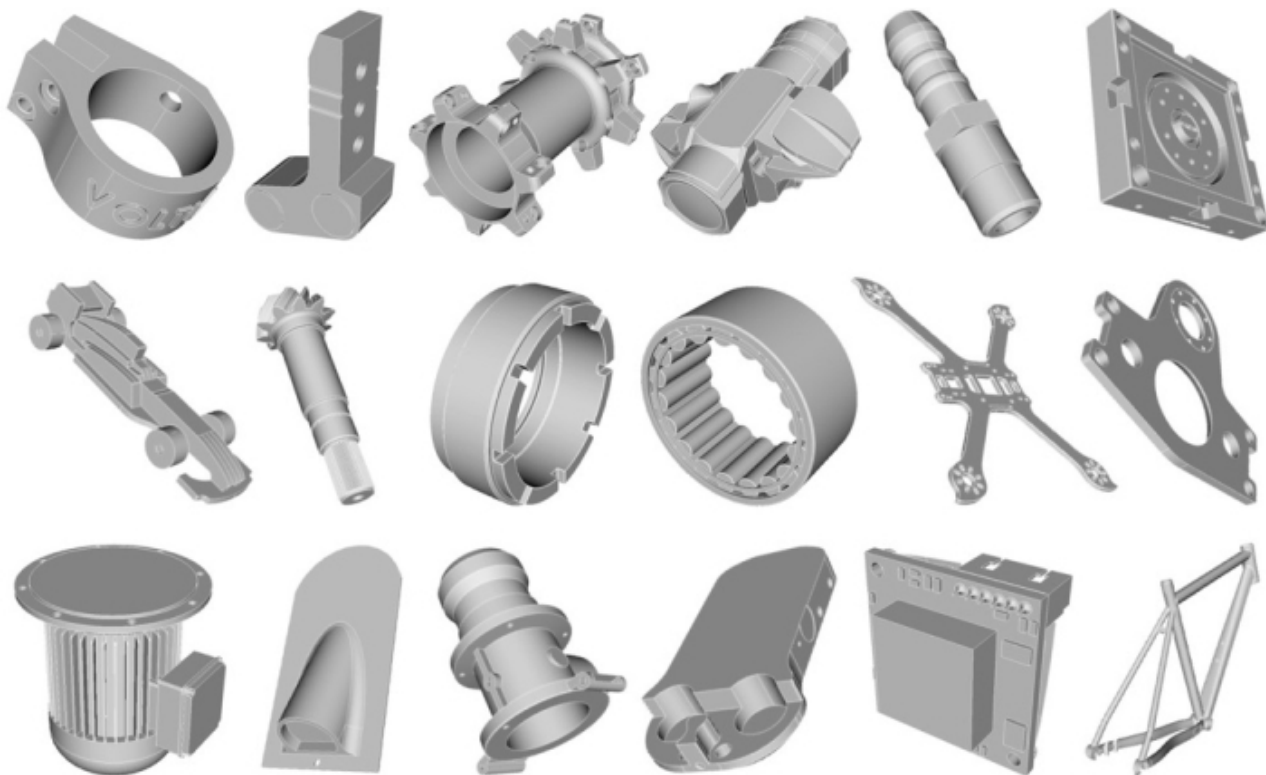
After

Collapse Curve

	Before	After
Scaled Jacobian	 Min SJ = 0.250	 Min SJ = 0.227
In-radius	 Min IR = 0.027	 Min IR = 0.267



# Machine Learning Study



94 single-part  
CAD models

	Num Obs.	Num Failed	Num Trained
vertex_no_op	1348	0	1348
curve_no_op	9842	0	9892
surface_no_op	5842	0	5842
remove_surface	17,624	10,026	7598
tweak_replace_surface	2569	1152	1417
composite_surfaces	43,551	5020	38,531
collapse_curve	13,830	2113	11,717
virtual_collapse_curve	14,955	14,743	212
remove_topology_curve	7056	5175	1881
remove_topology_surface	3890	3102	788
blunt_tangency	8059	3982	4077
remove_cone	232	20	212
<b>Totals</b>	<b>128,484</b>	<b>45,333</b>	<b>83,515</b>

Owen, Steven; Shead, Timothy M.; Martin, Shawn , "CAD Defeaturing Using Machine Learning",  
*28th International Meshing Roundtable (IMR)*, Buffalo, New York, USA, October 14-17, 2019

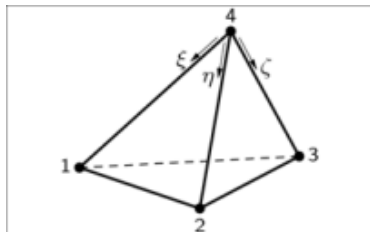
Patent Pending Application 17/016, 542 September 10, 2020



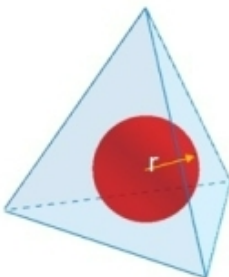


# Machine Learning Study – EDT (Ensembles of Decision Trees)

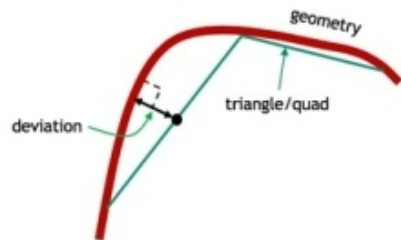
Scaled Jacobian



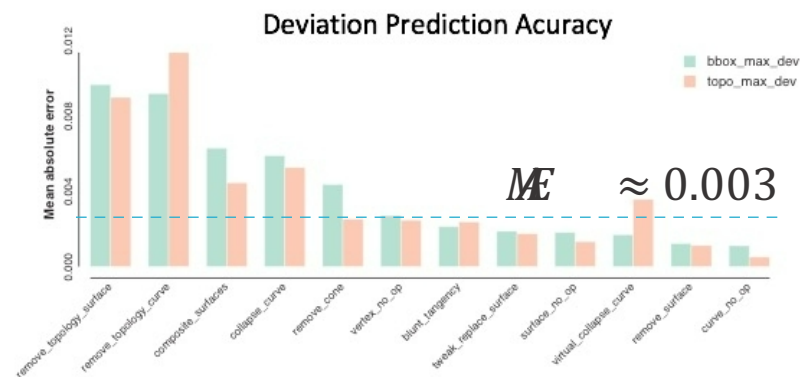
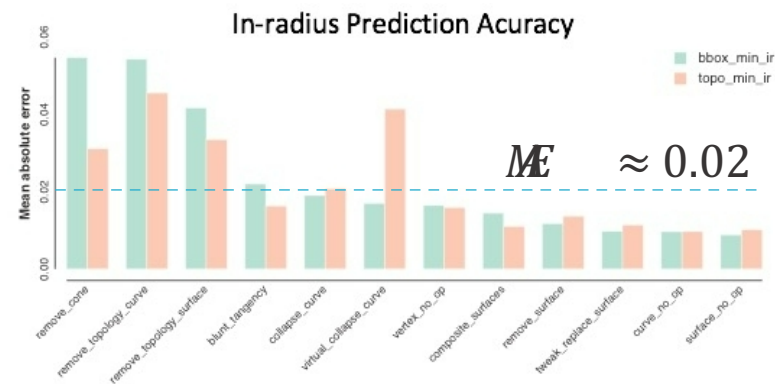
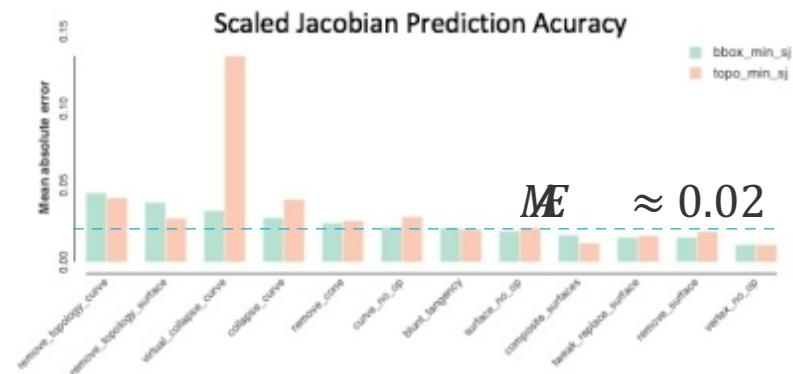
Scaled In-radius



