

Materials for Strategic Sectors

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Data-Driven Ballistic Property Prediction of Ceramic Armor Materials for Advanced Protection Systems

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Introduction and Objective

Ceramic armor selection still depends heavily on firing trials and destructive testing, which slows innovation and fails to systematically explore composition–microstructure–performance relationships. This work develops a machine learning–driven prediction system integrating crystallographic, mechanical, and thermal descriptors to forecast ballistic-relevant properties, enabling rapid screening of candidate armor ceramics.

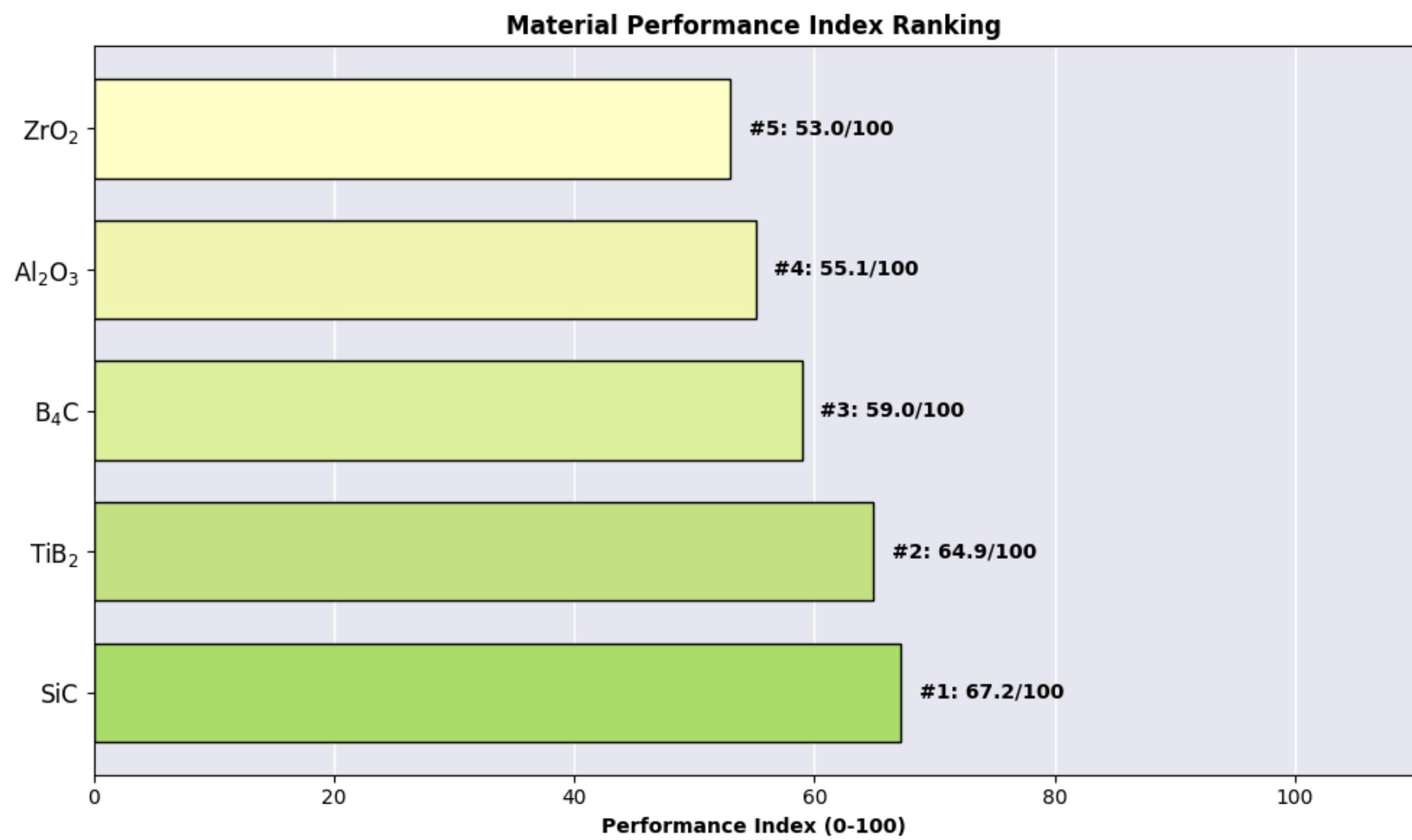
To develop an ML framework capable of:

