

# Neural Network-based Isogeometric Topology Optimization of Multi-material Structures under Thermal and Mechanical Loads

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## Introduction

The increasing complexity of engineering problems, especially in structural design, has led to the development of advanced optimization methods aimed at improving the performance of multi-material structures under various loading conditions. Topology Optimization (TO) is a powerful technique that allows for the design of structures with optimal material distribution to achieve desired mechanical, thermal, and other performance criteria. In recent years, Isogeometric Analysis (IGA) has gained popularity as a method for integrating design and analysis, bridging the gap between Computer-Aided Design (CAD) and Finite Element Analysis (FEA). Combining isogeometric analysis with topology optimization presents a promising approach for designing multi-material structures, especially when subject to complex thermal and mechanical loadings. The use of Neural Networks (NNs) further enhances this optimization process by improving efficiency and enabling real-time design adjustments. This report explores the integration of these advanced techniques and presents a numerical investigation into the isogeometric topology optimization of multi-material structures under thermal and mechanical loadings using neural networks. The fundamental goal of topology optimization is to find the optimal material distribution within a given design space, subjected to various constraints and loading conditions. In the case of multi-material structures, the design involves the distribution of several materials in such a way that the structure exhibits the best overall performance under mechanical and thermal stresses. The mechanical loadings could include forces, moments, or displacements, while thermal loadings often involve temperature gradients that lead to thermal expansion, contraction, and stress generation within the material. For complex engineering systems, the coupling of these two types of loadings—thermal and mechanical—poses additional challenges, making it necessary to employ sophisticated numerical methods to handle the coupled physics and optimize the design effectively.

## Description

Isogeometric analysis, introduced by Hughes et al., is a numerical method that integrates CAD and FEA. Unlike traditional methods, which rely on discretizing the design domain into simpler elements, isogeometric analysis utilizes the same basis functions used in CAD, such as B-splines or NURBS (Non-Uniform Rational B-Splines). This allows for a more direct and accurate representation of the geometry, which is particularly beneficial in topology optimization problems where preserving the geometry's fidelity is crucial. By employing isogeometric analysis in topology optimization, it is possible to maintain a high level of geometric accuracy while still solving the optimization problem efficiently. In multi-material structures, the accurate representation of the interfaces between different materials and the overall geometry is essential to achieve an optimal solution. The coupling of thermal and mechanical loadings

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in topology optimization introduces additional complexity to the problem. Thermal loads can cause expansion or contraction of materials, leading to internal stresses that affect the mechanical performance of the structure. For example, in structures exposed to varying temperatures, the difference in thermal expansion between materials can lead to warping, delamination, or other forms of failure. Therefore, the optimization process must account for the effects of both mechanical and thermal stresses, ensuring that the final design is capable of withstanding both types of loads simultaneously. The thermal-mechanical coupling requires solving a set of coupled partial differential equations that describe the thermal and mechanical behavior of the structure. This can be computationally intensive, especially in large-scale problems, making it important to develop efficient optimization algorithms [1].

The introduction of neural networks into the optimization process represents a significant advancement in the field of structural design. Neural networks, particularly deep learning models, have shown great potential in learning complex patterns and relationships in data, making them well-suited for optimization tasks. In the context of topology optimization, neural networks can be used in several ways. One common approach is to use neural networks to approximate the mapping between design variables (such as material distribution, geometry, and loading conditions) and the objective function (such as compliance, stress, or temperature). By training a neural network on a set of known design examples, it can then predict the objective function for new designs, significantly reducing the computational cost associated with the optimization process. In the case of multi-material structures under thermal and mechanical loadings, neural networks can be particularly useful in predicting the structural response to thermal and mechanical loads, allowing for real-time adjustments and more efficient optimization. Instead of solving the full set of governing equations (which can be computationally expensive), the neural network can learn the relationship between the design parameters and the structural response, enabling rapid evaluations of different design configurations. This approach not only speeds up the optimization process but also allows for the handling of more complex problems that would otherwise be too computationally expensive to solve directly [2].

The process of isogeometric topology optimization of multi-material structures typically involves several steps. First, the design space is discretized using an isogeometric representation, typically using B-splines or NURBS to describe the geometry of the structure. The structure's mechanical and thermal properties are then defined, including Young's modulus, Poisson's ratio, thermal conductivity, and thermal expansion coefficients. The mechanical and thermal loads are applied, and the corresponding boundary conditions are set. Next, the optimization problem is formulated, with an objective function typically focused on minimizing compliance (for mechanical performance) or maximizing heat dissipation (for thermal performance) while satisfying constraints on stress, temperature, and other factors. The optimization is performed by iterating over the design variables, adjusting the material distribution and geometry to find the optimal solution. Neural networks can be integrated into this process in a few different ways. One common approach is to use neural networks to approximate the material distribution and geometry at each iteration of the optimization process. The neural network can be trained using data from previous iterations, allowing it to predict the optimal material distribution based on the current state of the design. This approach can reduce the number of iterations needed for optimization, speeding up the overall process. Another approach is to use neural networks for surrogate modeling, where the neural network approximates the objective function (such as compliance or thermal response) based on the design variables. This surrogate model can be used

to guide the optimization process, reducing the need for computationally expensive simulations.

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## Conclusion

The benefits of using neural networks in isogeometric topology optimization are clear. Neural networks can dramatically speed up the optimization process by reducing the need for repeated simulations. They can also handle highly complex design spaces, including multi-material structures with thermal and mechanical coupling, which would otherwise be difficult to optimize using traditional methods. Additionally, neural networks can adapt to different problem types, making them highly versatile for a wide range of engineering applications. However, the successful integration of neural networks into topology optimization requires careful consideration of the network architecture, training data, and optimization algorithm. The accuracy of the neural network's predictions is critical, and insufficient training data or poorly chosen network architectures can lead to suboptimal solutions. In conclusion, the integration of isogeometric analysis, topology optimization, and neural networks offers a powerful approach for designing multi-material structures under thermal and mechanical loadings. This combination allows for highly efficient, accurate, and real-time optimization of complex structures, reducing computational costs and enabling the design of more efficient and robust systems. While challenges remain, particularly in terms of neural network training and integration, the potential benefits of this approach make it a promising direction for future research in structural optimization. As computational power increases and machine learning techniques continue to advance, it is likely that neural

network-based isogeometric topology optimization will play a significant role in the design of next-generation engineering structures.

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