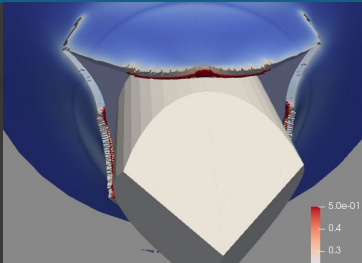
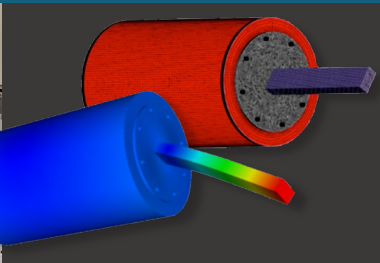
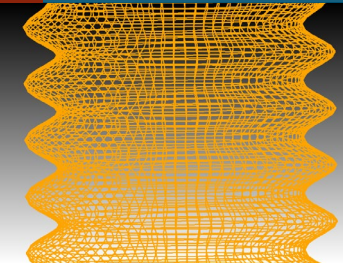
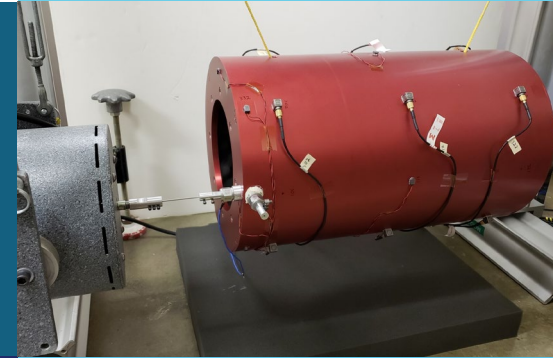


# Mapping from Low Fidelity to High Fidelity Analysis for Failure Quantities of Interest



*Mentors:*

Mark Merewether, Edmundo Corona, Peter Grimmer, & Brendan Donohoe

*Students:*

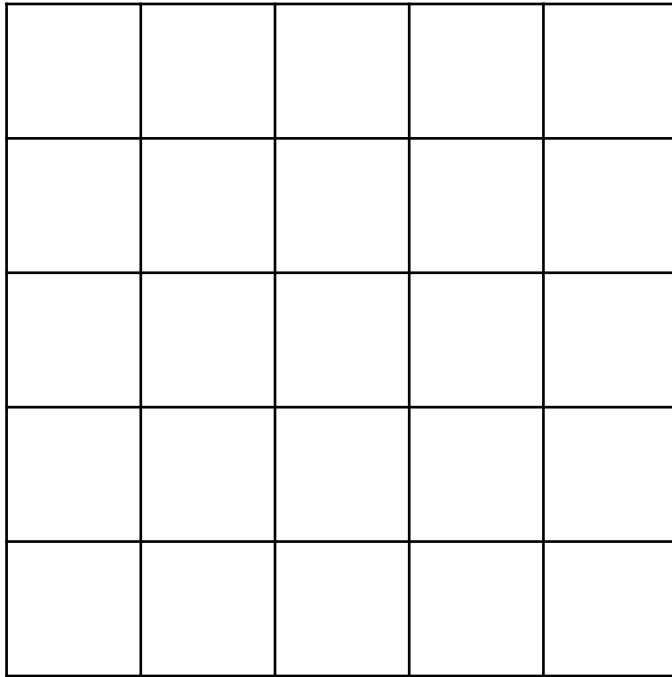
Christopher Leonard, Joseph Redmond, & Leah Brinkman



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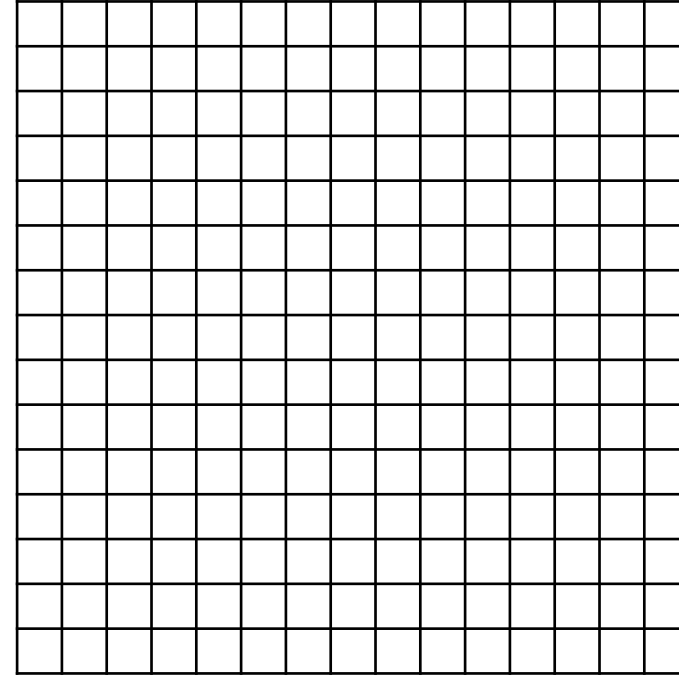
Using a low fidelity model, can one predict failure seen in higher fidelity models?

Lower Fidelity



shell element models  
element death failure models

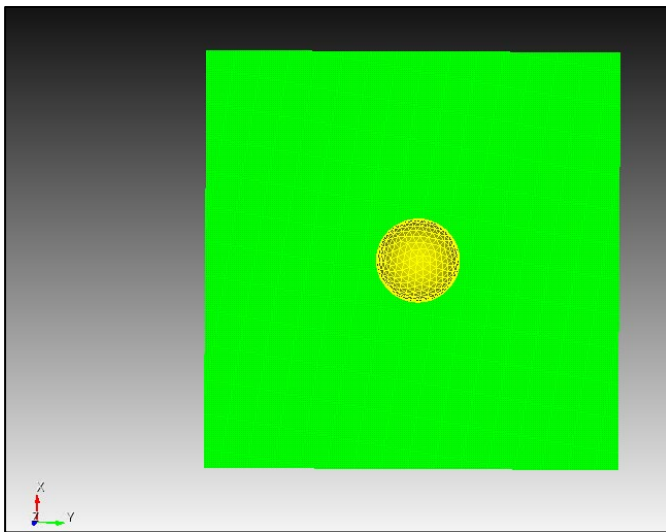
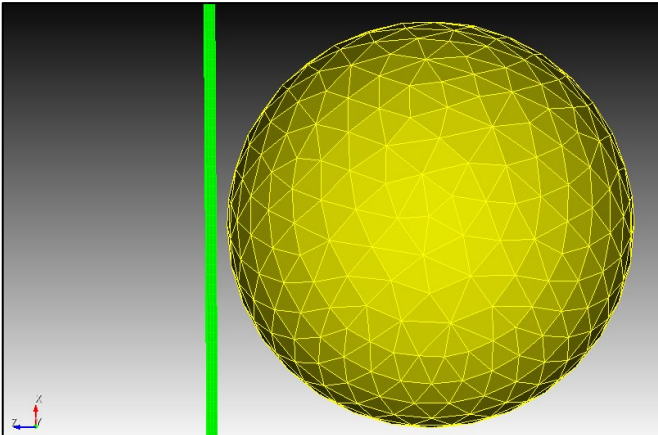
Higher Fidelity



hexahedral element models  
XFEM failure models



- Introduction
- Material Models
- Hexahedral (Hex) Model
- Shell Model
- Comparison of Hexahedral and Shell Models
- Neural Network
- Conclusions