Scaled Deviation



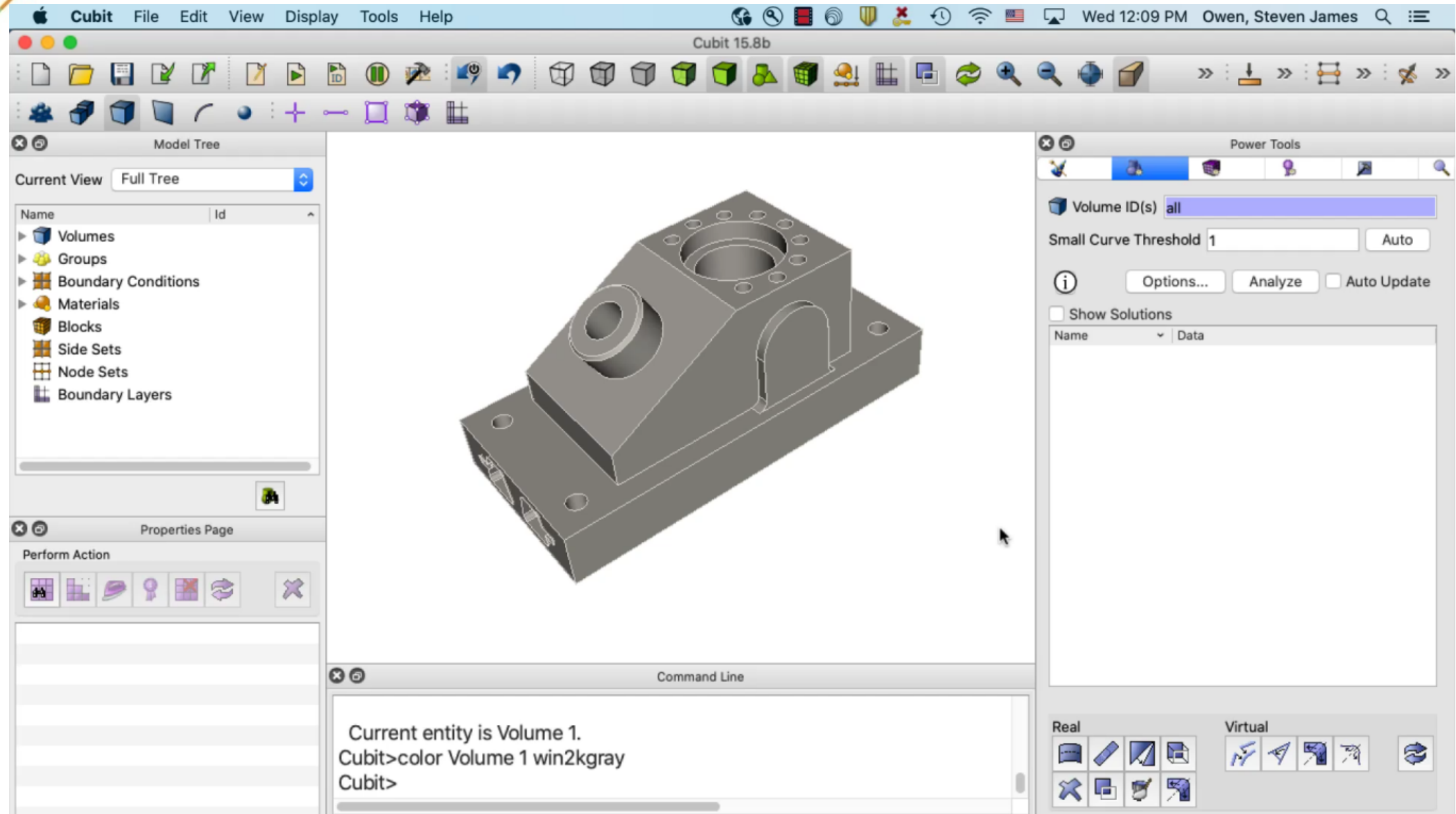
Mean Absolute Error (MAE)

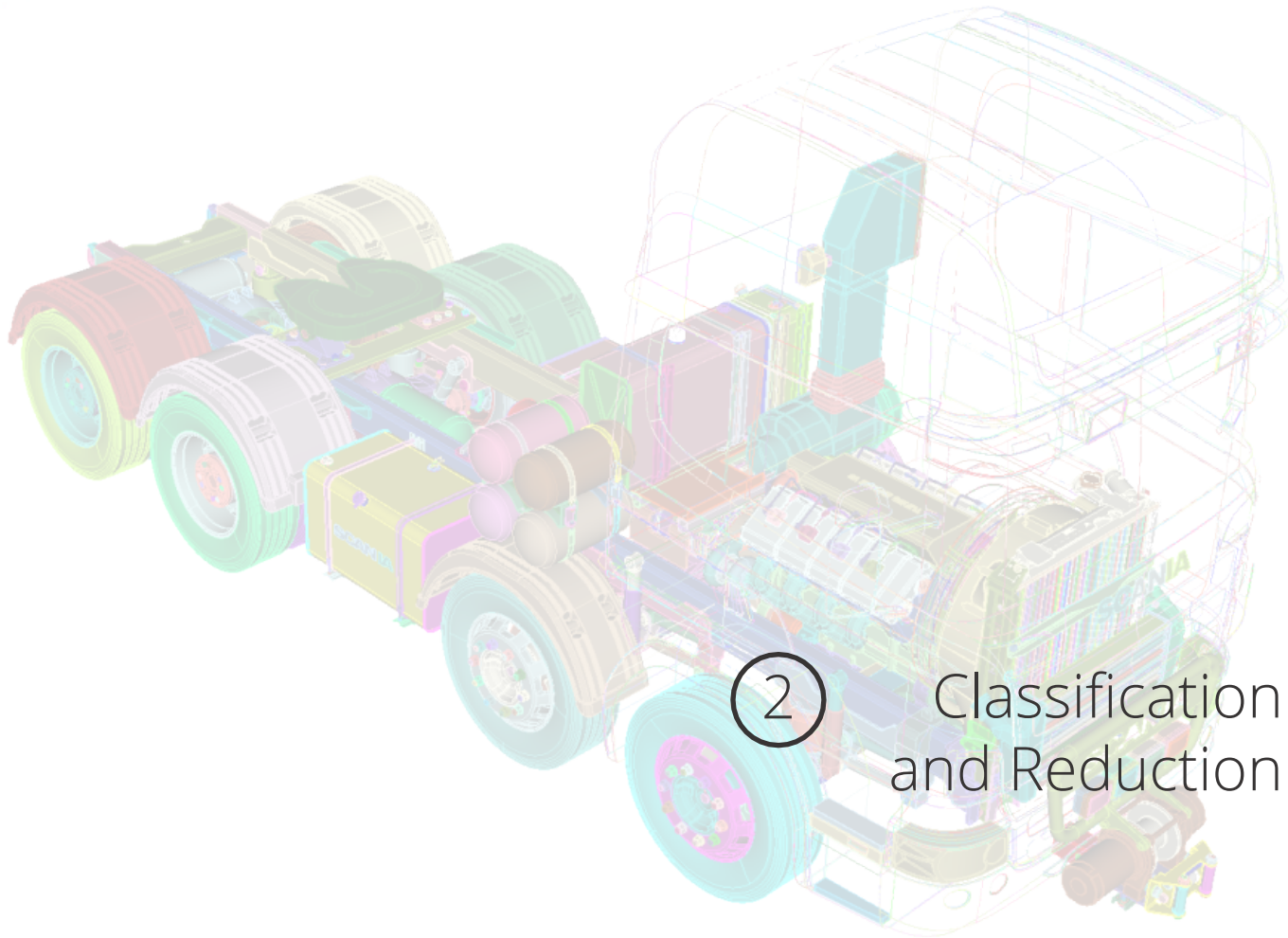






# Defeaturing for Meshing Example





2 Classification and Reduction

## Example workflow

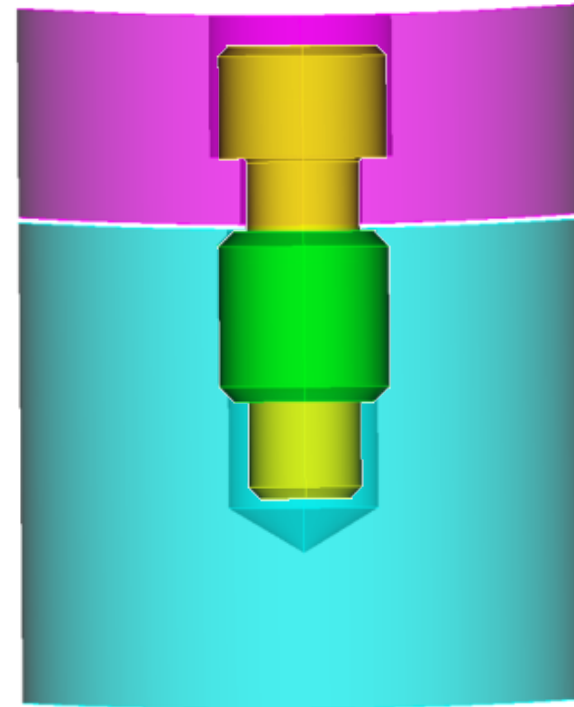
- Break into subsystems for teaming and/or avoid large model size
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# Mechanism Part Classification

