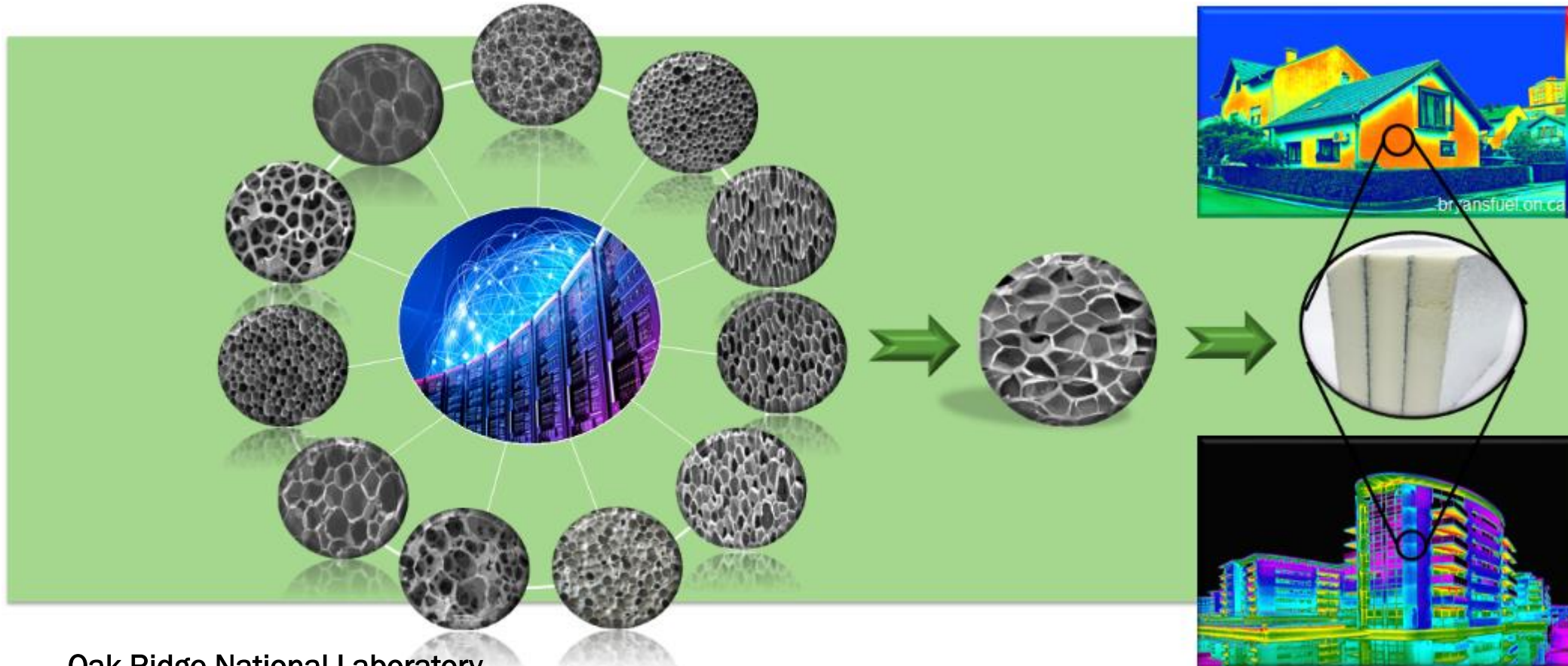


# Multi-Scale Simulations and Machine Learning-Guided Design and Synthesis of High-Performance Thermal Insulation Materials



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BTO-03.01.03.124, BENEFIT FOA 2196

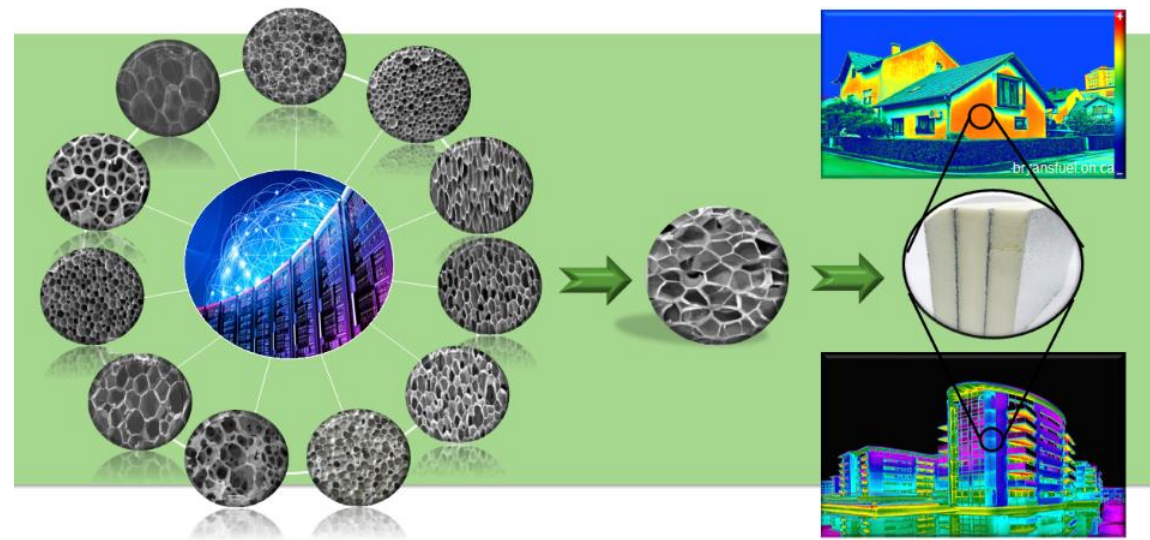
# Project Summary

## Objectives

- Develop novel foam insulation with higher thermal performance and comparable cost compared to the state-of-the-art.
- Enable building envelopes with tight spaces to meet thermal building code requirements.

## Outcome

R10/in foam that does not require vacuum and can be produced with minimal adjustments to current manufacturing practices.



## Team and Partners



## Stats

Performance Period: 10/1/21 to 9/30 24

DOE budget: \$1,600k, Cost Share: \$400k

Milestone 1: Develop foam design concepts for  $\geq R10/in$ .

Milestone 2: Produce  $\geq R10/in$  foams and cost of goods sold (COGS) analysis.

Milestone 3: Identify large-scale manufacturing process to produce new insulation with minimum modification to current process.