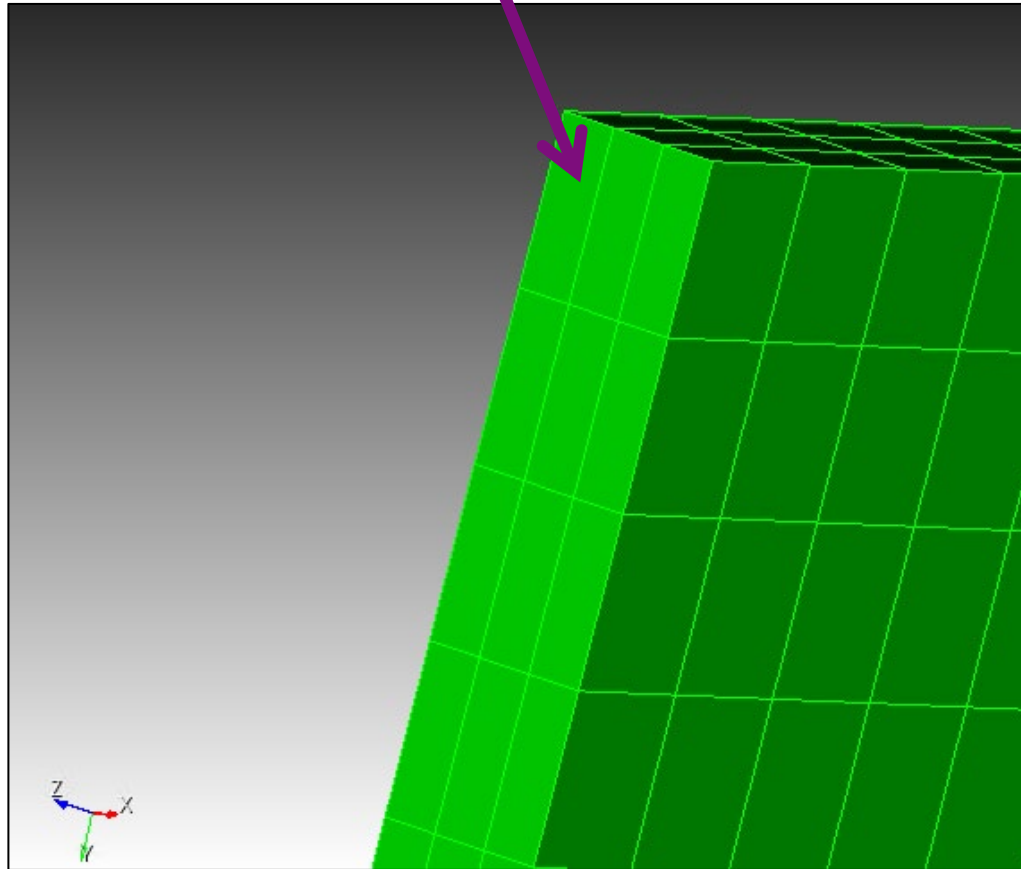
- Ball
  - 5" diameter
  - Made up of solid tetrahedral elements
  - Mesh size of 0.5"
  - 0.2" from plate in - z-direction
  - Initial velocity
- Plate
  - 25" x 25" x 0.12" square plate
  - Made up of either hexahedral (hex) or shell elements
  - Varying mesh sizes
  - Similar hex and shell models developed for comparison
  - Fixed on edges

Hexahedral and shell models developed at different levels of fidelity for comparison.

# Hexahedral vs. Shell Elements

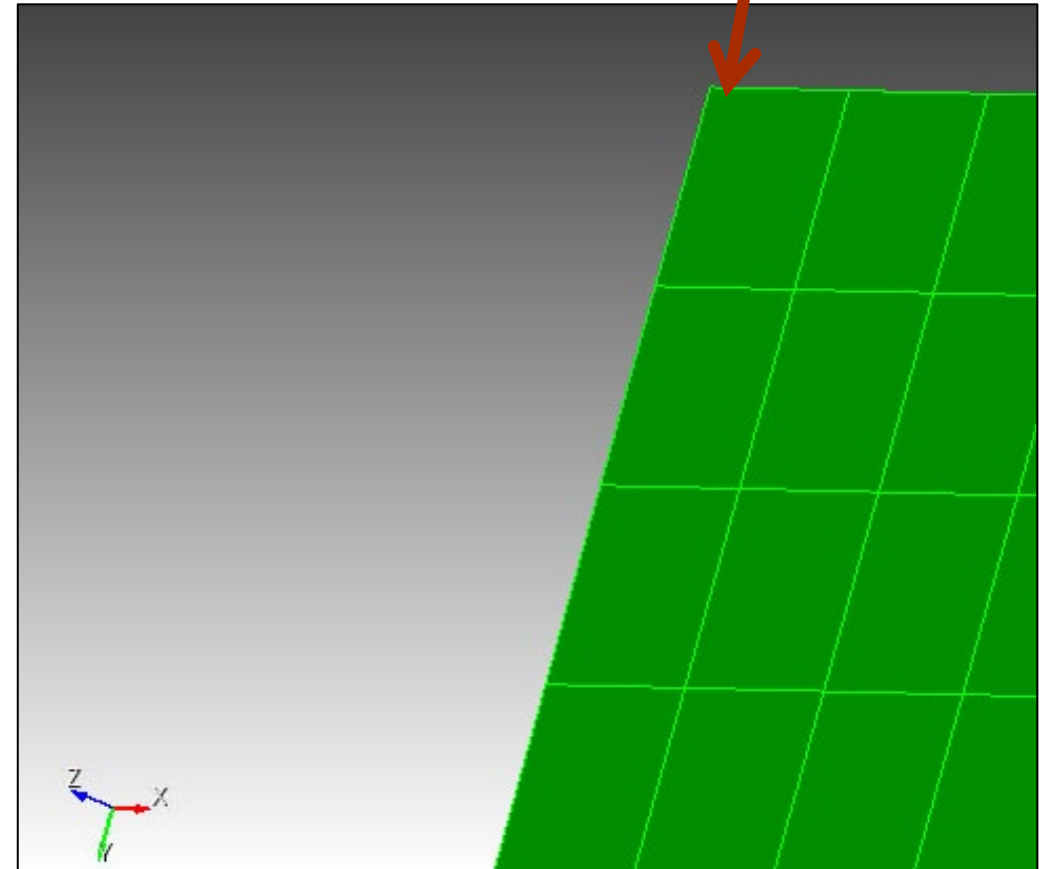


solid element



Hexahedral Plate

thickness  
stored as  
parameter



Shell Plate

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# Relating Hexahedral & Shell Elements



Relating hexahedral and shell elements will be achieved by comparing the following:

- Breakthrough velocity magnitude
  - How fast must the ball travel to break through the plate?
  
- Kinetic energy change
  - What is the change in energy of the projectile from the beginning time step to the end?
  
- Size of hole at ball speed of 5000 in/s
  - How much destruction is measured for each case?





**Plate: 6061-T651 Aluminum Alloy**

Hex-Based Setup

J<sub>2</sub> Plasticity Model

Shell-Based Setup

Modular Plane Stress Plasticity Model

All components are set up with ductile failure models: J<sub>2</sub> plasticity for solid elements and modular plane stress plasticity for shell elements.

**Ball: 304L Stainless Steel Alloy**

All Setups

J<sub>2</sub> Plasticity Model

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# J<sub>2</sub> Plasticity



## Fixed Parameters

### Aluminum

Parameter	$\rho$ (blob/in <sup>3</sup> )	E (psi)	$\nu$	$c_p$ (in-lb/(blob °C))
Value	$2.5 \times 10^{-4}$	$10.4 \times 10^6$	0.33	$1.36 \times 10^6$

### Steel

Parameter	$\rho$ (blob/in <sup>3</sup> )	E (psi)	$\nu$	$c_p$ (in-lb/(blob °C))
Value	$7.49 \times 10^{-4}$	$28 \times 10^6$	0.27	$0.776 \times 10^6$

Fixed parameters do not vary with plastic deformation.