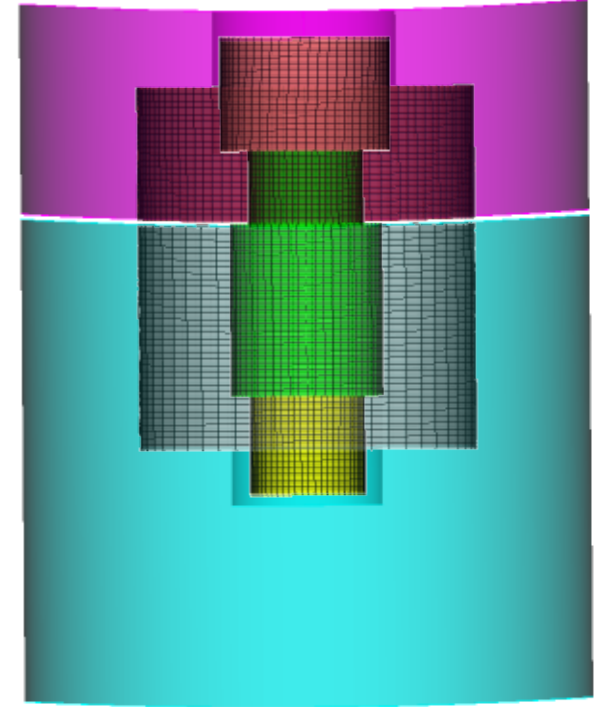
Classify Common Mechanisms



Rapidly Prepare for Analysis



Bolt in CAD assembly before preparation

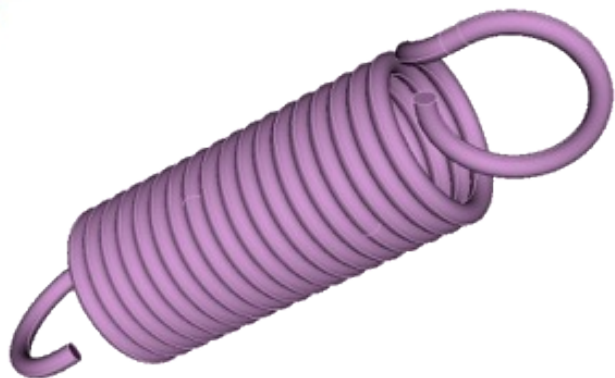


Bolt ready for analysis



## Mechanism Part Classification - Features

### Features of a CAD Volume



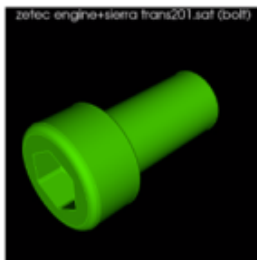
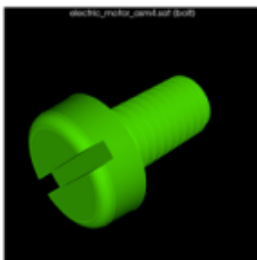
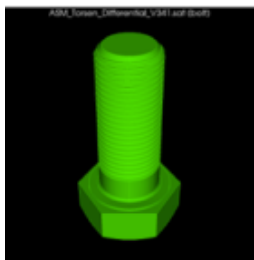
- Genus
- Min aspect ratio
- Max aspect ratio
- Tight Bounding Box Volume / Volume of Part
- Principal Moments
- Area of Surfaces / Volume of Part
- Ratio of Cylindrical Surface Area
- Ratio Planar Surface Area
- Ratio High Curvature Surfaces
- Ratio of exterior curve angles 90 degrees
- Ratio of exterior curve angles 180 degrees
- Ratio of exterior curve angles 270 degrees
- Etc...



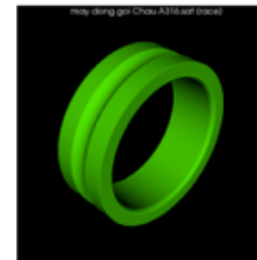
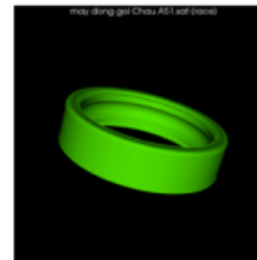
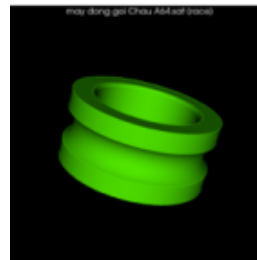


# Mechanism Part Classification - Labels

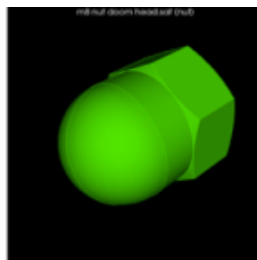
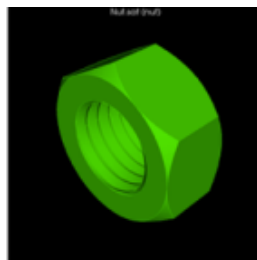
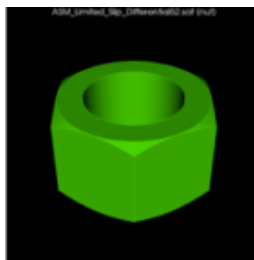
Bolt



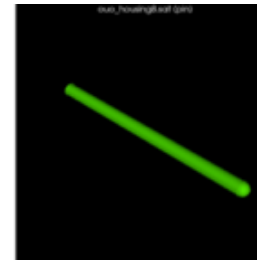
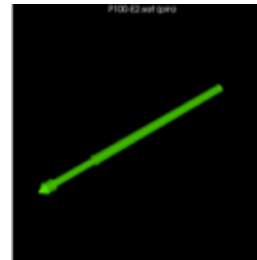
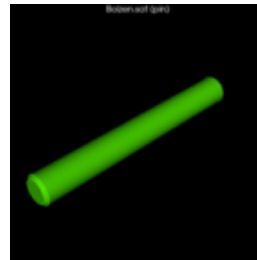
Race



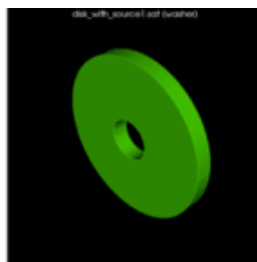
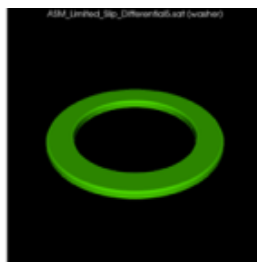
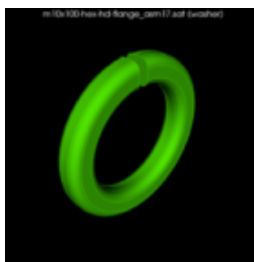
Nut



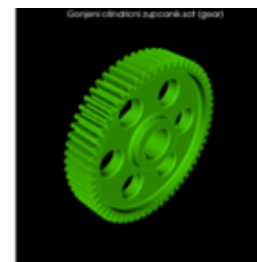
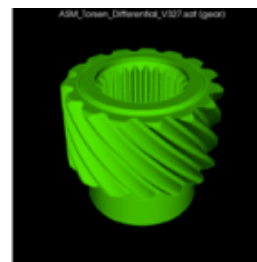
Pin



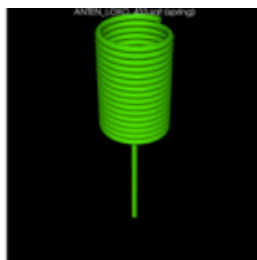
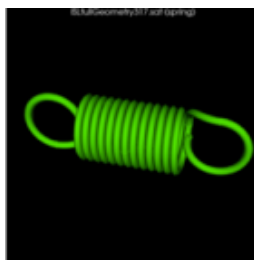
Washer