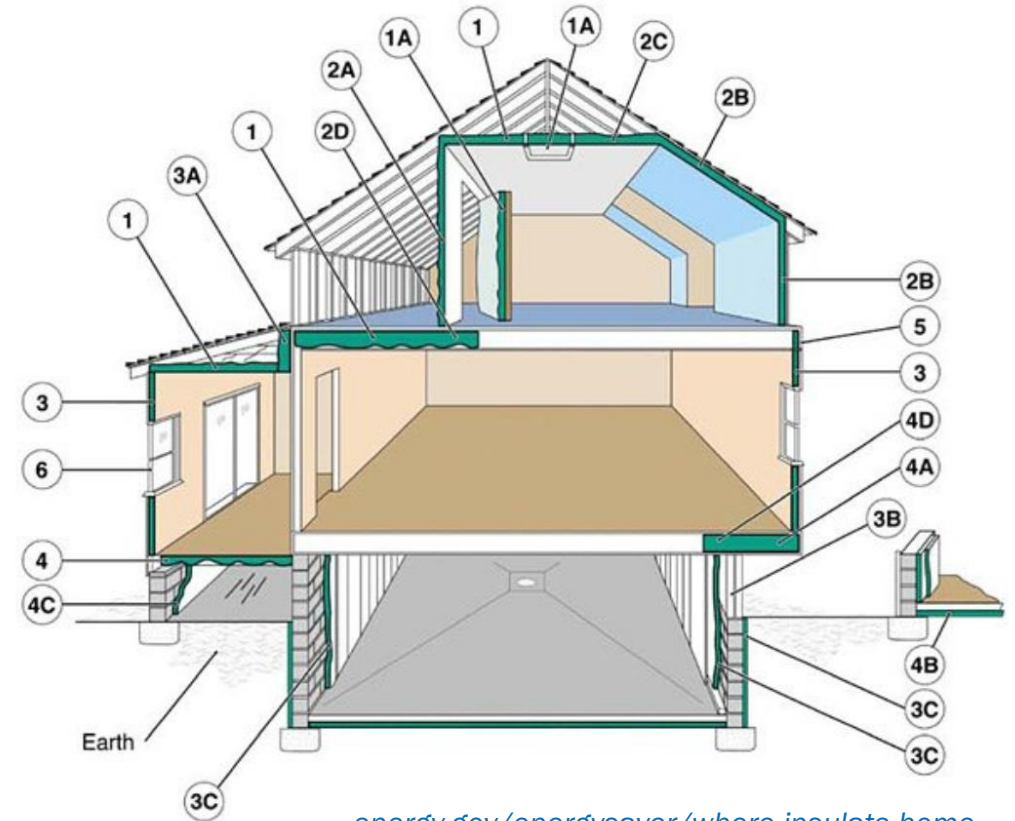
# Problem

- Around 44 % of US households spend > 8.6 % of their income on energy bills\*.
- High-performance insulation can improve building energy efficiency but currently used insulation limits ~R6/in.

## Solution:

- Multi-scale simulations and machine learning can guide the design and development of R10/in. foam that does not require evacuation or nanopores.

Envelope areas that require higher thermal performance

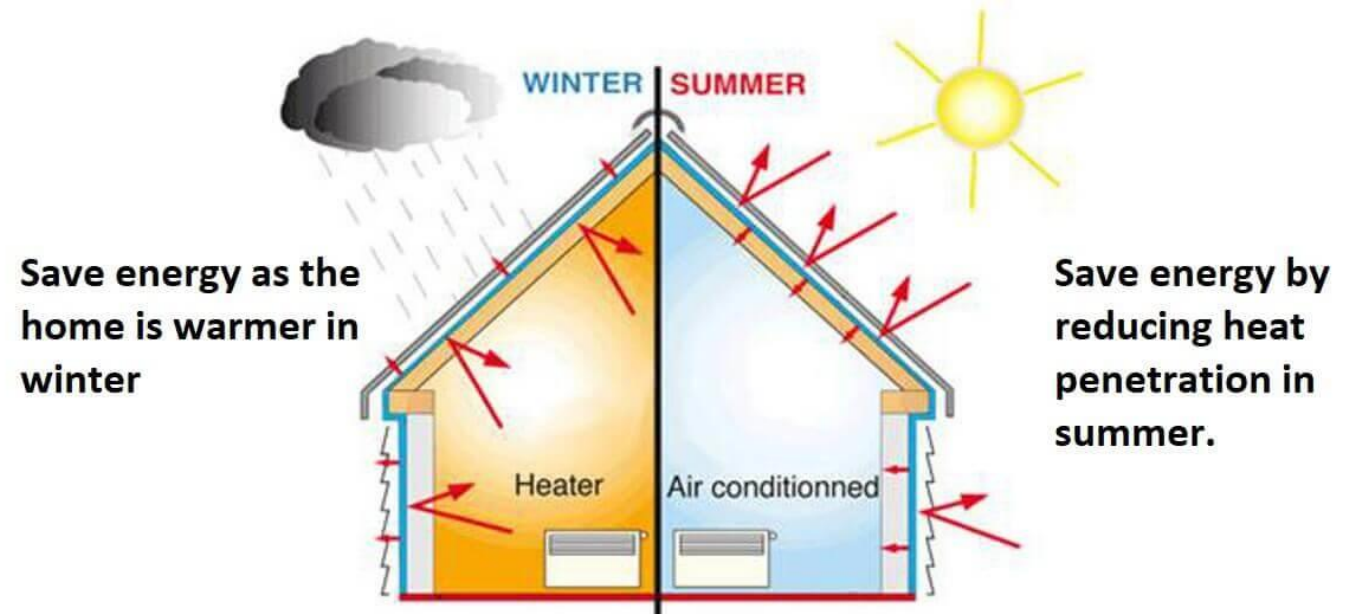


\* DOE LEAD Tool <https://www.energy.gov/scep/slsc/lead-tool>

# Alignment & Impact

- Insulation is a key contributor heating and cooling costs.
- Higher R-value of insulation
  - Can save **>1 quad energy/year\*\*** (~\$10 billion in energy cost).
- Successful R10 foam
  - Minimal modifications are required to the current manufacturing process.
  - Cost competitive with available insulation materials

**Increase building energy efficiency**  
Reduce onsite energy use intensity in buildings 30% by 2035 and 45% by 2050, compared to 2005



\*\* DOE, Research and Development Opportunities Report for Opaque Building Envelopes

Image: [roofinsulations.co.za/summer-insulation-benefits/](http://roofinsulations.co.za/summer-insulation-benefits/)