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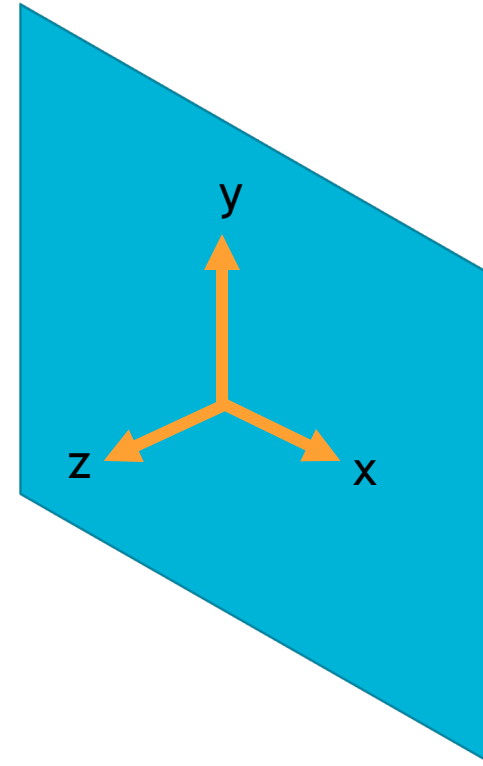


# Modular Plane Stress Plasticity



- A  $J_2$  plane-stress model with modified forms for hardening
- Uses the same values from the  $J_2$  plasticity model
- Developed for use with shell elements

Only the plane stress state is allowed in shell elements.



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- Element death will be defined using the damage variable  $D$
- Factors of the damage variable
  - Calculated such that material failure occurs when damage  $\geq 1$
  - Accumulates with plastic deformation
  - Functional dependency chosen to be on the stress, equivalent plastic strain rate, and temperature histories

$$D = \frac{1}{d_{crit}} \int_0^{\epsilon^p} f(\sigma, T, \dot{\epsilon}^p) d\epsilon^p$$

Element death occurs when the variable damage  $\geq 1$ , which accumulates with plastic deformation.

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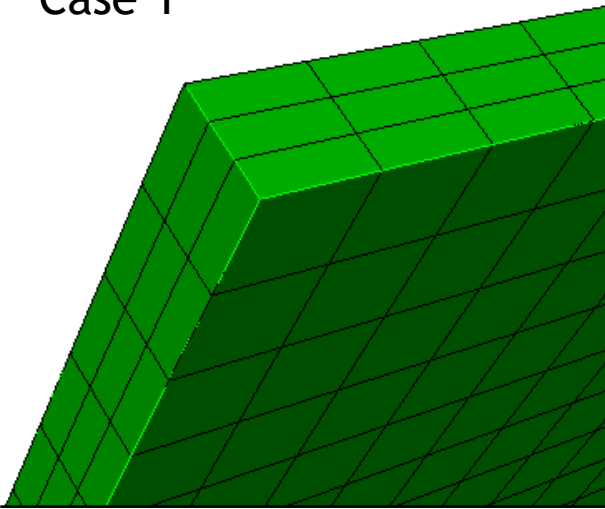
Conclusions

## Plate Mesh Sizes Considered

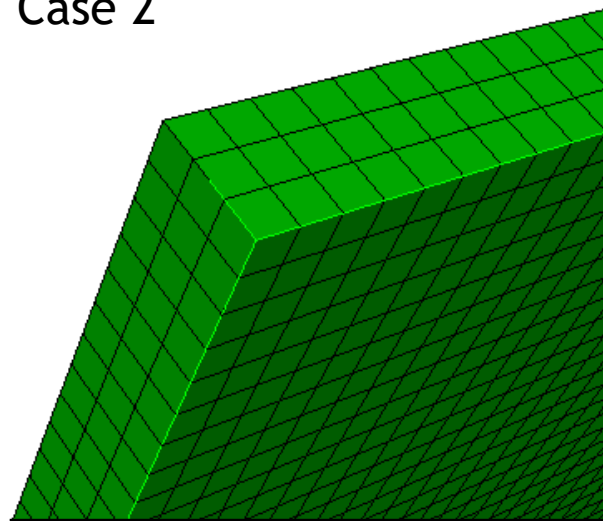
Case Number	Element Side Length Across Face (in)	Number of Elements Through Thickness	Total Number of Nodes	Aspect Ratio
1	0.12	3	175,848	3
2	0.04	3	1,568,268	1
3	0.02	6	10,956,131	1

Varied number of elements through the thickness and across faces of plate.

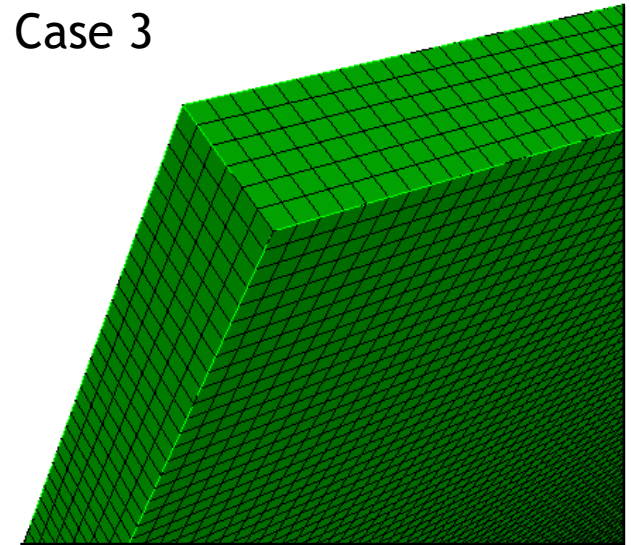
Case 1



Case 2



Case 3



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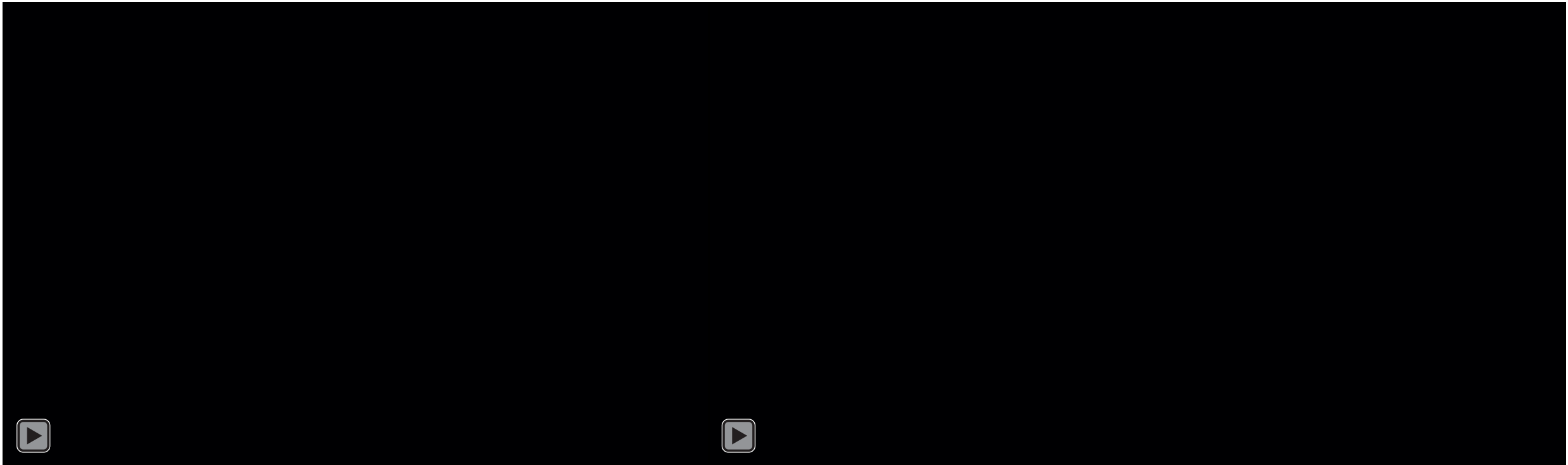
Neural Network

Conclusions



## Case 3 - Most Refined

## Case 1 - Least Refined



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Case Number	Breakthrough Velocity (in/s)
1	1388
2	1363
3	1187-1250

Breakthrough velocity decreases with mesh refinement.

