

Solving Variable-Entangled Partial Differential Equations (PDEs) via Atomic Tensor Decomposition and Gated Expert Neural Networks: A Multi-Expansion Framework for Accurate and Adaptive PDE Approximation

Dhulipalla, Tejas (School: Paradise Valley High School)
Parashar, Maatank (School: Paradise Valley High School)

We propose a gated neural framework for solving partial differential equations (PDEs) of the form $L(u(x)) = f(x)$, where $x \in \mathbb{O} \subset \mathbb{R}^n$ and $u : \mathbb{O} \rightarrow \mathbb{R}$. The solution $u(x)$ is expressed as a residual-weighted sum of atomic separable expansions: $u(x) \approx \sum_{j=1}^S w_j(x) \times f_j(x)$, where each f_j captures univariate structure along dimension j . Atomic components are drawn from six novel analytically derived expansion families, each corresponding to a distinct structural mode of the PDE, including first-order, second-order, nonlinear coupling, fractional, eigenfunction, and PGD-based decompositions. For each atomic candidate f_j , we compute a pointwise residual $R_j(x) = \|L f_j - f(x)\|^2$. These residuals are passed to a gating network that assigns spatially adaptive weights using $w_j(x) = \exp(-R_j(x)) / \sum_{j=1}^S \exp(-R_j(x))$, enabling the model to prioritize expansions that best match the local operator behavior. The PDE system, including L , f , and boundary data g , is embedded as a structured feature vector and fused with spatial coordinates. This representation is processed through six expert subnetworks, each generating a candidate solution aligned with a specific expansion class. The aggregated output $\hat{u}(x)$ maintains operator consistency, admits SVD-based low-rank compression, and remains interpretable. Unlike monolithic solvers such as PINNs, our model operates in the Fréchet space $C^8(\mathbb{O})$, ensuring $\sup_{x \in \mathbb{O}} \|\partial^k \hat{u}(x) - \partial^k u(x)\| \leq \epsilon$ for $0 \leq k \leq N$ via a hybrid soft-penalty/hard-solve loss, preserving termwise differentiability. This yields reduced approximation error in regions with steep gradients, faster residual decay $\|L(\hat{u}) - f\| \approx 0$, and consistent recovery of operator-aligned solutions in systems where ∂u , u^2 , or mixed derivatives $\partial^2 u / \partial x^2 \partial y$ dominate the solution behavior.

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