# State-of-the-art Insulation Materials

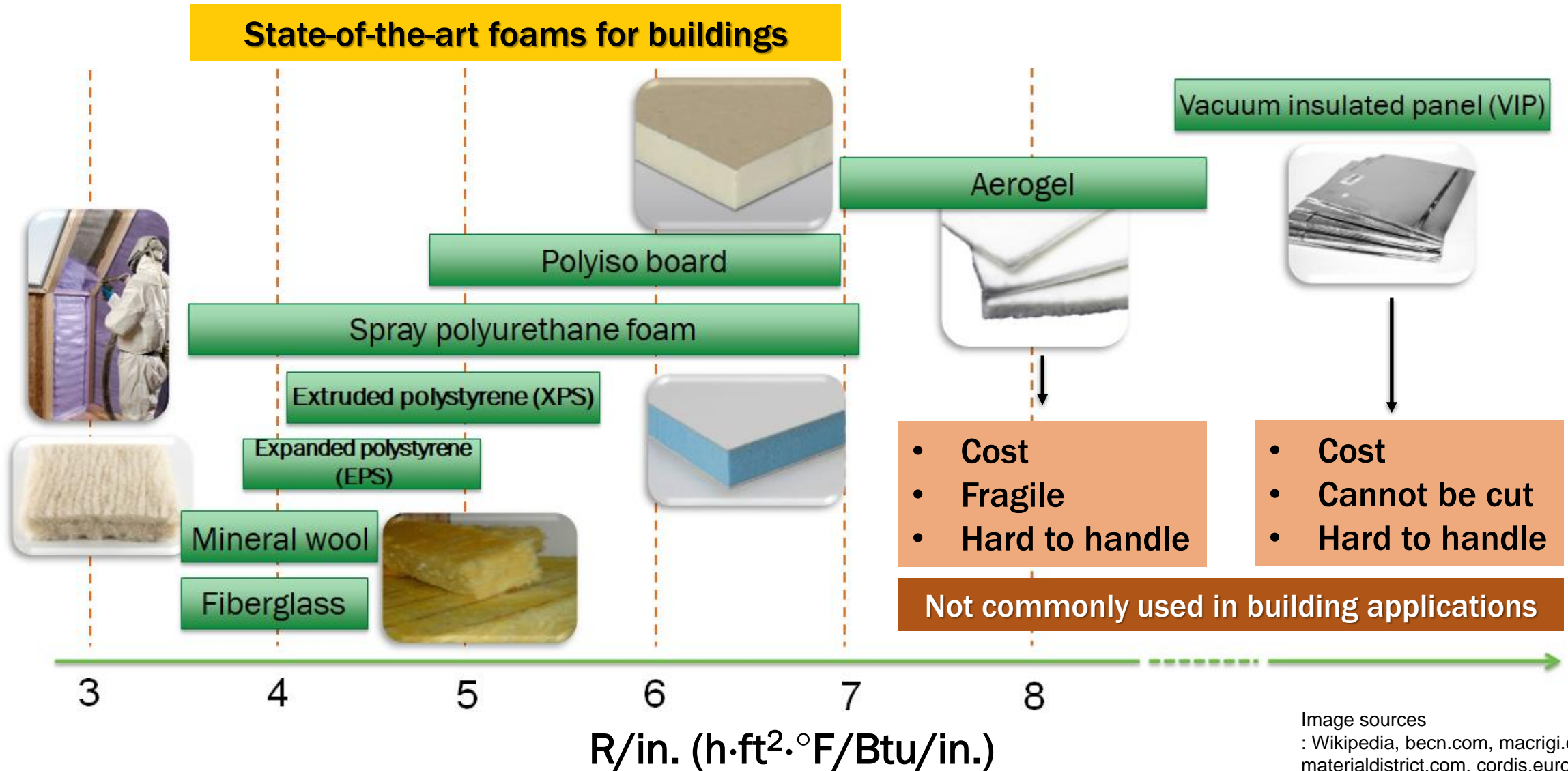


Image sources  
: Wikipedia, becn.com, macrigi.com,  
materialdistrict.com, cordis.europa.eu

# Challenges, Risks, Validation, and Commercialization

## Technical challenges

Translate simulation results into actual foam designs

## Risks and mitigation strategies

Numerous parameters affect thermal performance of foam insulation

- Use experimental results from the literature and this project to train ML algorithms that will identify and guide most relevant parameters

## Validation

Reproducibility of foam prototypes and thermal performance

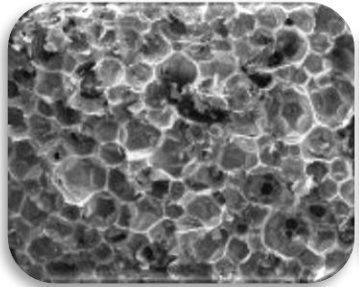
- Work with GAF to translate lab-scale to large-scale production

## Commercialization

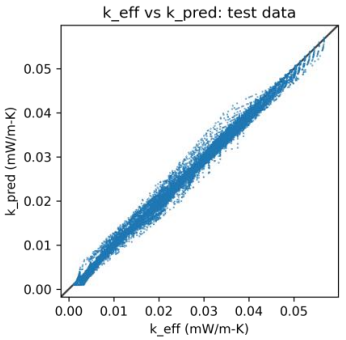
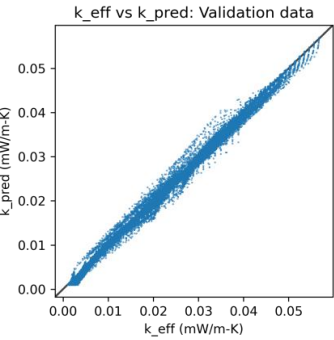
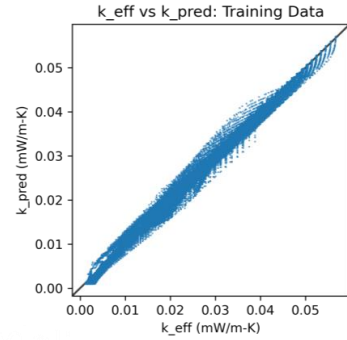
Feasible designs that consider scalability, cost, and durability

- Identify the design that requires minimum modifications in the current manufacturing process
- Cost comparable with current foam insulation on  $\$/\text{ft}^2/\text{R}$  basis

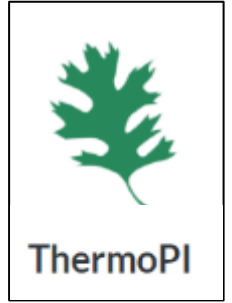
# Approach and Novelty



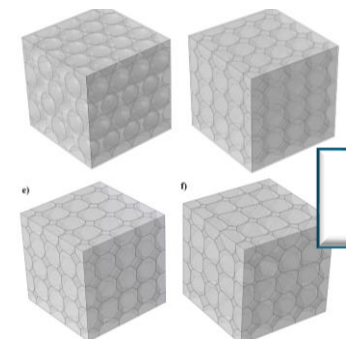
Experiment



Validation



+

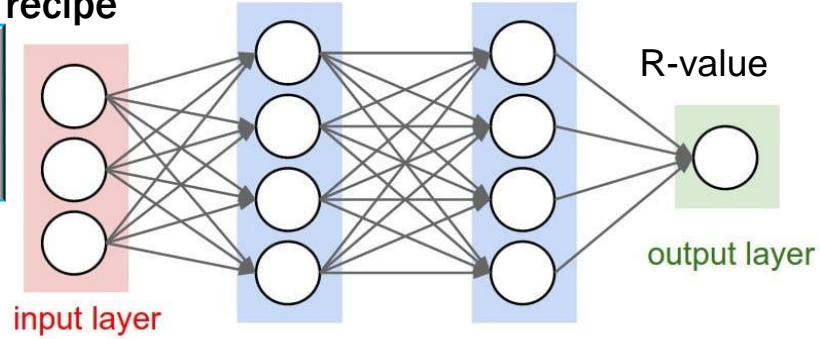


Simulation

Machine learning

Using experimental results to develop/train ML

Input geometry or recipe



hidden layer 1 hidden layer 2

Machine learning

Multi-scale model for  $\lambda_{gas}$ ,  $\lambda_{solid}$ ,  $\lambda_{rad}$

High-throughput finite-element

Simulation guides the design of foam insulation → Reduce trial and error

# Current Solutions and Challenges

Thermal conductivity of foams → three coupled heat transfer mechanisms

$$\lambda_{foam} = \lambda_{gas} + \lambda_{solid} + \lambda_{rad}$$

↑

- Porosity
- Cell size
- Gas species

↑

- Porosity
- Cell structure
- Material

↑

- Porosity
- Cell size
- Cell-wall thickness
- Opacifier

- Tradeoff between heat transfer mechanisms.
- Current commercial PIR foam is already at the optimal point (for pentane) → limitation of current solutions
- ~97% porosity is optimum.