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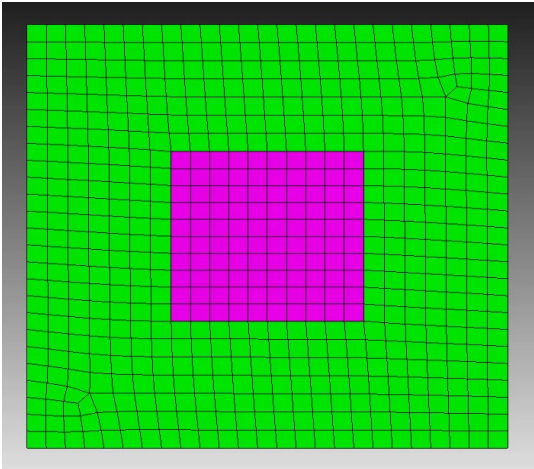
Neural Network

Conclusions

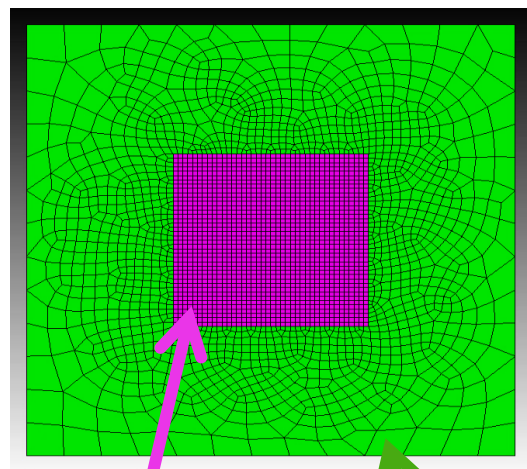
# Shell Development



Case 1



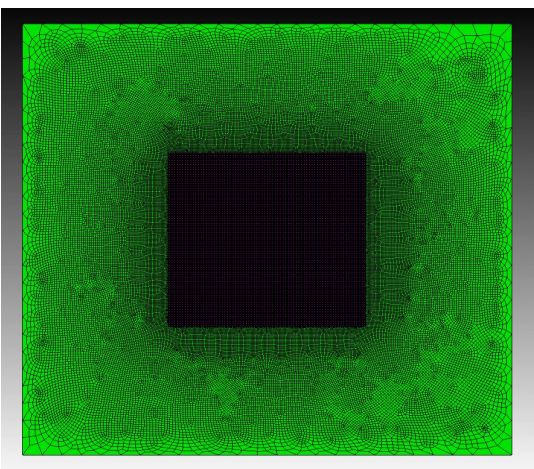
Case 3



Fine Interval Mesh

Graded Pave Mesh

Case 5



## Shell Intervals Evaluated

Case	Element Interval (in)	Outer Interval Pave (in)	Total Number of Nodes
1	1	1	1764
2	0.5	2	2,006
3	0.25	2	4,193
4	0.12	1	14,627
5	.04	1	110,920
6	.02	N/A	1,566,125

Shell Models have greater utility in lower fidelity schemes, as they can be localized, and do not have to adhere to aspect ratio limitations





Case 1 - 1" Mesh

Case 4 - 0.04" Mesh

Case 6 - 0.02" Mesh



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### Breakthrough Velocity

Mesh Size (in)	Breakthrough Velocity (in/s)
1	1083
.5	844
.25	795
.12	765
.04	516

Disparity in Velocity Threshold is greater relative to Hex models, but all thresholds are underestimated relative to hex models

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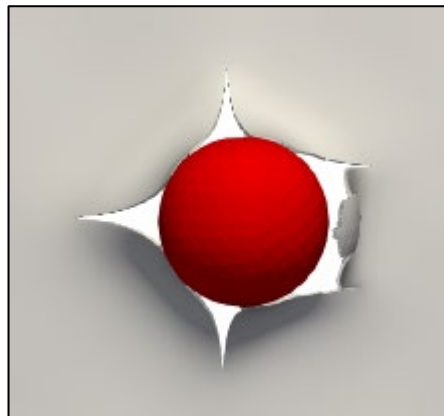
Conclusions



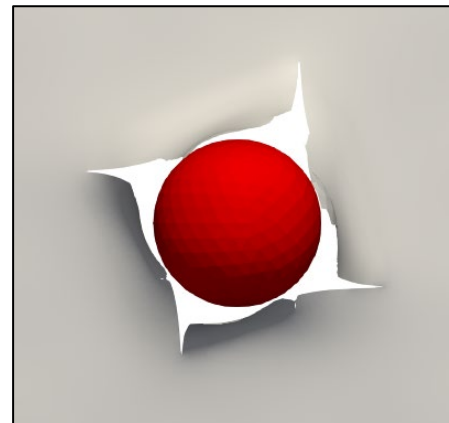
# Comparison of Shell and Hex



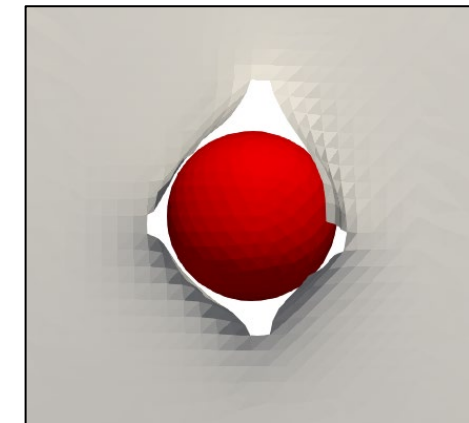
	Hex Model (.04)	Shell (.04)	Shell (0.5)
Computation Time (hour : minutes : seconds)	96:05:40	11:53:52	00:04:04
Projectile Kinetic Energy Loss (J)	7.02	7.80	12.67



Hex Model (.04)



Shell Model (.04)



Shell Model (0.5)

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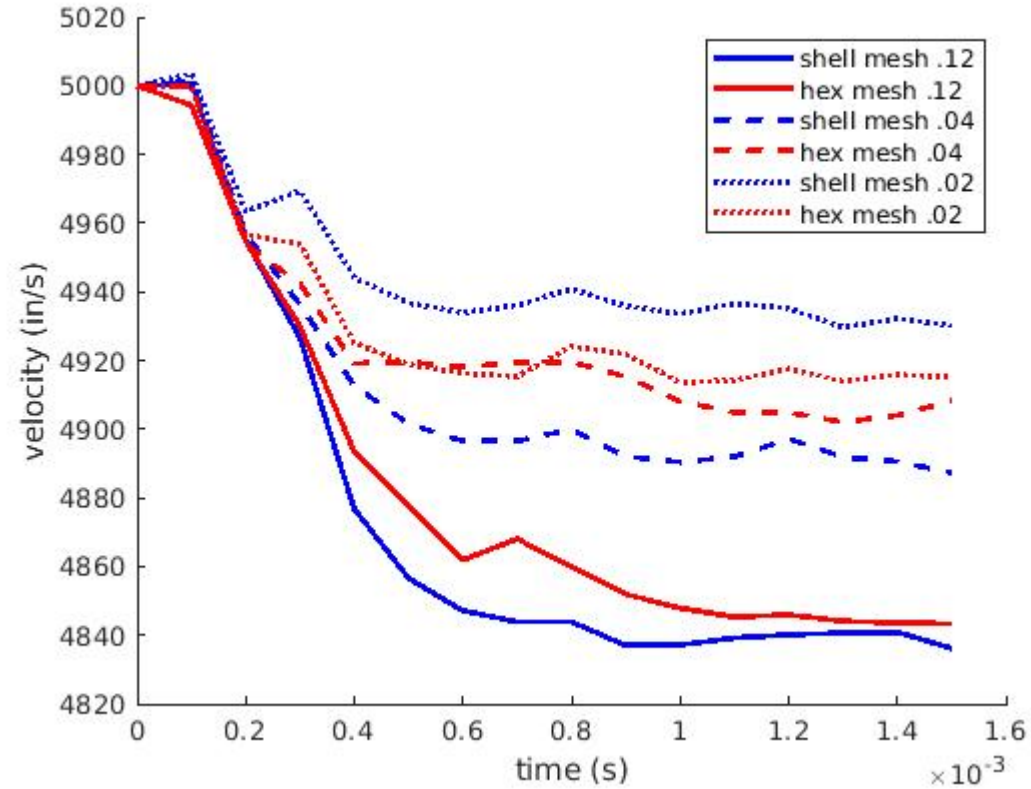
Shell Model

