

Conformal Defects in Neural Network Field Theories

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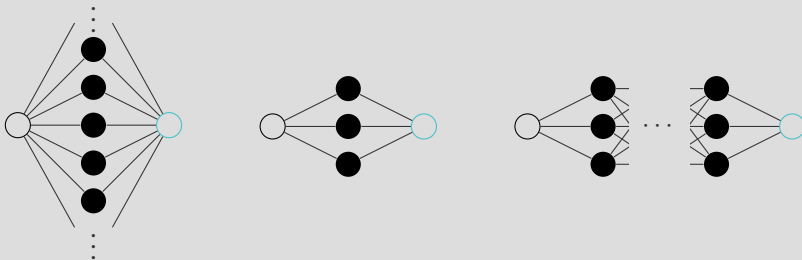
What is a neural-network QFT and why do we care?

1 Neural-Network CFT

Neural Network: a map $f : \mathbb{R}^m \rightarrow \mathbb{R}^n$

Operations: composition, addition, scalar multiplication

Universal Approximation Theorems (UATs): the NNs are dense in a space of maps (in some limit).



$\hookrightarrow$ Depends on choice of basis, and choice of infinite-node limit.

What is a neural-network QFT and why do we care?

1 Neural-Network CFT

QFT: Generating functional: $Z[J] = \int \mathcal{D}\phi e^{-S[\phi] + \int d^D x \phi(x) J(x)}.$

Key Idea: Approximate the functional integral using the UATs and NNs. [Halverson, Maiti and Stoner; 2008]

NNs come with a set of parameters: eg. Sigmoids = $\{\sigma(a \cdot x + b)\}.$

Integrate over those with a measure:

$$\int \mathcal{D}\phi e^{-S[\phi]} \longleftrightarrow \lim_{N \rightarrow \infty} \int \prod_i^N d\theta_i P(\theta_i) \quad (1)$$

In the limit where the UAT applies, both sides are equal.

Result: Novel way of constructing QFT correlators.

What about neural-network CFTs?

1 Neural-Network CFT

CFT: More symmetry $\rightarrow$ more constrained dynamics.

Full CFT data encoded in set of numbers (Δ_i, λ_{ij}) .

NN-CFT: Approximation is exact at any number of neurons! No need to take a UAT limit!

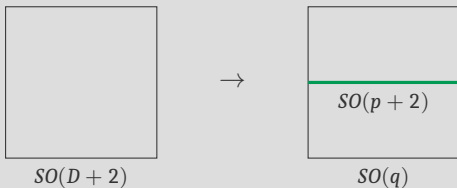
[Halverson, Naskar, Tian; 2024]

Conformal primary	
$\phi(x) \mapsto \lambda^{-\Delta} \phi(x)$	$\Phi(X) \mapsto \lambda^{-\Delta} \Phi(X)$
Partition function	
$Z[J] = \int \mathcal{D}\phi e^{-S[\phi]} e^{\int d^d x J(x) \phi(x)}$	$Z[J] = \int d\theta \mathcal{P}(\theta) e^{\int d^d x J(x) \Phi(x)}$
Correlators	
$\langle \phi(x) \phi(y) \rangle$	$\mathbb{E}[\Phi(x) \Phi(y)]$

Extending this to theories with defects

2 Neural-Network Defect CFT

Break the symmetry group: $SO(D + 2) \rightarrow SO(p + 2) \times SO(q)$



dcFT: Full data encoded in set of numbers $(\hat{\Delta}_i, \Delta_i, \lambda_{ij}, \lambda_{ijk})$

Result: Neural-network approximation is still exact at finite number of neurons! [Capuozzo, Robinson, Suzzoni; 2025]

We can build new defect CFTs simply by specifying an invariant distribution and conformal primaries fields!

Conclusion

3 Conclusion

What was known

1. Neural-Network CFTs can be engineered from a finite number of building blocks (neurons)
2. Correlators are well defined (standard statistical averages)
3. Can combine neurons to form new CFTs! → playground for generating CFT data

What we did

1. Extended the notion of nn-CFT to the defect case
2. Constructed simple examples of fully solvable nn-dCFTs

Formalism doesn't yet exist for $\text{spin} \neq 0$ $\hookrightarrow$ Important for condensed matter, string theory, AdS/CFT, ...

Thank you

dCFT correlator constraints

3 Conclusion