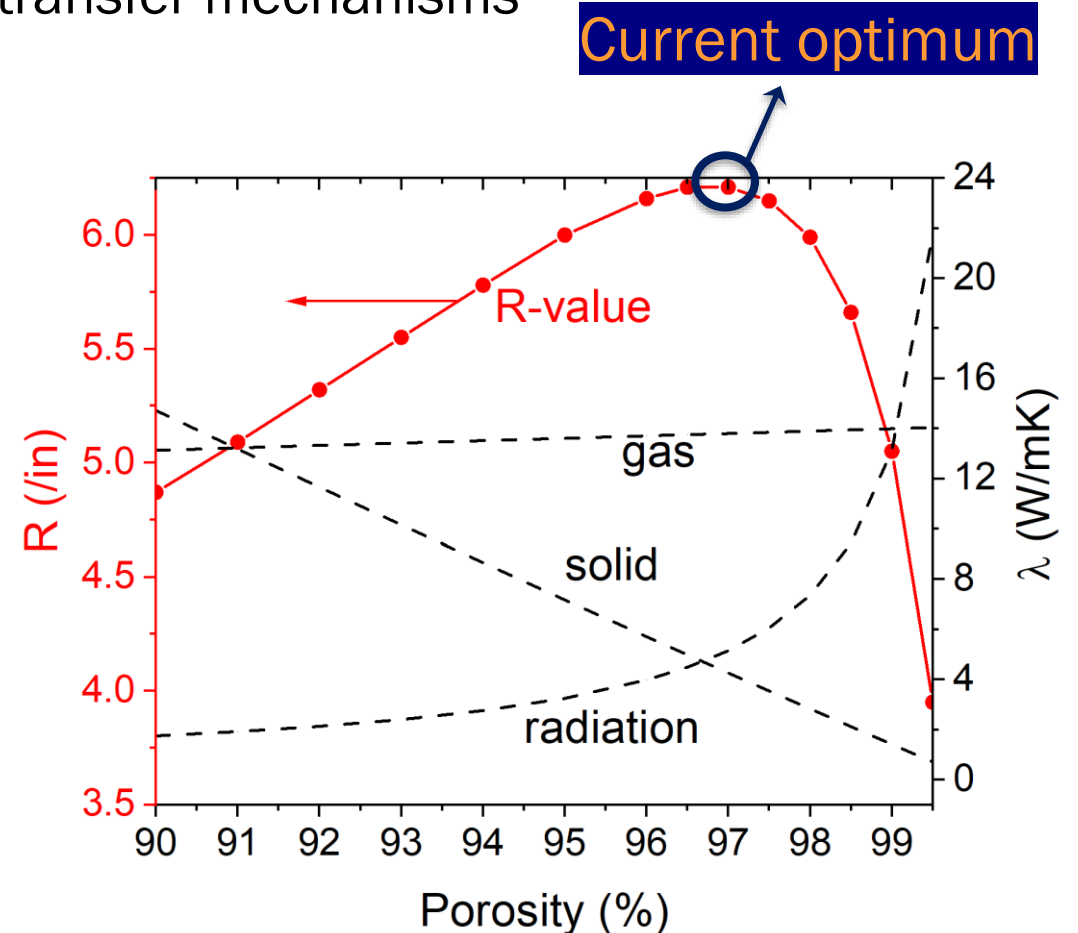
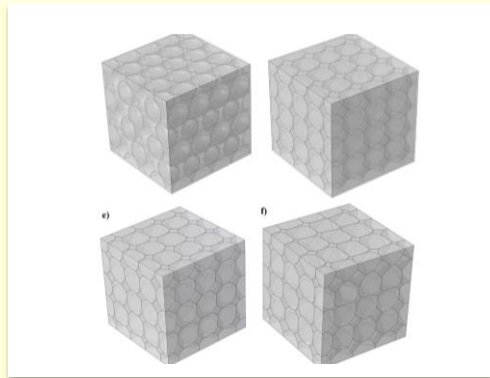


Figure generated using **ThermoPI (ORNL)**  
[thermopi.ornl.gov/calculate](http://thermopi.ornl.gov/calculate)

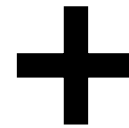


# Approach and Novelty: Simulation Guidance to Foam Designs

## Simulation



Pore geometry  
(FE simulation)



$$\lambda_{foam} = \lambda_{gas} + \lambda_{solid} + \lambda_{rad}$$

- ↑
- Porosity
  - Cell size
  - Gas species

- ↑
- Porosity
  - Cell structure
  - Material

- ↑
- Porosity
  - Cell size
  - Cell-wall thickness
  - Opacifier



ThermoPI

Consider different approaches to overcome current limits  
(multi-scale simulation)



- Open-cell foam
- or
- Closed-cell foam ✓

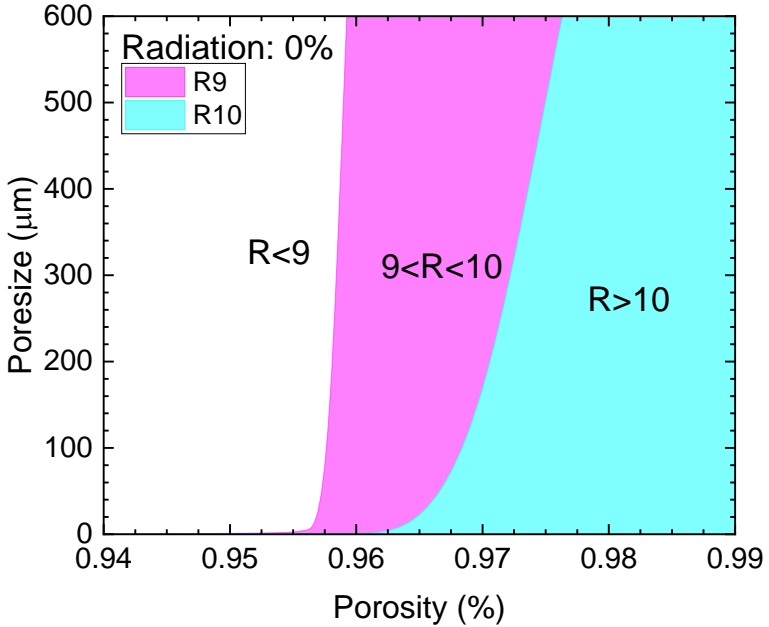
- Reduce cell size to nanometers ✓
- Control cell morphology ✓
- Remove radiation (add opacifier) ✓

- Optimize formulation ✓
- Optimize co-blowing agents ✓

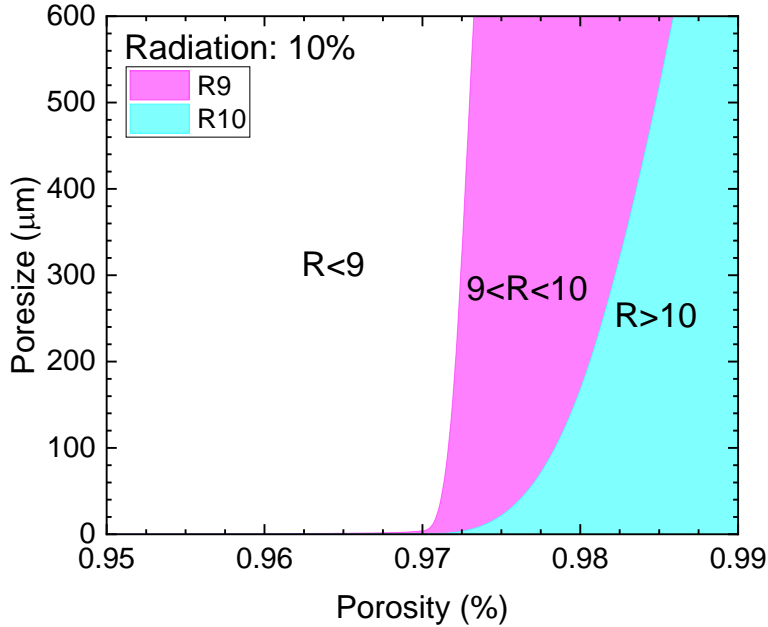
✓ : Picks for our work

# Approach/Progress: R-10/in. Design Concepts for Closed-cell Foam

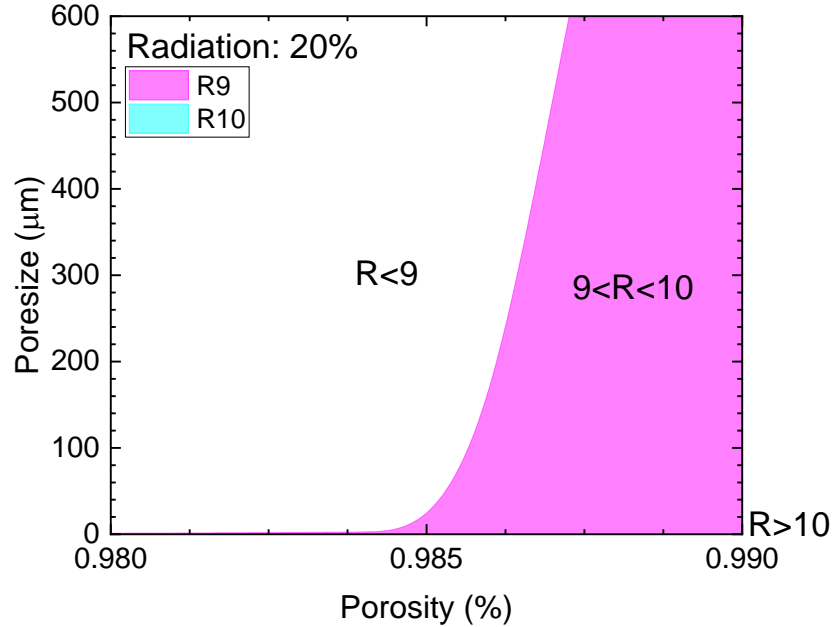
Closed-cell (Polyiso) design space with HFO as blowing agent:



- If radiation = 0
- R10 is feasible



- If radiation = 10%
- R10 is feasible



- If radiation = 20%
- R10 is challenging
  - R9 is possible

With reduced radiation, R10/in. can be achieved without nm-scale cell structure

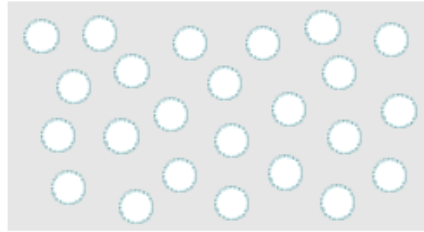
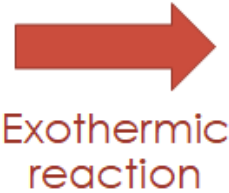
# Approach/Progress: Polyiso Foam Formation

- Experimental procedure:

Expanding until gas is confined in cells



Dispersed physical blowing agent (liquid status)

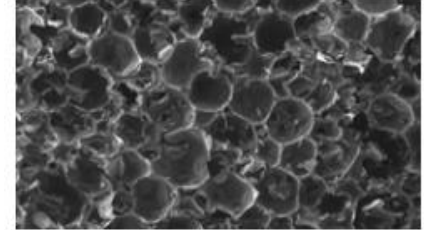


Expansion of blowing agent (Gaseous status)

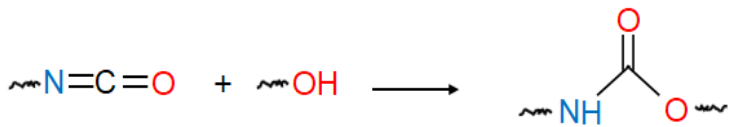
Reaction happens quickly



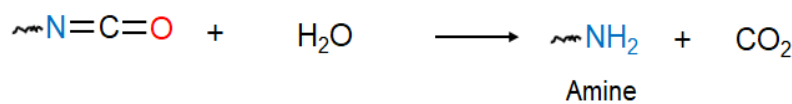
Packing of cells (cell-foam formation)



Exothermic reactions : Isocyanate reacts with polyol, water, isocyanate



Isocyanate Polyol Urethane



Amine



Isocyanate water Urea + CO<sub>2</sub> ↑



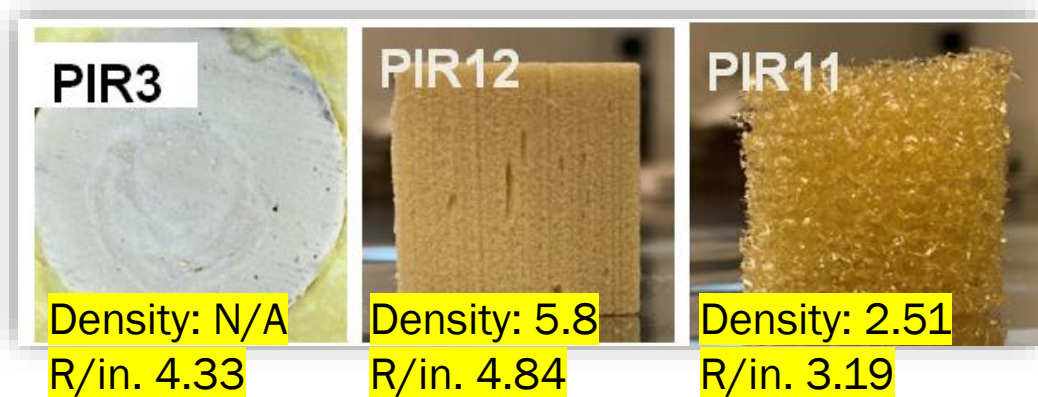
Isocyanate Polyisocyanurate

Simulation results indicate polyiso foam's R-value can be increased to R10/in.

# Progress: Experimental Challenges

Density: lb/ft<sup>3</sup>  
R/in: h.ft<sup>2</sup>.F/Btu/in.

- Too much blowing agent → too brittle, large pores



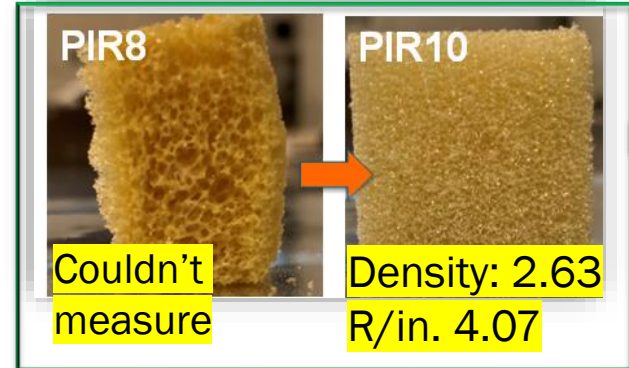
Poor structure & thermal performance →

- Excessive Isocyanate → poor thermal performance

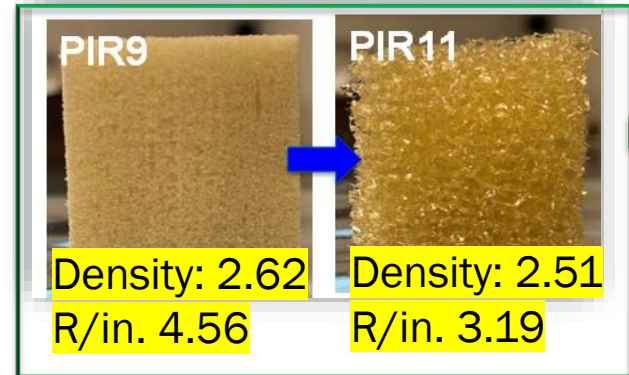
	NPIR7	NPIR14
pMDI	Iso Index ~380	Iso Index ~270
Density	2.94	3.0
R/in.	6.28	6.69