Comparisons

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Conclusions

# Comparison of Shell and Hex



Velocity of the ball vs. time for different levels of fidelity.

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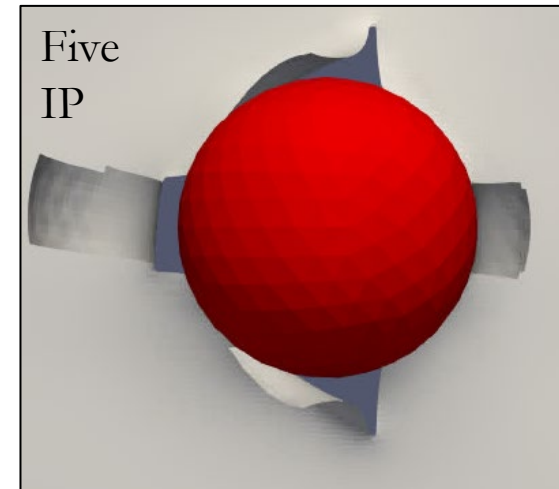
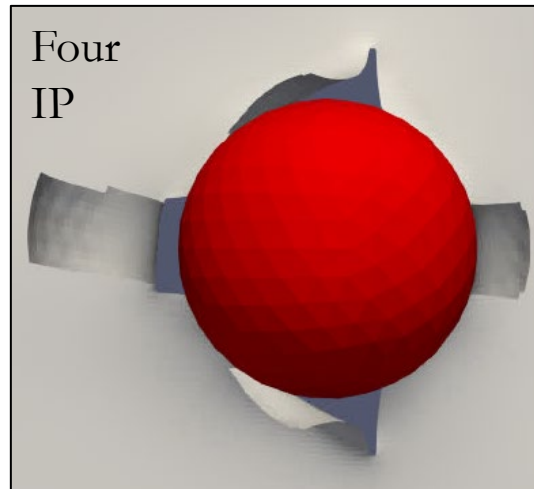
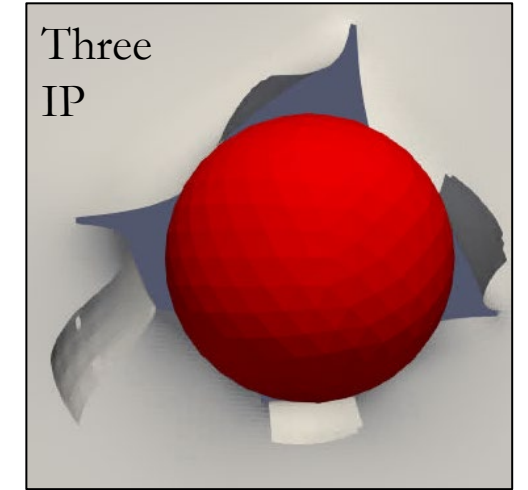
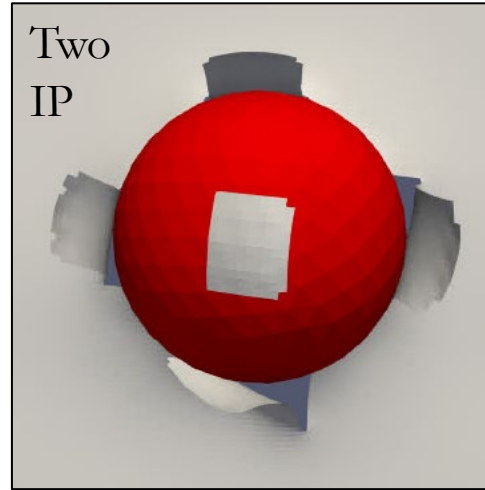
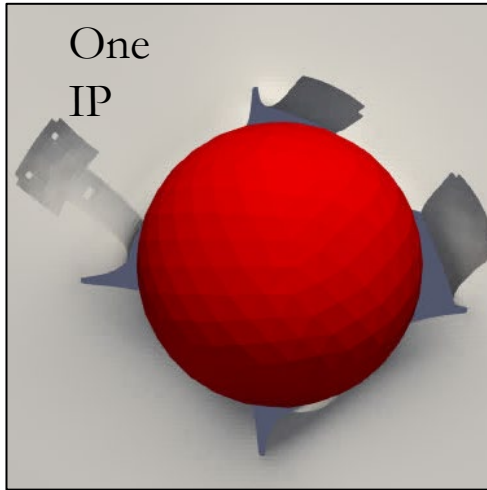
Neural Network