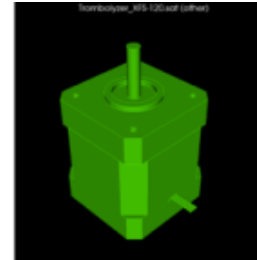
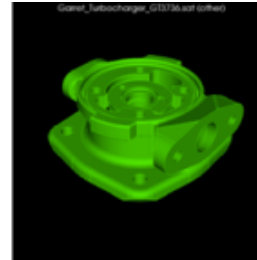
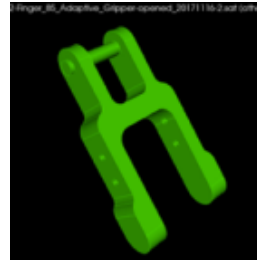
Gear



Spring



Other





## Mechanism Part Classification - Accuracy



label	precision	recall	support
<hr/>			
is_bolt	98.1%	98.3%	423
is_nut	95.5%	82.4%	44
is_washer	95.7%	97.8%	92
is_spring	100.0%	100.0%	34
is_ball	100.0%	100.0%	262
is_race	95.2%	100.0%	83
is_pin	97.8%	98.3%	180
is_gear	99.0%	89.9%	99
is_other	97.3%	98.1%	1033
Total			2250

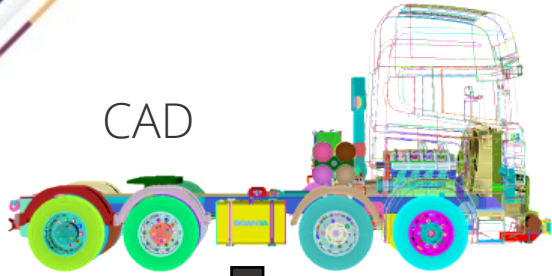
Ensembles of Decisions Trees  
with 5X2 cross validation

Models Source: -OUO Models from Sandia Analysts  
-GrabCAD Open Source Repository

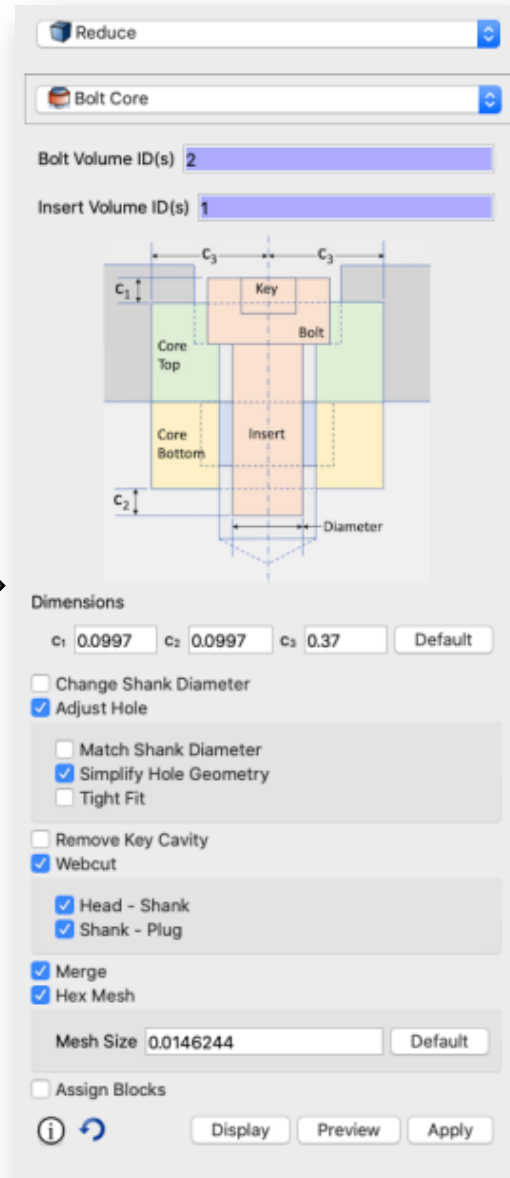
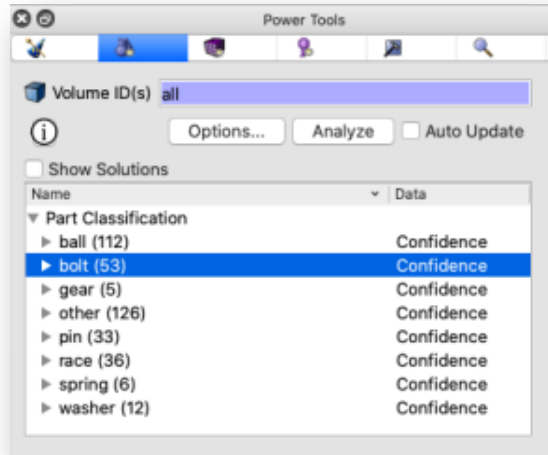