- Predicting mechanical & ballistic performance of ceramics with high accuracy
- Optimizing ceramic compositions for lightweight, high-protection armor
- Reducing experimental screening effort for defense materials discovery

Methodology: The ML Pipeline

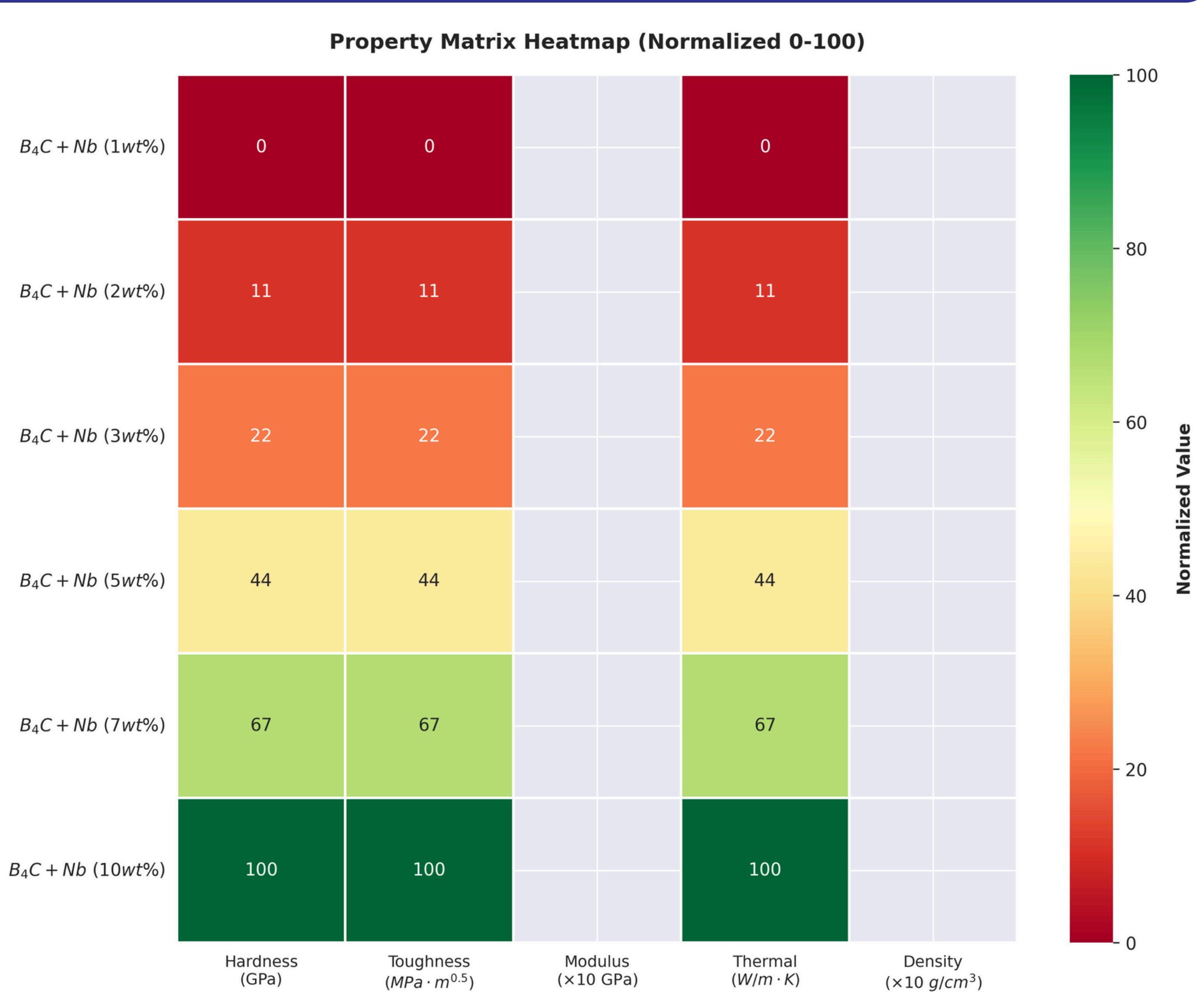
- Data Sources: Integrated repository combining the Materials Project, NIST Ceramics Database, and ballistic datasets for SiC, B₄C, WC, TiC, and Al₂O₃.
- Feature Engineering: Optimization performed on key ballistic indicators: Fracture Toughness, Hardness, Density, and Penetration Resistance.
- Algorithm Performance: The predictor achieves R² > 0.85 for mechanical properties and R²> 0.80 for ballistic indices.