Conclusions

# Shell – Varying Integration Point Thresholds



## Number of Integration Points (IP) to Reach Death Criterion Before Element Killed



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Material Models

Hex Model

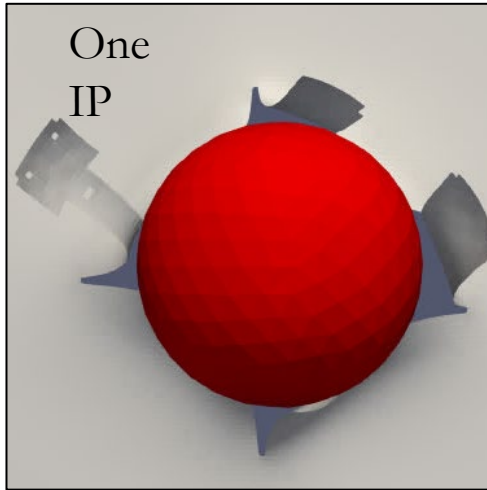
Shell Model

Comparisons

Neural Network

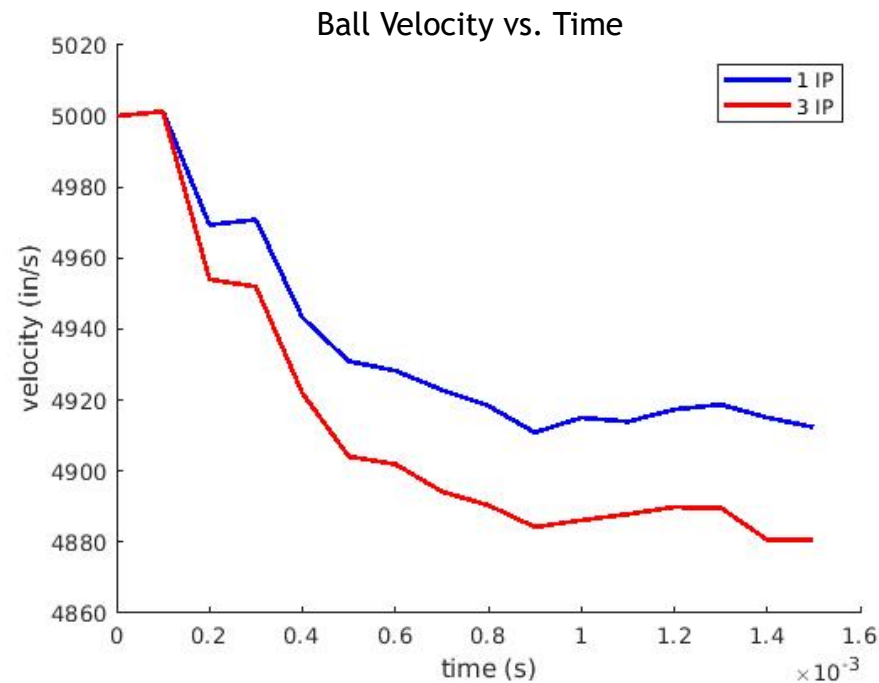
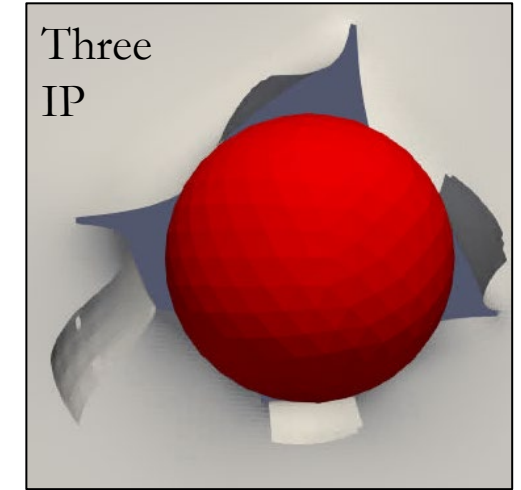
Conclusions

# Shell – Varying Integration Point Thresholds



Number of Elements  
Killed  
198

Number of Elements  
Killed  
196



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Material Models

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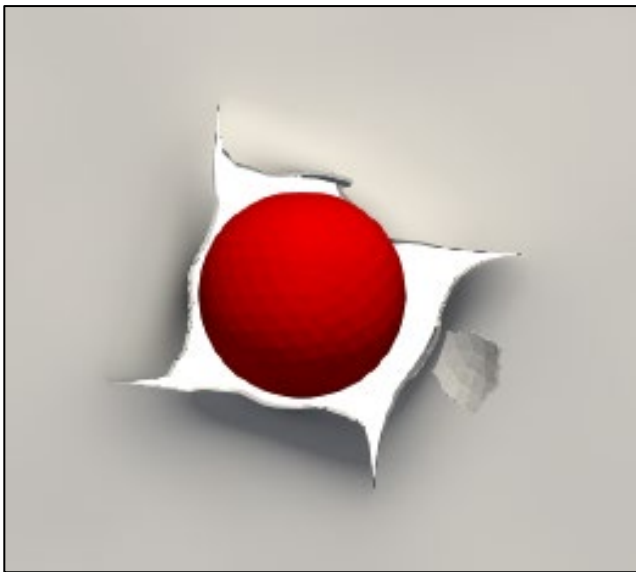
# Comparison of Hexahedral Element Death vs Shell XFEM



## Mass Lost

- Shell with XFEM - 0 lb
- Shell with Element Death -  $2.1 \times 10^{-5}$  lb
- Hex with Element Death -  $2.076 \times 10^{-5}$  lb (0.1% of starting mass)

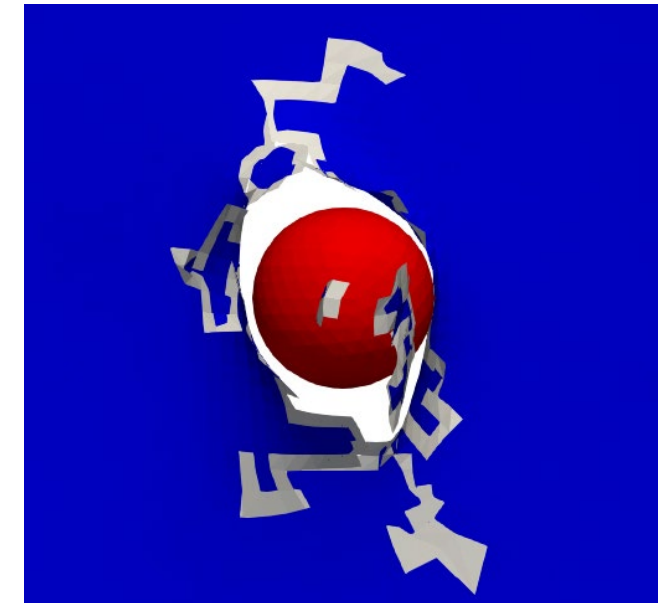
Case 3 Hex Model  
(Most Refined)



Case 2 Shell Model



Case 2 Shell Model  
XFEM



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Shell Model

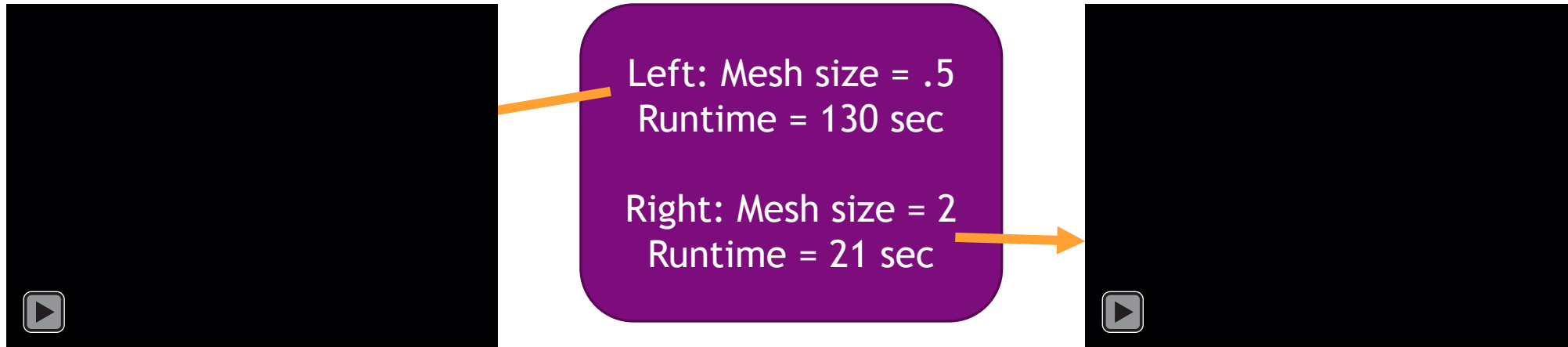
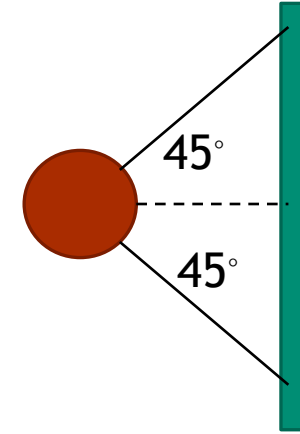
Comparisons

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# Neural Network problem

- A fully connected neural network is used to determine if there is a break in a plate given the initial velocity of the projectile.
- To train the neural network, data was gathered from simulations where the initial velocities magnitude and directions varied. This simulation was then used to determine if there was a break in the plate or not.
- With this neural network, we can run lower fidelity simulations and predict if there was break in the plate or not.