# Reduce for Simulation (Fasteners)

CAD



classify

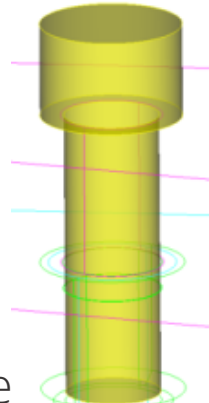


reduce

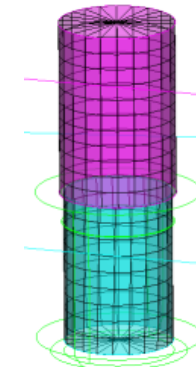


simulation-ready

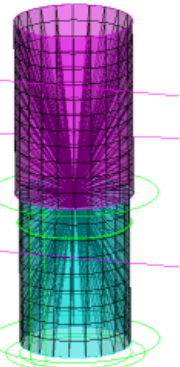
simple reduce



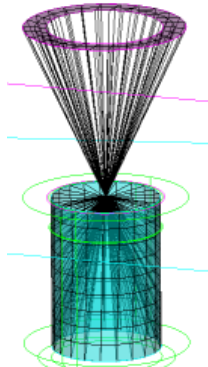
spider wagon wheel



spider j2g



spider countersink

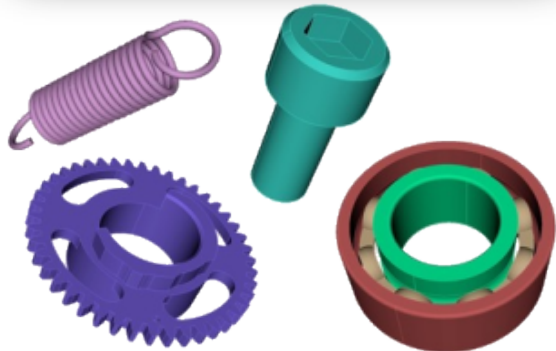


tweak hole diameter to shaft; webcut head, shank, plug

Hex meshed

Include insert and tweak to hole, shaft

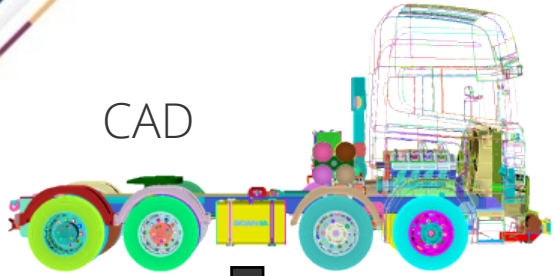
core from surrounding volumes



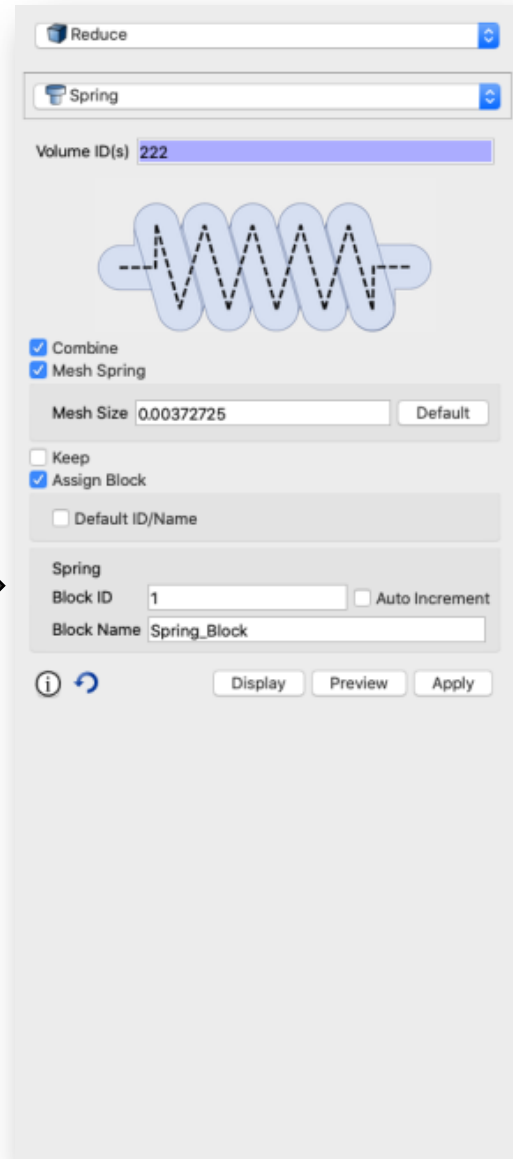
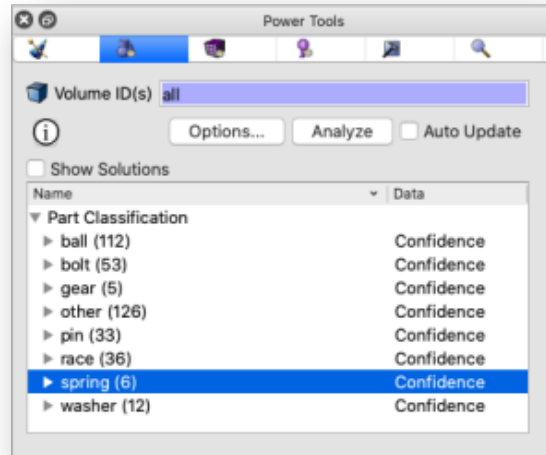


# Reduce for Simulation (Springs)

CAD



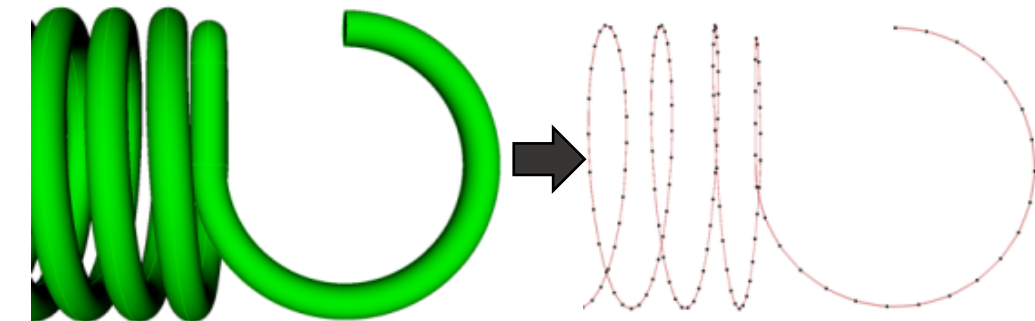
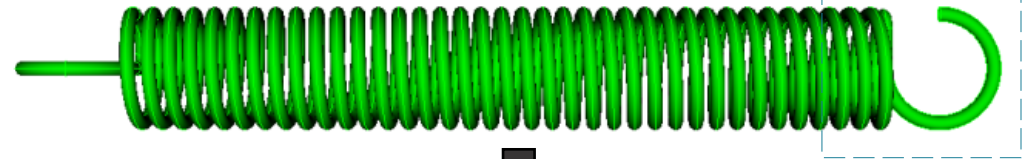
classify



reduce



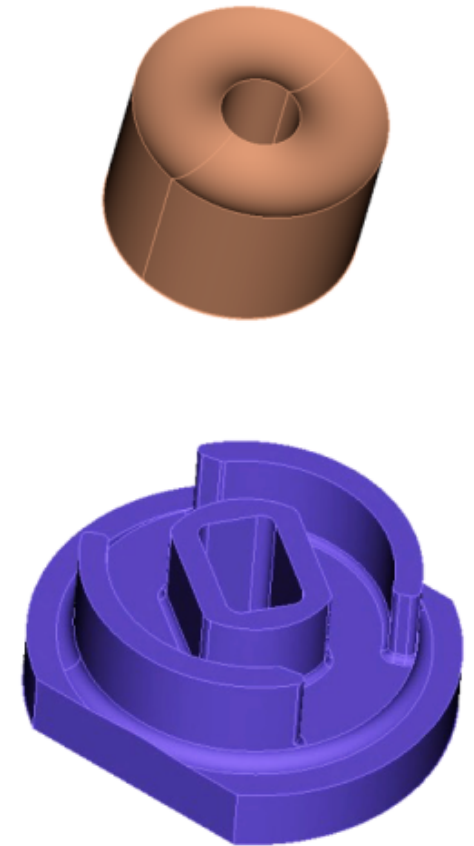
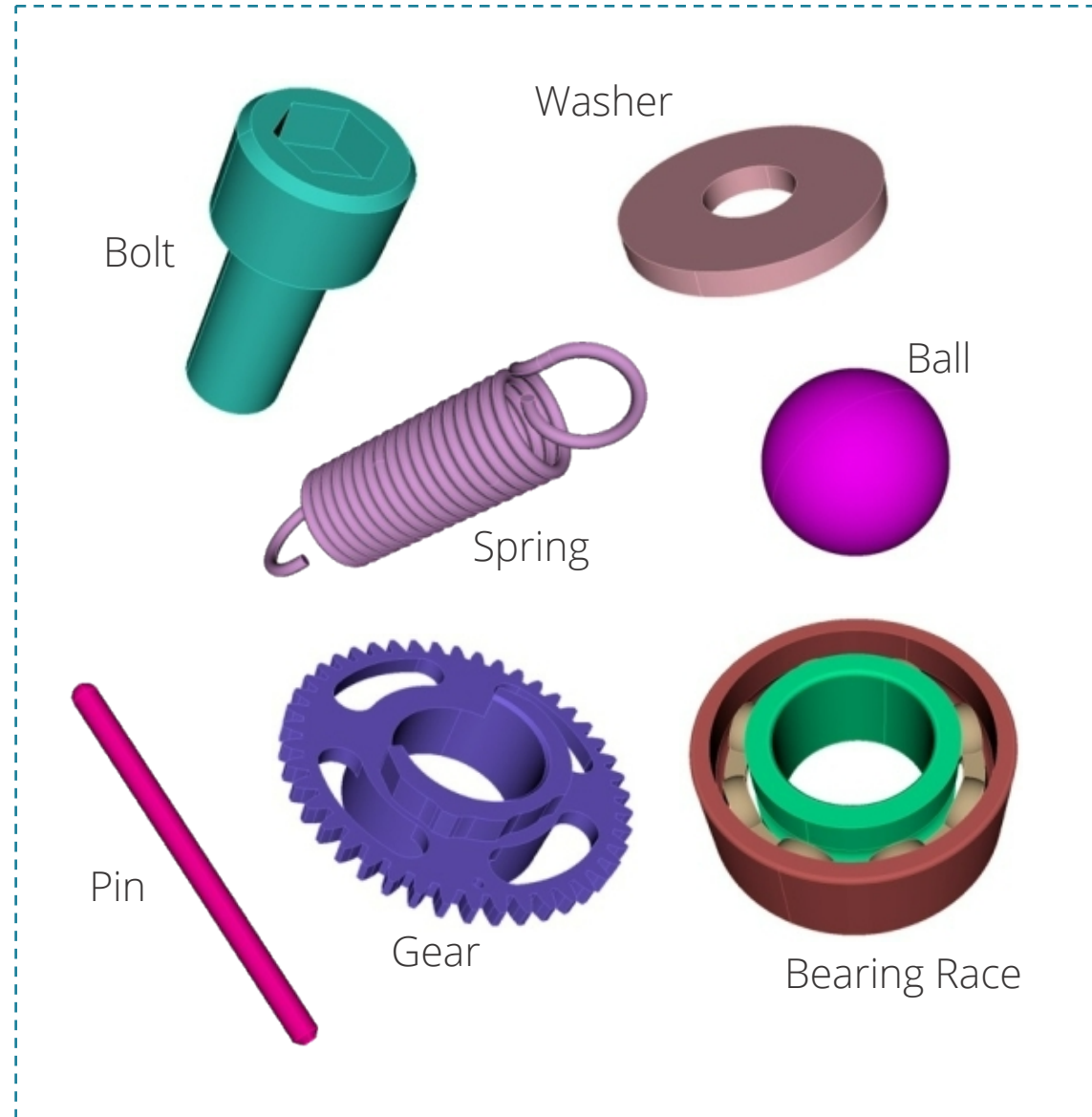
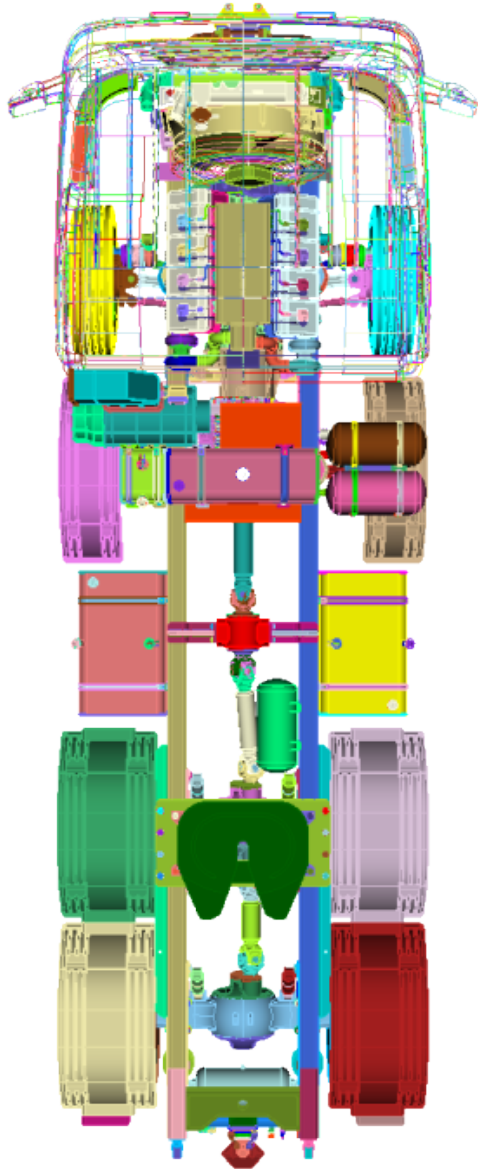
simulation-ready







