Quantum mechanics (QM) modeling enables accurate predictions of molecular interactions but is computationally expensive. Graph Neural Networks (GNNs) offer a faster alternative but struggle to generalize across diverse, unseen compounds. This work demonstrates that incorporating element features improves generalization to compounds with new elements. Our results show that excluding up to 10\% of elements from the dataset does not degrade performance.