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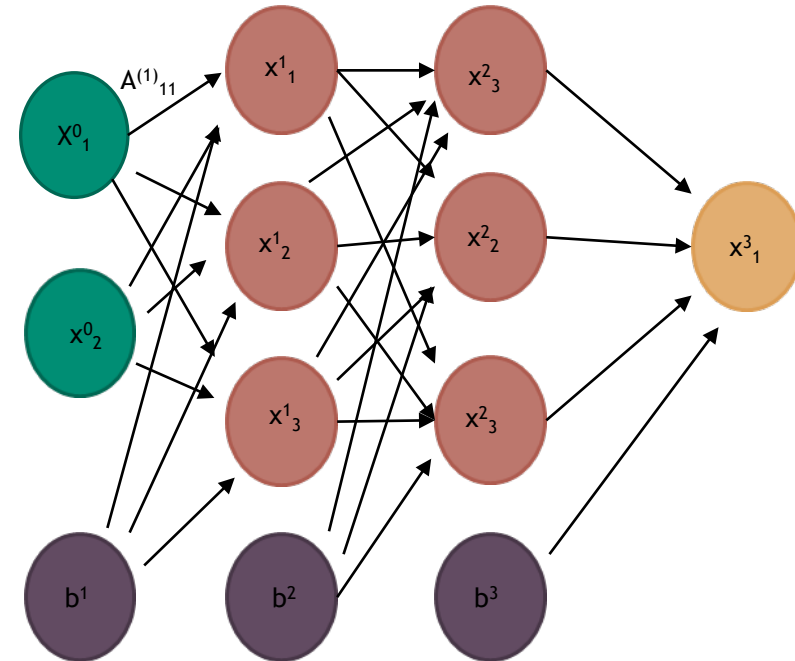
# Fully Connected Neural Network



- For a fully connected neural network each connection between layers can be represented as

$$\varphi^i(\mathbf{A}^i \mathbf{x}^{i-1} + \mathbf{b}^i) = \mathbf{x}^i$$

- Here  $i=1,2,\dots,n$ , where  $n-1$  is the number of hidden layers.
- The vectors  $\mathbf{x}^{i-1}$  are the inputs into the  $i$ th layer of the neural network.
- $\mathbf{A}^i$  and  $\mathbf{b}^i$  are the weight matrix and bias vectors respectively..
- The function  $\varphi^i$  is a an element wise function known as the activation function. This is used to add nonlinearity to the neural network.

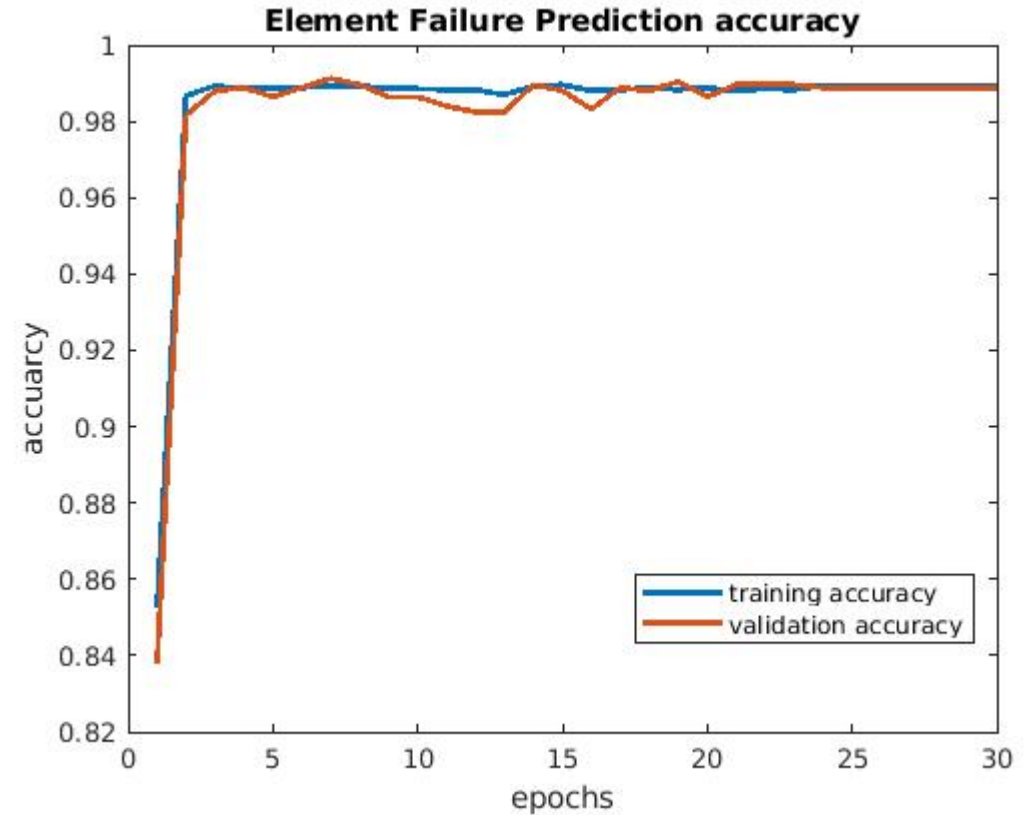


# Neural Network Training



- Trained the network using 30 epochs.
- Use the adam optimization algorithm.
- Total training time approximately 20 secs.

Prediction Accuracy  $\approx$  99%



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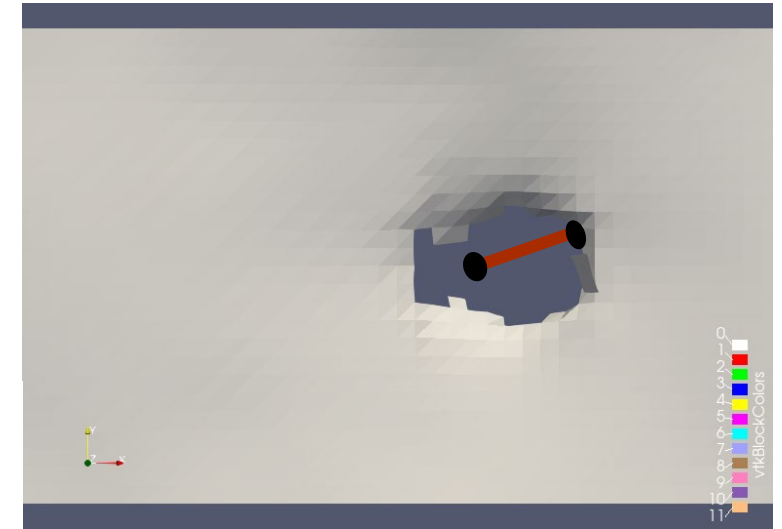
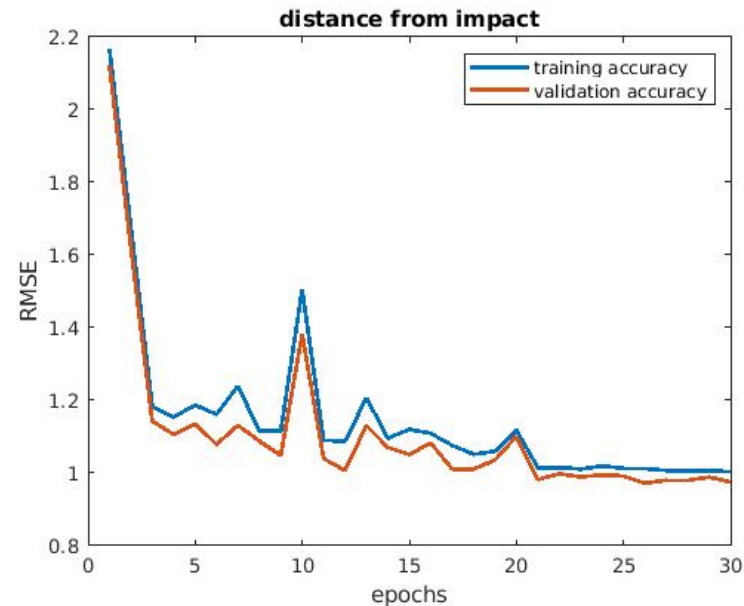
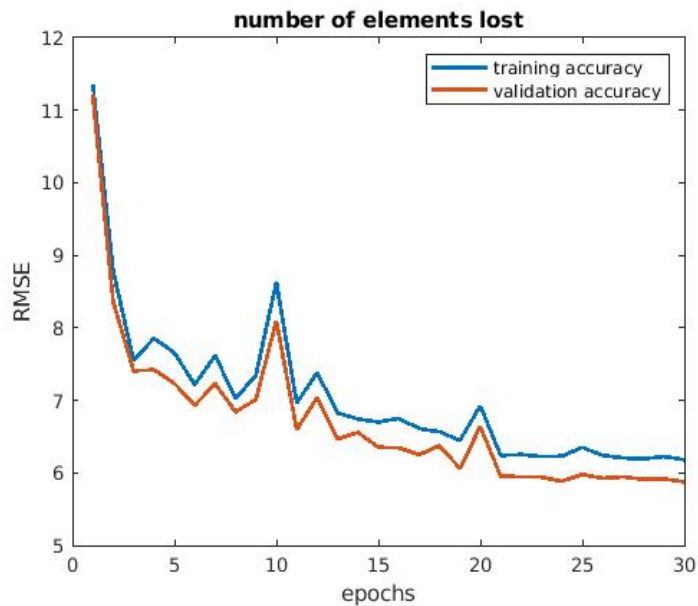
Neural Network