## In-situ user driven training



User-defined parts



# In-situ user driven training



All available training data


Apply learning algorithm

Finished Model  
with N categories

Make many predictions

“Normal” developer driven training

Existing Model  
with N categories

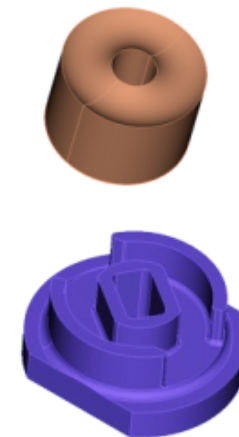
New category examples


Extend model and  
apply learning algorithm

New Model  
with N+1 categories

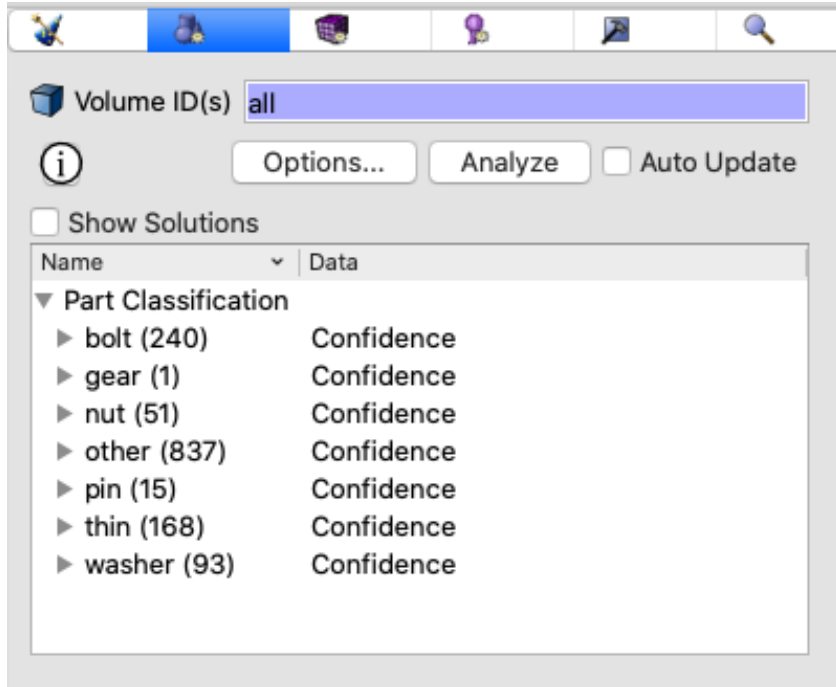
Make predictions

Online user driven training

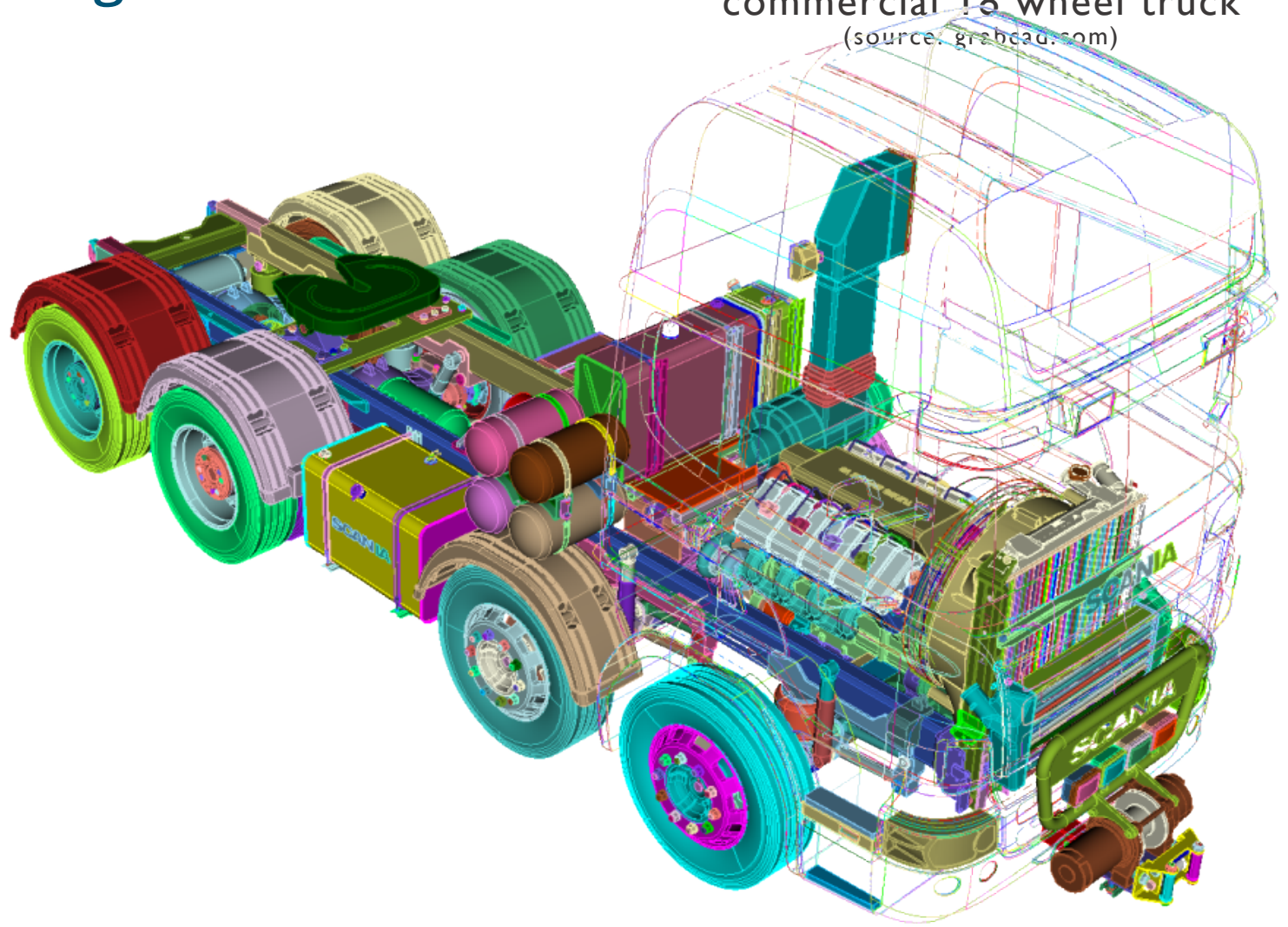


# In-situ user driven training

CAD assembly of  
commercial 18 wheel truck  
(source: grabcad.com)