Results & Discussions

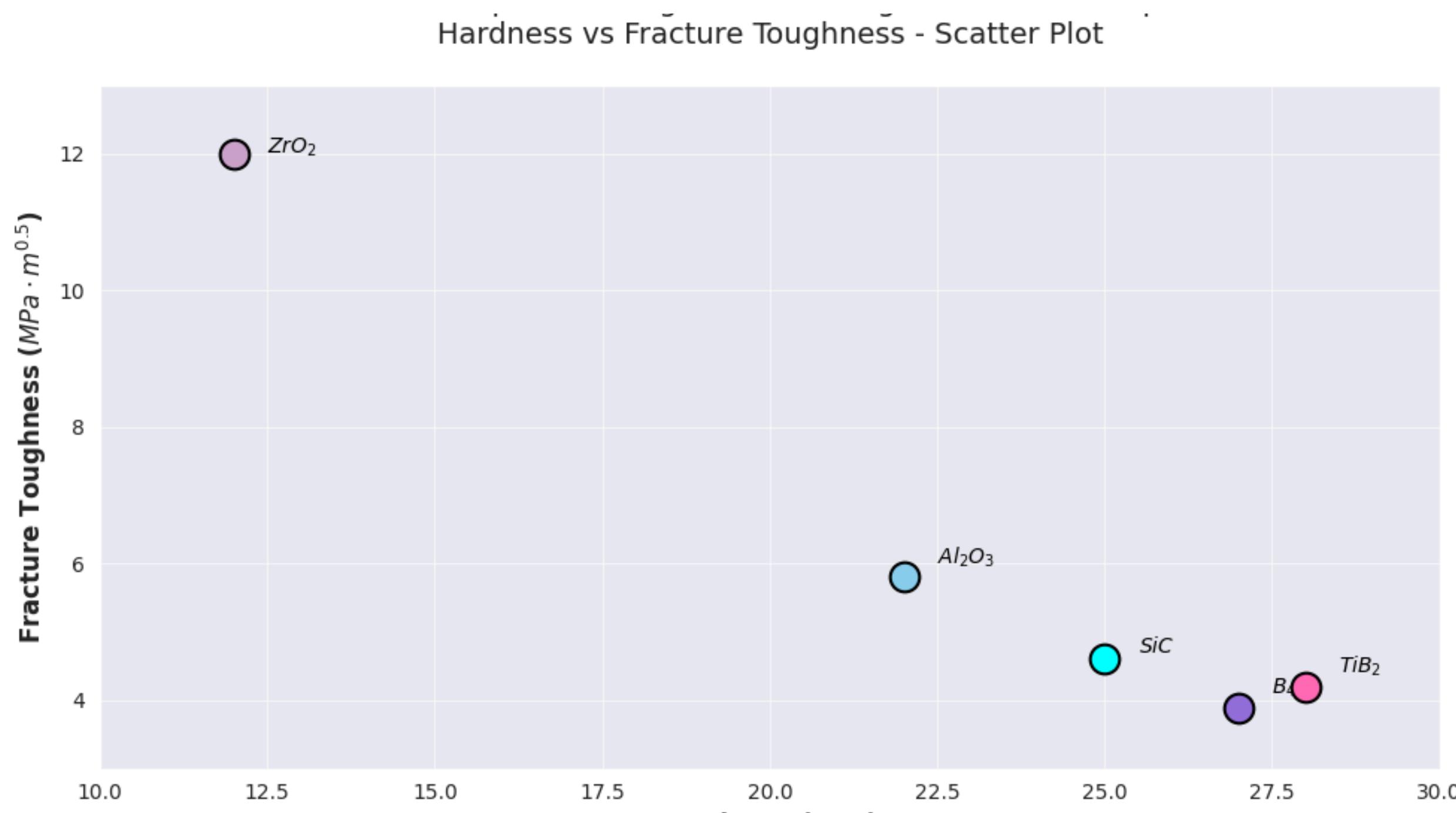
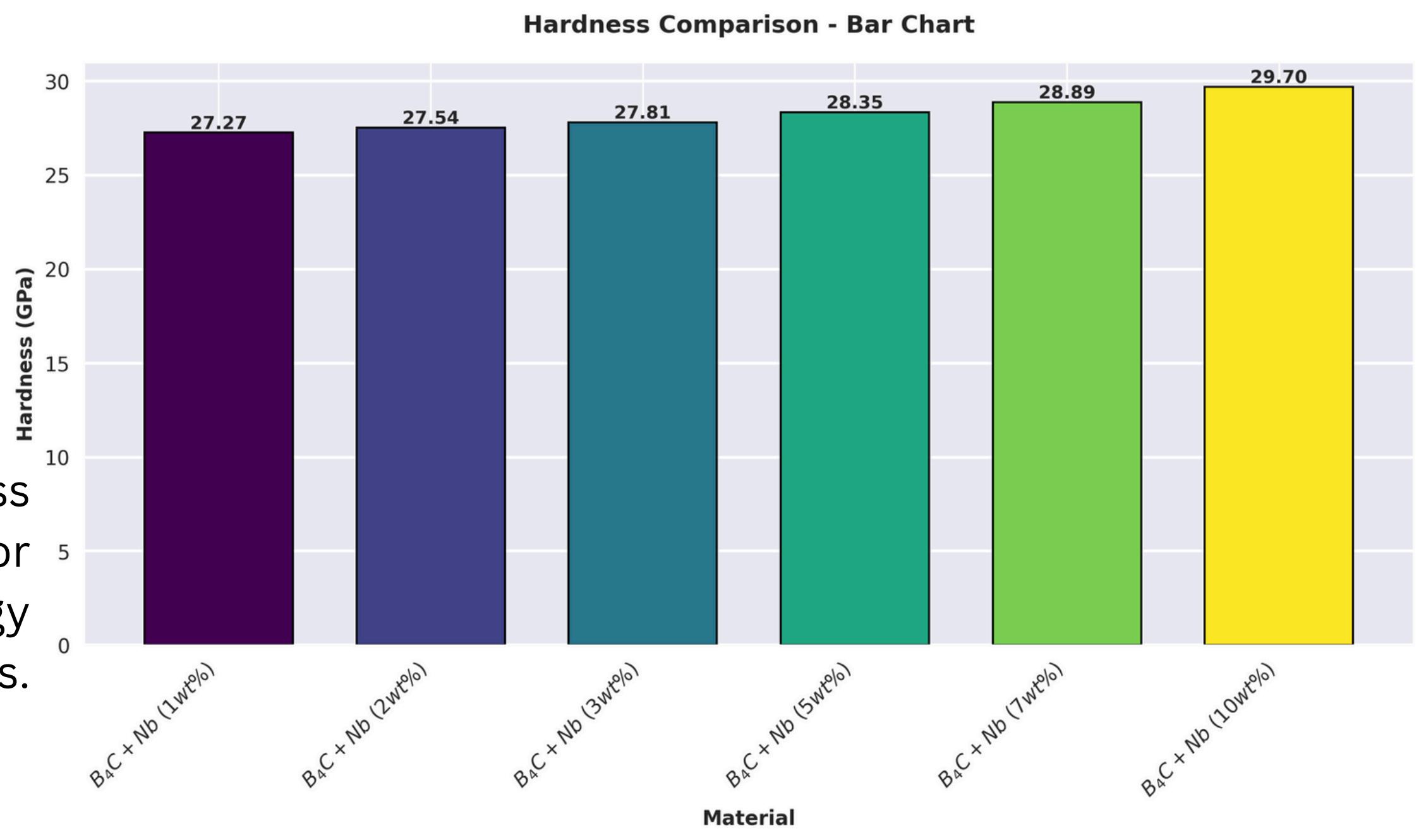


SiC achieves the highest composite score (**67.2/100**), offering the optimal balance of weight, thermal, and mechanical properties.