Formulation → control cellular structure → use the results for ML

- Same additive behaves opposite on different formulation



More additives → improved structure & thermal performance



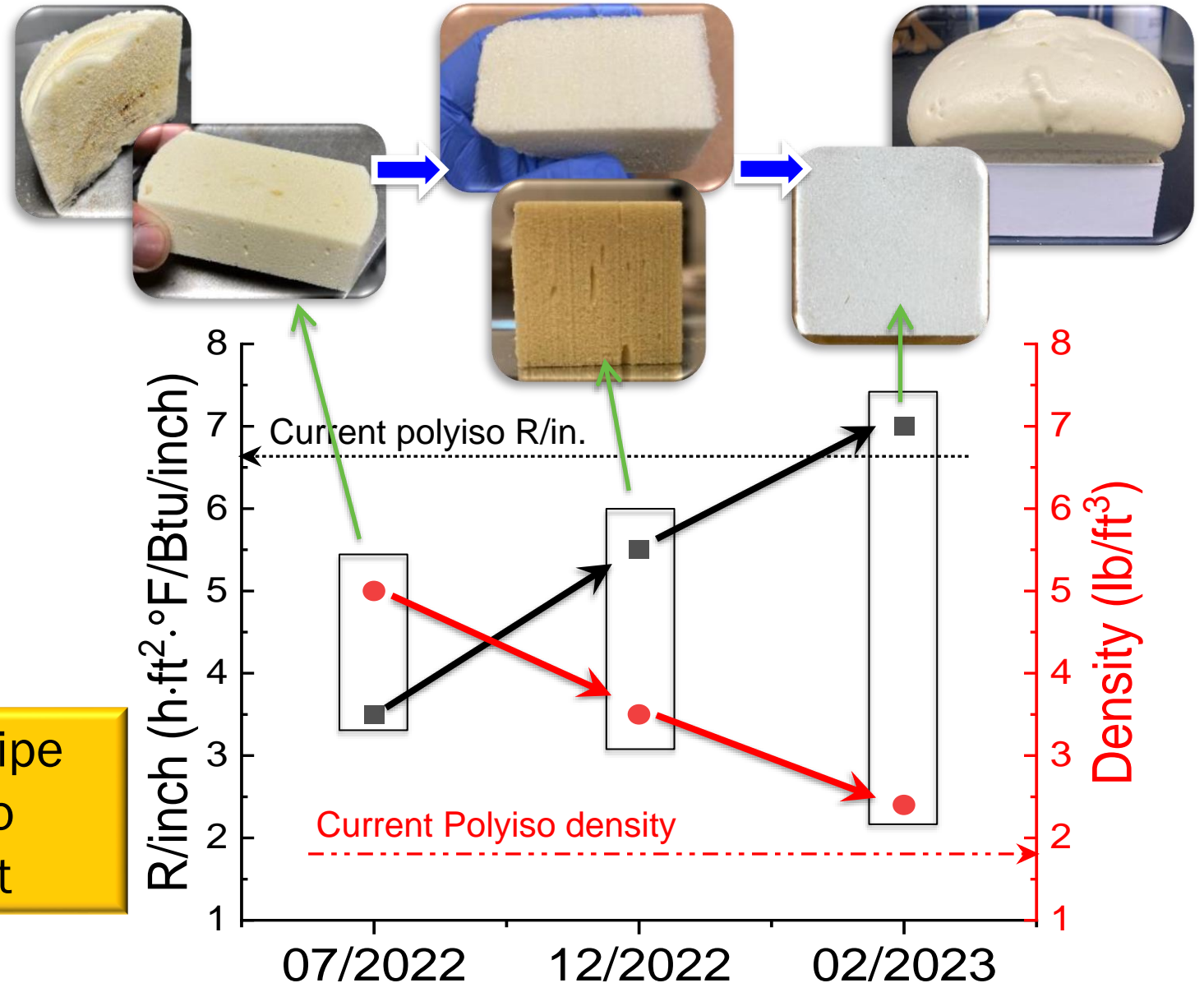
More additives → poor structure & thermal performance

# Progress: Experimental Obstacles and Significant Progress

## Formulation

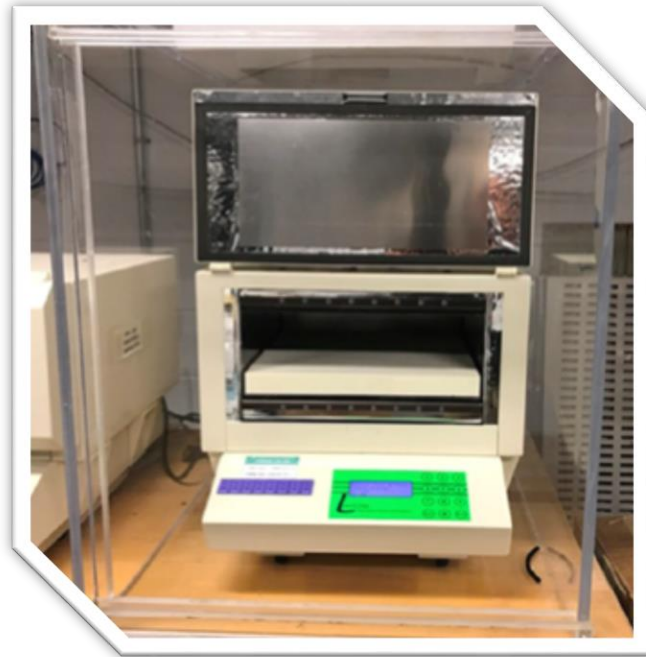
Material	
A-side	
Iso Index	270
B-side	Parts per weight (PBW)
Polyol	100
Catalysts, water, and other additives	varies

- Significant improvement in our recipe
- Wide range of experimental data to feed for ML algorithm development



# Progress: Thermal Performance and Morphology Analysis for ML

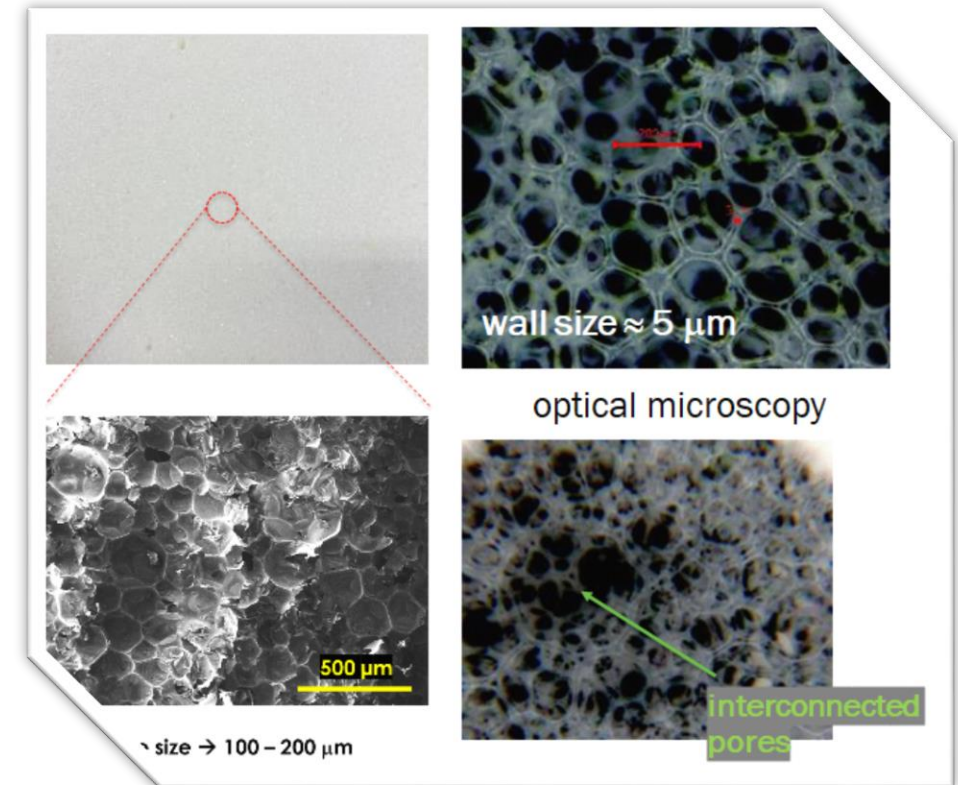
- Heat flow meter  
(thermal conductivity)  
ASTM C518