1048 parts classified into  
7 categories

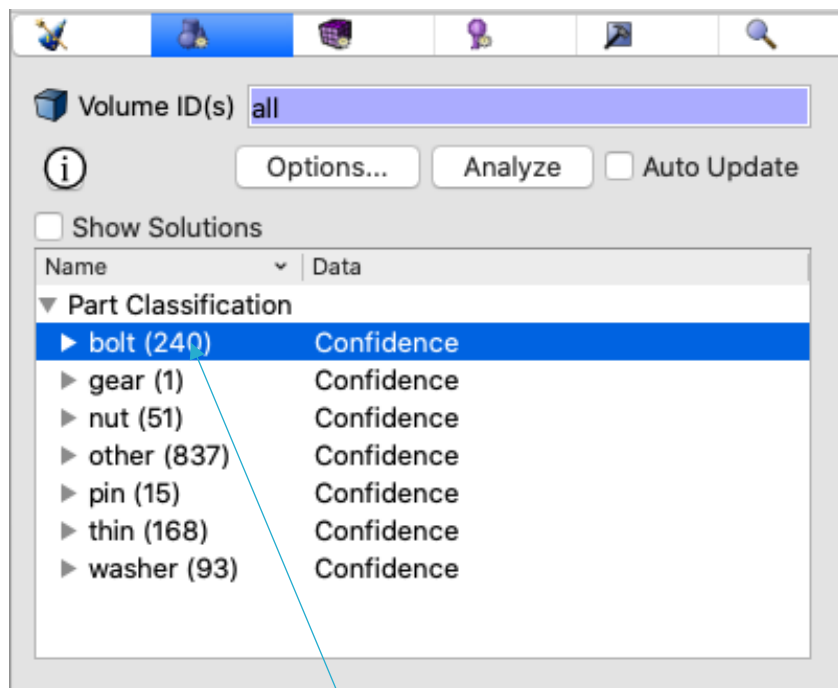


3D Solid Model: Full Assembly  
(1048 Solid Volumes)

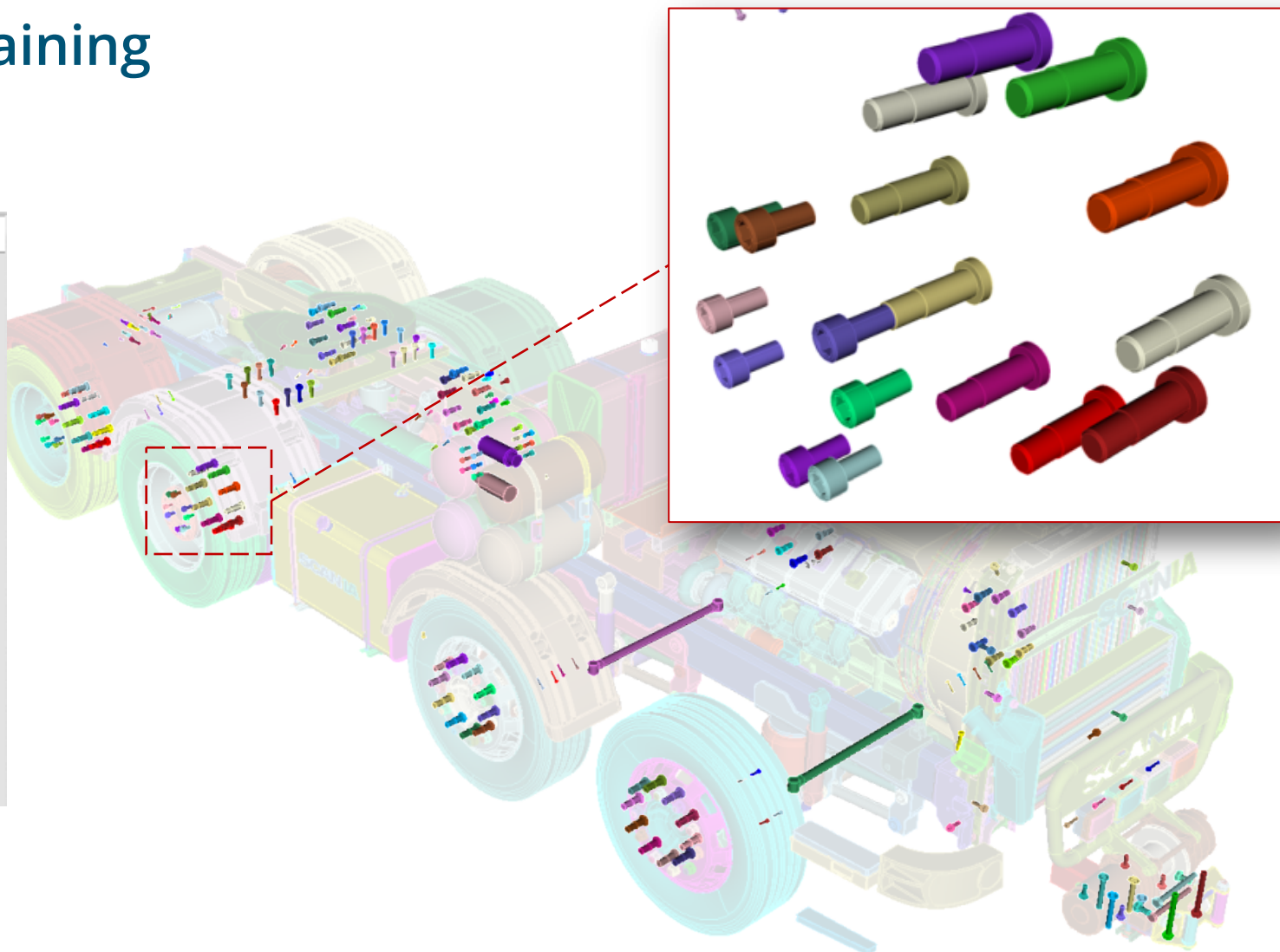




# In-situ user driven training



Existing “bolt” category  
Classified 240 bolts

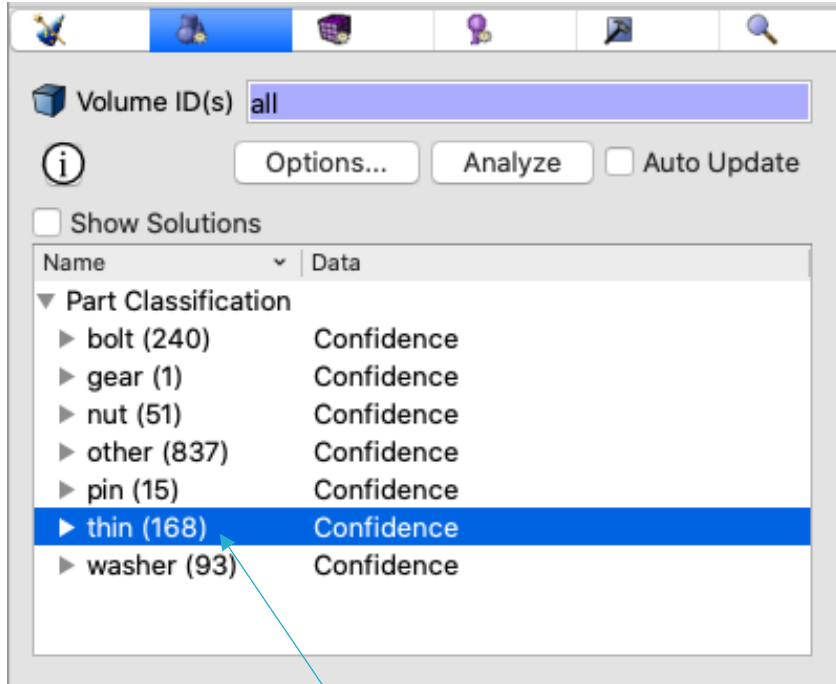


3D Solid Model: “Bolt” Volumes  
Only  
(240 Volumes)