TiB₂ leads with a maximum hardness of **28.0 GPa**, providing superior resistance against kinetic energy penetrators.

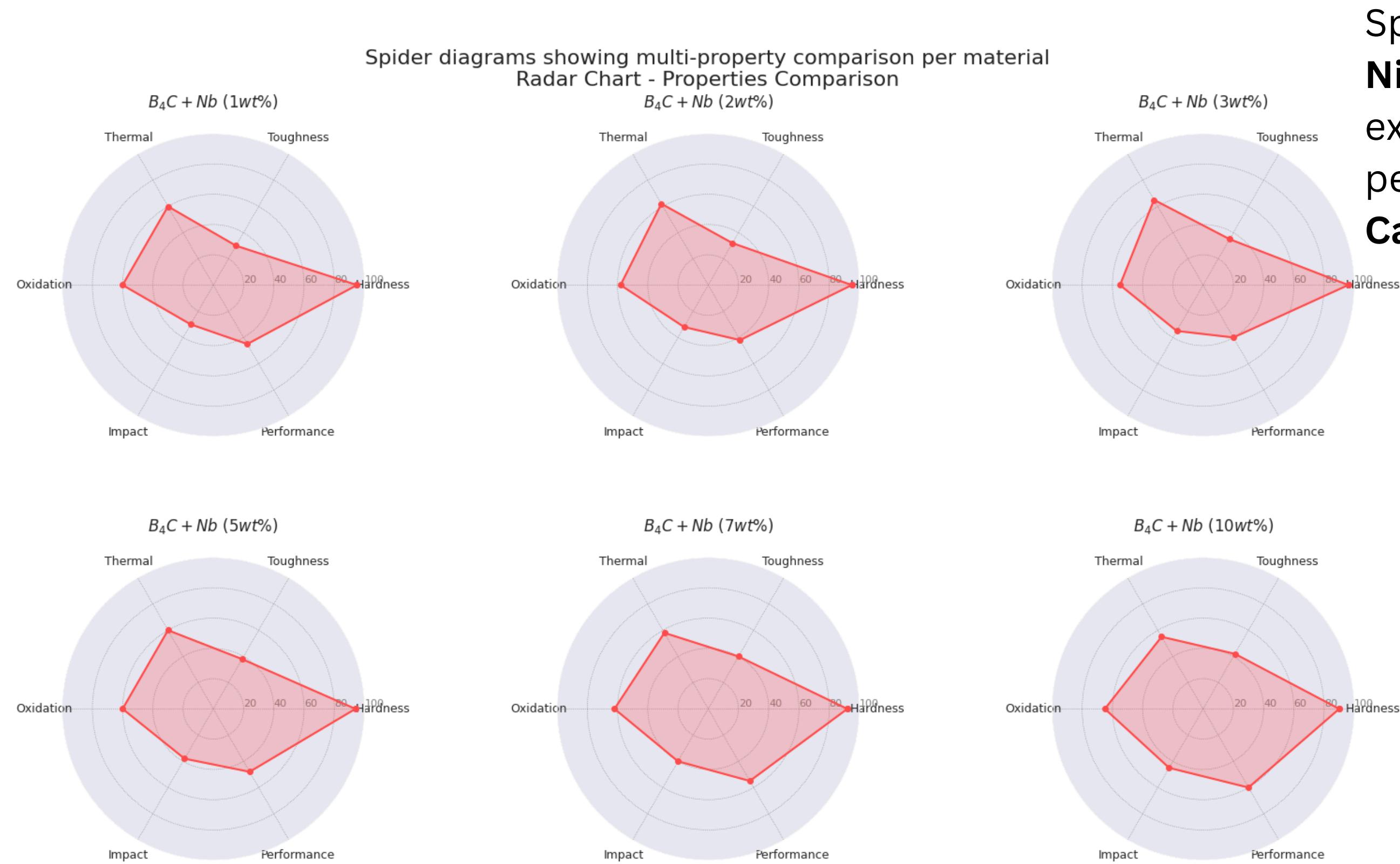
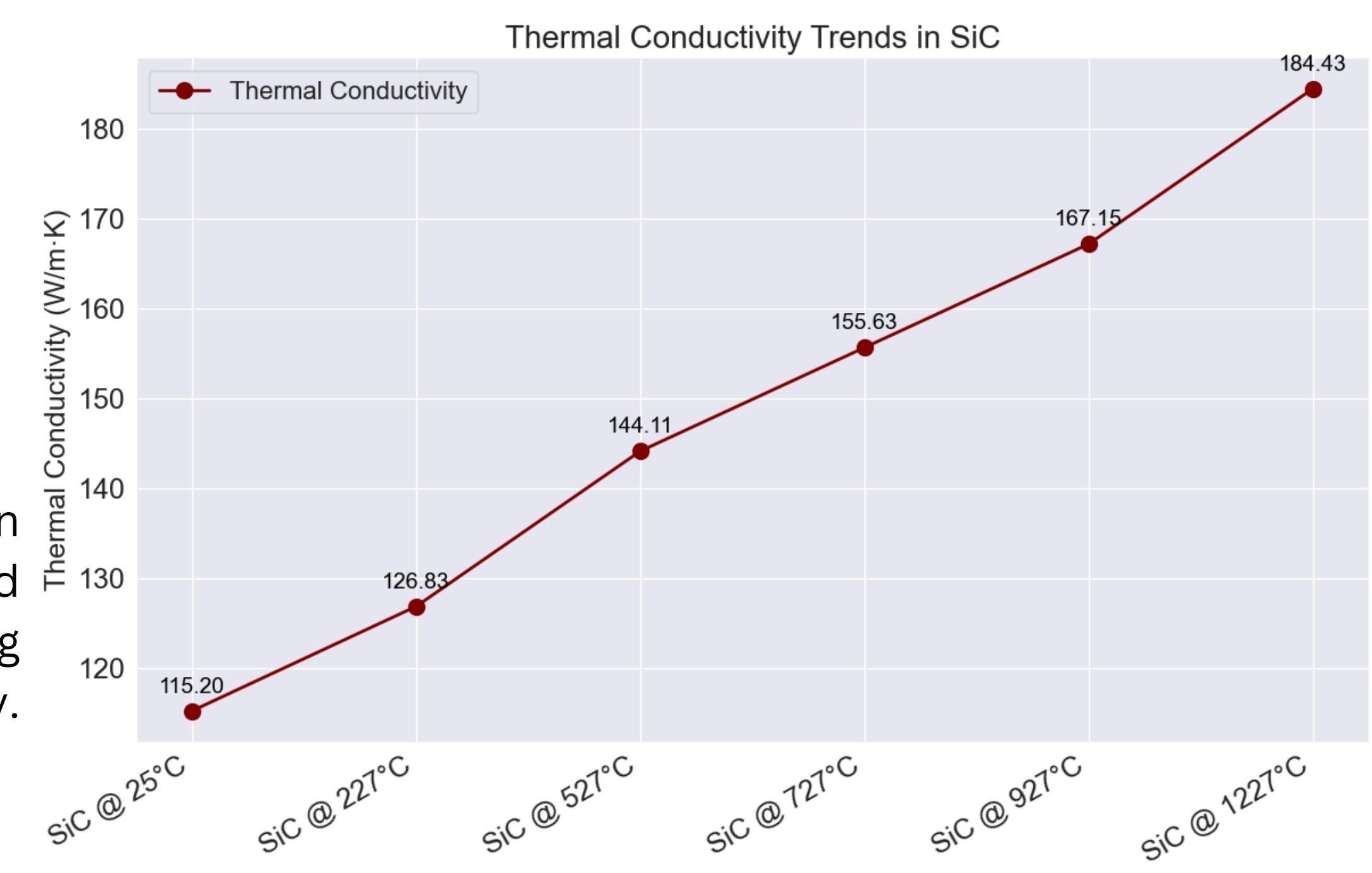