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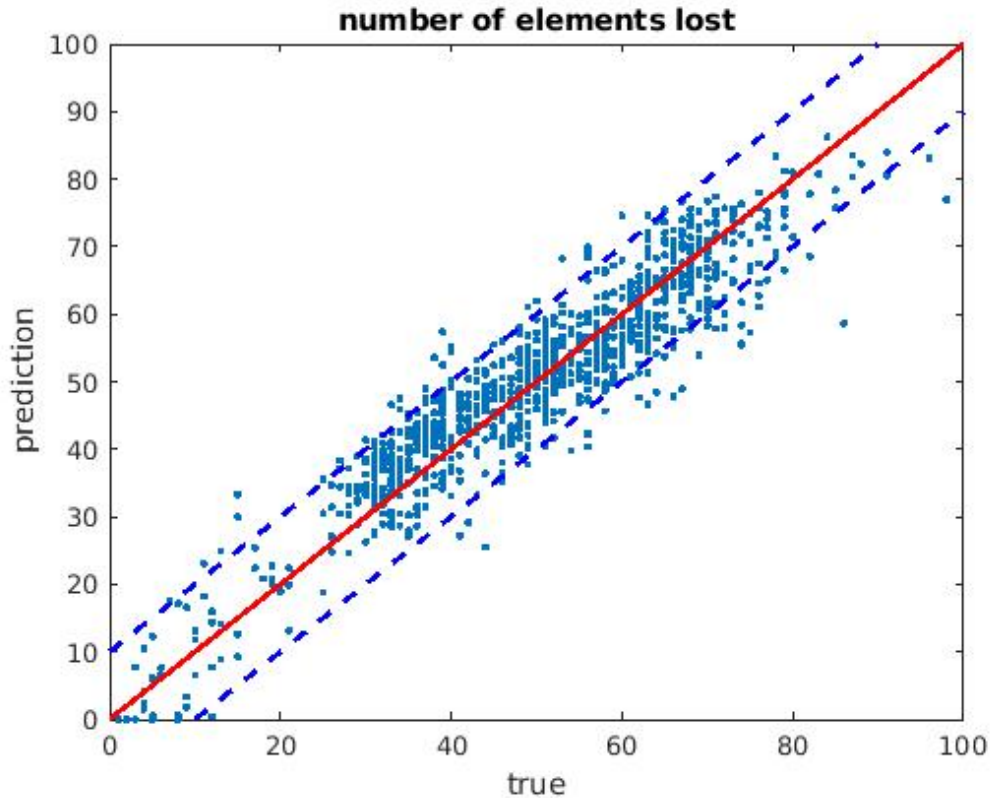




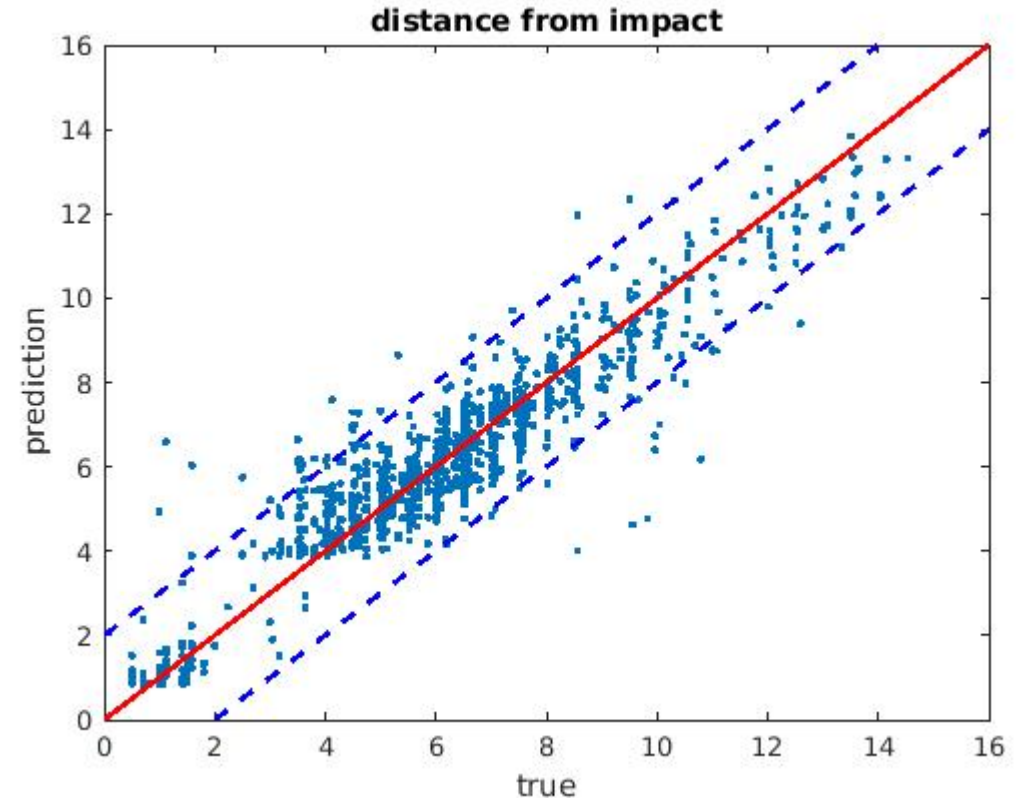
- Here, simulations are run to predict how large a tear there will be when there is element failure in the model.
- One fully connected neural network was used to make predictions on how wide the hole is and the total number of elements destroyed.



# Impact Predictions on the Validation Set



62.59% of  $|y_{\text{true}} - y_{\text{pred}}| < 5$  elements lost  
92.29% of  $|y_{\text{true}} - y_{\text{pred}}| < 10$  elements lost



77.44% of  $|y_{\text{true}} - y_{\text{pred}}| < 1$  inch  
94.92% of  $|y_{\text{true}} - y_{\text{pred}}| < 2$  inches

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- Developed quantitative and qualitative comparison of shell and hex models
- Looked at the usefulness of XFEM in coarse shell models for crack propagation compared to a refined hex model
- Quantified disparity in model behavior dependent on mesh resolution
- Able to accurately predict if there will be a tear in the plate given the projectiles velocity and give an estimate on how large the tear will be.





- Predict shape of the hole
- Train neural networks with other inputs, such as stress, strain, contact force, etc.
- Train neural network with higher fidelity model
- Study the differences in kinetic energy of the plate loss in element death vs. XFEM



# Acknowledgements



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