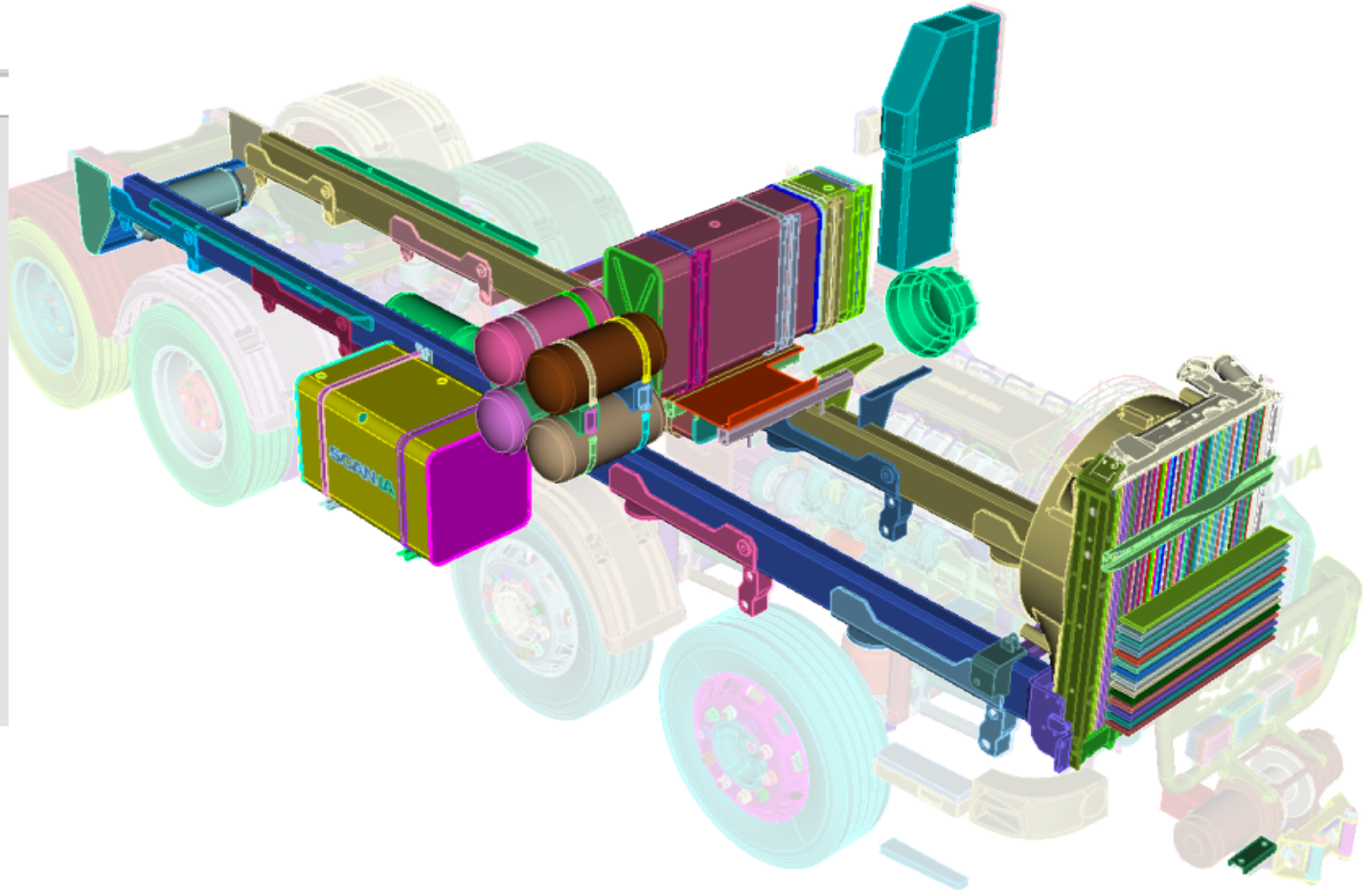


# In-situ user driven training

CAD assembly of  
commercial 18 wheel truck  
(source: grabcad.com)



New user-defined “thin” category  
Classified 168 thin volumes



3D Solid Model: “Thin” Volumes  
Only  
(168 Volumes)



# Rapid CAD to Simulation with Machine Learning

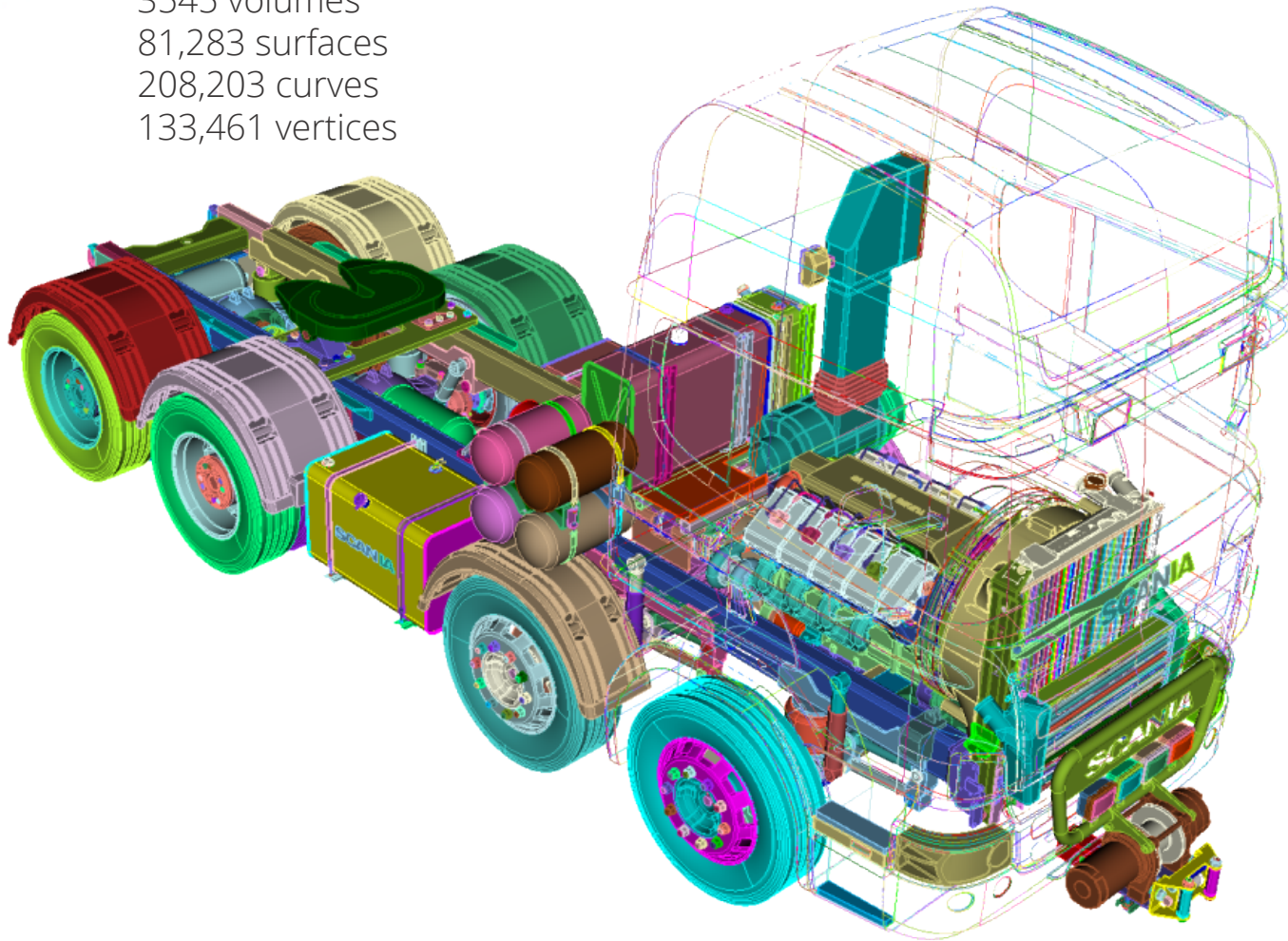
Commercial Truck CAD Model

3545 volumes

81,283 surfaces

208,203 curves

133,461 vertices



## Machine Learning for Analysis Preparation:

- Defeaturing
  - Predicts potential problems in mesh
  - Predicts tools that will best improve
  - Allows user to select and validate
- Part Classification
  - Identifies common mechanism types
  - Tools to quickly reduce to analysis-ready
  - In-situ training on user-defined categories

## What's next

- New analyst-driven custom **reduce** solutions
- Reinforcement Learning
- ML-driven Beam and Shell Modeling