- Pycnometer  
(closed-cell contents)  
ASTM D 6226

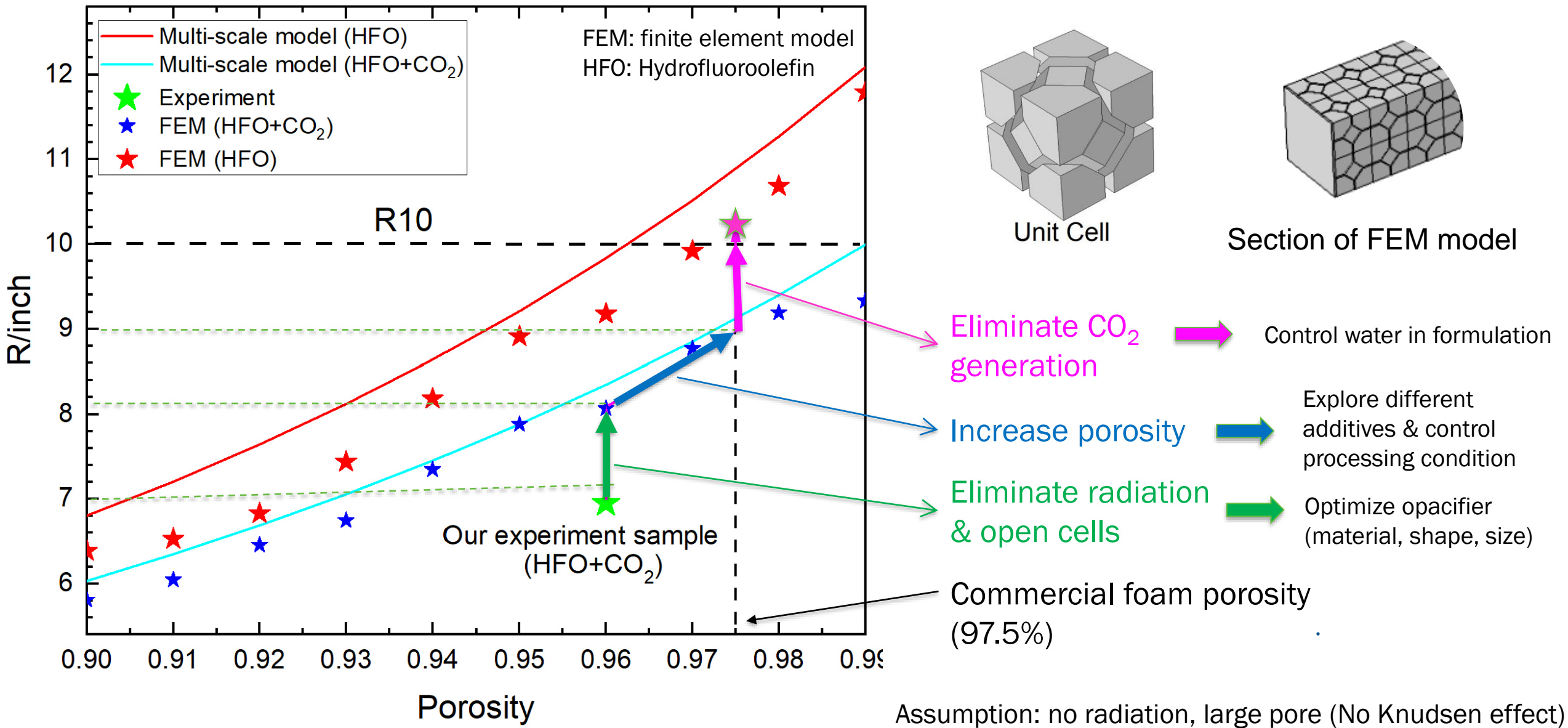


- Microscopy  
(morphology investigation)



Collecting important performance parameters to feed to ML algorithm

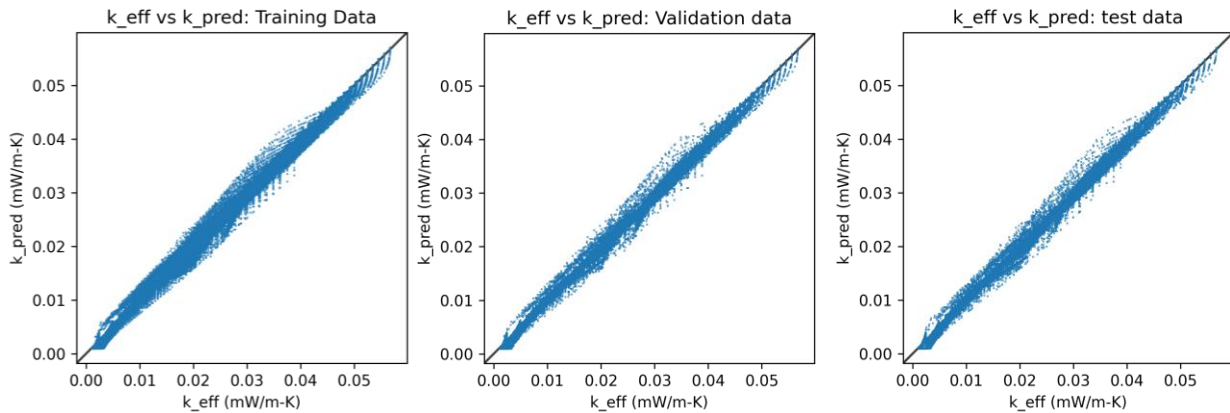
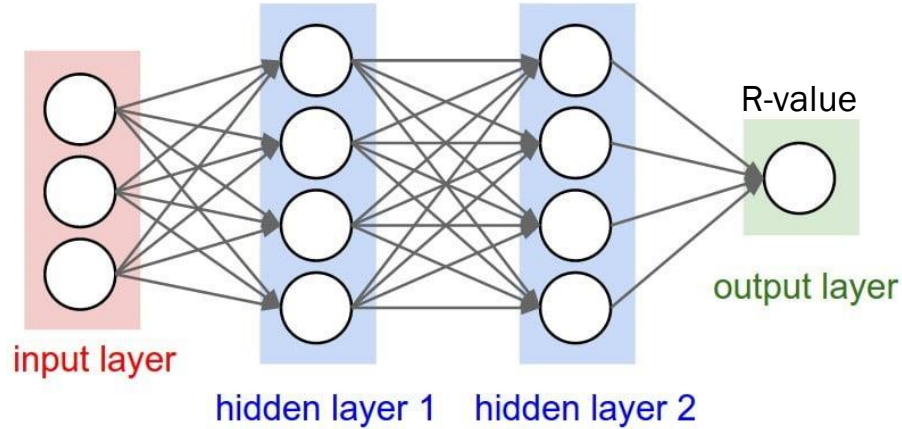
# Next Experimental Steps Guided by Simulations to Reach R-10/in.



# Next modelling effort: Machine learning

## R-value prediction by foam structures

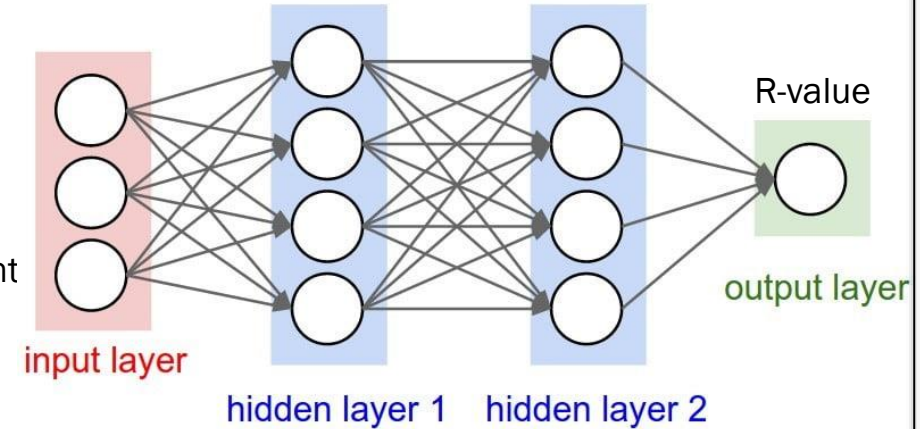
Porosity  
Pore size  
Gas composition  
Solid  
Anisotropic ratio  
.....



Our ML model predicts very well thermal conductivity

## R-value prediction by recipe

Materials  
Temperature  
Mixing time  
Catalyst  
Polyol  
Flame retardant  
Surfactant  
Opacifier  
.....