Normalized input dataset visualizing the property distribution across material classes used to train the ML predictor.



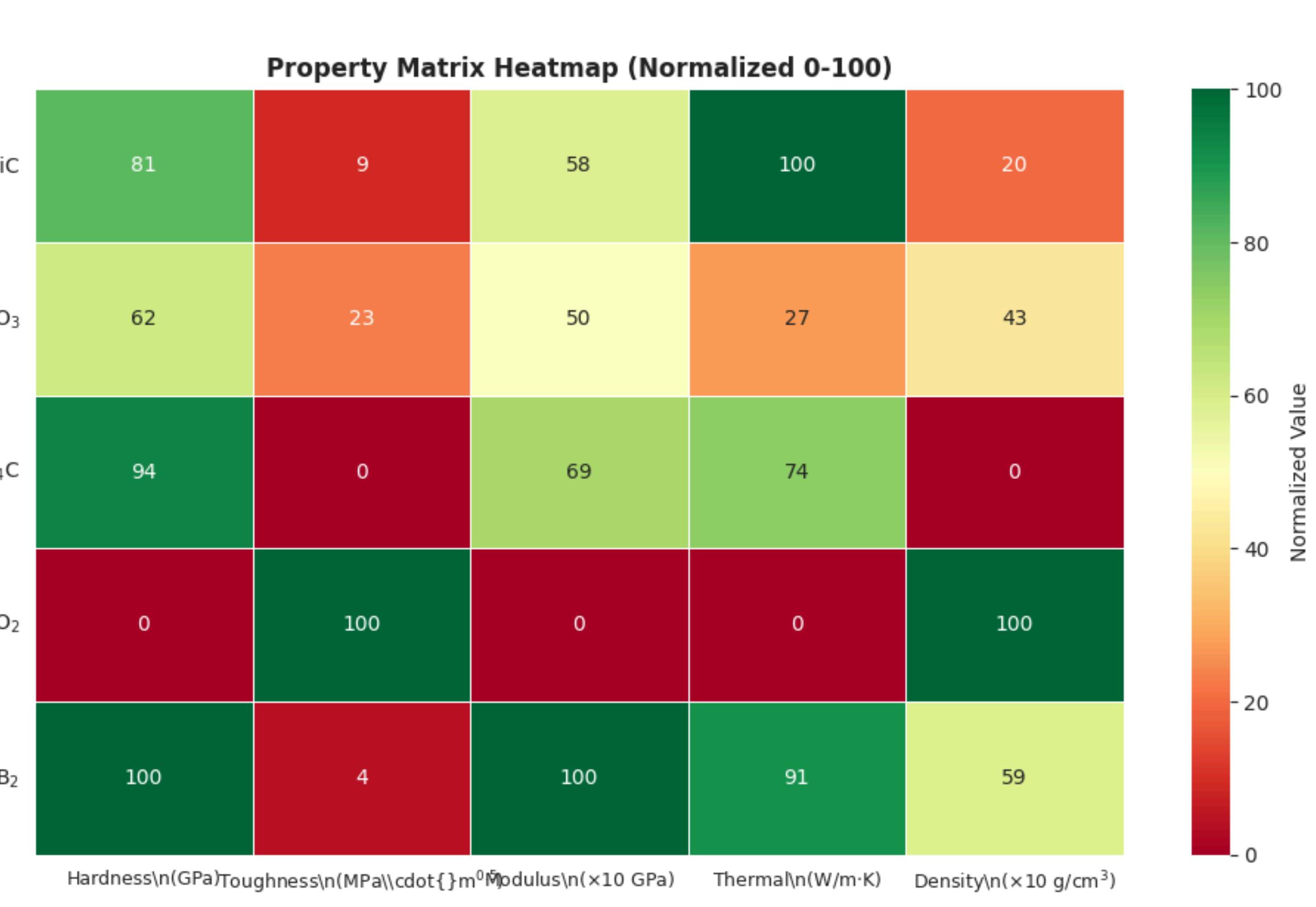
Illustrates the inverse relationship where **ZrO₂** maximizes toughness, whereas **SiC** and **TiB₂** prioritize ballistic hardness.

Line analysis showing the increase in **SiC** thermal conductivity at elevated temperatures, critical for dissipating impact energy.



Spider diagrams revealing how **Niobium** doping systematically expands the multi-property performance envelope of **Boron Carbide**.

Normalized matrix highlighting optimal property clusters, with green indicating superior performance across standardized metrics.



Conclusions

- ML achieves R²>0.80 ballistic; NSGA-II optimizes 3x faster.
- B₄C+Nb(10wt%): Top perf (61.9/100)—hardness 29.7 GPa, density 2.52 g/cm³.
- Framework extensible to composites for defense-energy sustainability.