- ML model is built and ready
- As soon as we have more experimental data, we will be able to train the model and optimize the recipe



# Future Work

- Optimize the manufacturing conditions for a minimum foam thermal conductivity by using the trained ML model
- Optimize the porous structures to achieve  $\geq R10/\text{in.}$  by multiscale simulations, FE analysis, and ML
- Fabricate the ML-optimized foam prototypes and characterize the thermal conductivity with a target of  $>R10/\text{in.}$
- Identify a method for mass production of the new foam and minimize the cost

# Thank you

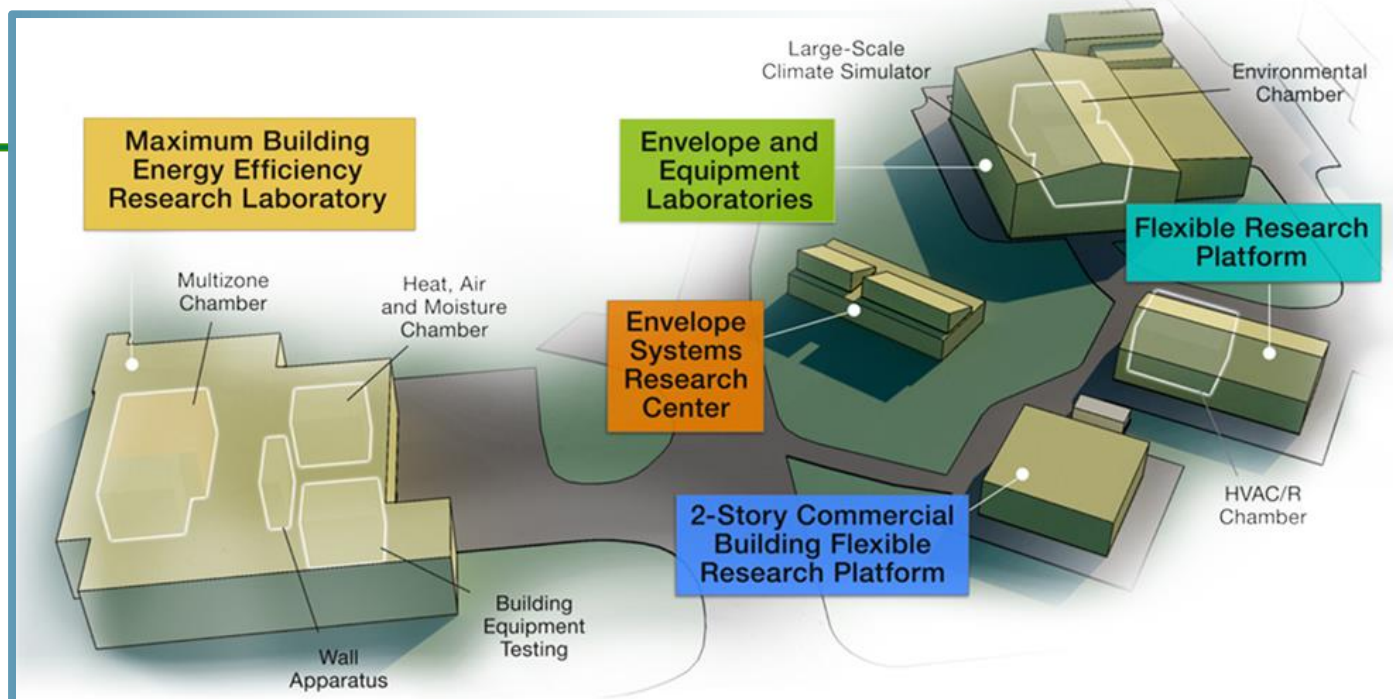
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Bokyung Park, Postdoctoral Research Associate

WBS # BTO-03.01.03.124, BENEFIT FOA Project # 2196-1726



**ORNL's Building Technologies Research and Integration Center (BTRIC)** has supported DOE BTO since 1993. BTRIC is comprised of 60,000+ ft<sup>2</sup> of lab facilities conducting RD&D to support the DOE mission to equitably transition America to a carbon pollution-free electricity sector by 2035 and carbon free economy by 2050.

## Scientific and Economic Results

236 publications in FY22

125 industry partners

54 university partners

13 R&D 100 awards

52 active CRADAs

***BTRIC is a  
DOE-Designated  
National User Facility***

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# REFERENCE SLIDES

# Project Execution

	FY2022	FY2023	FY2024
Planned budget	500,000	600,000	500,000
Spent budget	316,000	417,000	NA

No	Deliverable/Milestones	Year 1				Year 2				Year 3				
		Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	
Task 1. Finalize Overall Project Management and Planning														
1.1	Finalized Overall PMP	■												
Task 2. Collect and analyze the experimental data														
2.1	Collect the experimental details and TC for at least 20 foam designs				■									
2.2	Process and analyze at least 20 foam design data.				■									
Task 3. Optimize the Manufacturing Conditions for a Minimum Foam Thermal Conductivity														
3.1	Optimize the manufacturing conditions to minimize the foam thermal conductivity by using the existing manufacturing methods				■									
Task 4. Optimize the Porous Structures to Achieve $\geq R10/in.$ by Multiscale Simulations														
4.1	Obtain foam morphologies to achieve $\geq R10/in.$				■									
Task 5. Obtain the Foam's Cell Morphology Details Needed for Performing Machine Learning Optimization														
5.1	Obtain the foam's cell morphology details for the at least 20 foam designs					■								
Task 6. Optimize the Porous Structures to Achieve $\geq R10/in.$ by Multiscale Simulations, FE Analysis, and ML														
6.1	Train ML algorithm and predict the TC and foam structure with 90% accuracy							■						
6.2	Obtain the TC of porous structures by using multiscale models and FE simulations.								■					
6.3	Obtain at least three optimized porous structures with $\geq R10/in.$ by ML									■				
Task 7. Conduct a cost of goods sold (COGS) analysis to understand the cost distribution of foam manufacturing														
7.1	Collect cost details to conduct cost of goods sold (COGS) analysis								■					
7.2	Complete the COGS analysis and rank-order cost objects. Identify cost reduction opportunities to attain the cost of polyiso foam.									■				
Task 8. Identify new gas blends with a lower TC than currently industry used ones														
8.1	Obtain the pressure-dependent TC of CO <sub>2</sub> + HCFO blends								■					
8.2	Identify new gas blends with lower TC and condensation point of $< -40$ °C									■				
Task 9. Develop a method to fabricate foams with small pore size and high porosity														
9.1	Produce at least two types of foams with pore size $\leq 0.5$ $\mu m$ and porosity $\geq 0.9$ .										■			
Task 10. Fabricate the ML optimized foam prototypes and characterize the TC with a target of $> R10/in.$														
10.1	Produce at least two prototype foams that achieve $\geq R10/in.$											■		
10.2	Produce closed-cell foams with facers to achieve $\geq R10/in.$												■	
10.3	Fabricate at least one 6x6x1 in. foam that achieves $\geq R10/in.$													■
Task 11. Identify a method for mass production of the new foam and minimize the cost														
11.1	Identify options to modify the current commercial foam manufacturing processes that will be needed to produce new insulation that will potentially achieve $\geq R10/in.$												■	
11.2	Identify and trail a large-scale manufacturing process that will enable producing new insulation with $\geq R10/in.$ with a minimum modification in the current manufacturing													■

■ Milestone ■ Go/No-Go

# Team member



## Thermal Analysis

**Som Shrestha, Diana Hun,  
Andre Desjarlais**

Buildings Envelope Materials Research, ORNL

## Design / Synthesis of Foam

**Tomonori Saito, Sungjin Kim**

Chemical Sciences Division, ORNL

**Bokyung Park**

Buildings Envelope Materials Research, ORNL

## Physical Characterization

**Catalin Gainaru**

Chemical Sciences Division, ORNL

## Industry Partner

The GAF logo, consisting of the letters "GAF" in white on a red square background.

**Kevin McGrath,  
Kenneth Willoughby, Kevin Pollack**  
GAF Materials Corporation



## Simulation / Computational Analysis

**Tianli Feng, Janak Tiwari**

University of Utah

**Daniel Howard**

Buildings Envelope Materials Research, ORNL