Measuring the "size" of hypothesis spaces over $\mathbb R$ from the perspective of No Free Lunch

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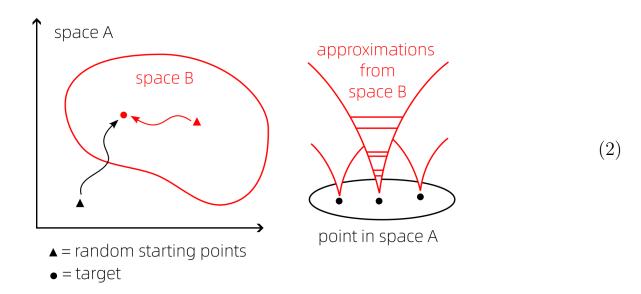
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Motivating example

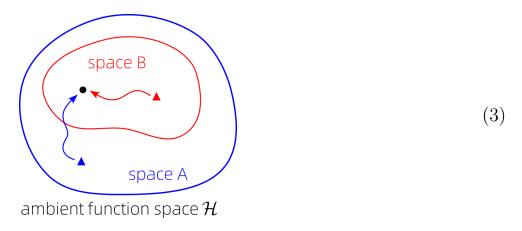


The **input space** is $X = \mathbb{R}^n = n$ -dimensional hypercube. Permutation symmetry implies that only one corner of the hypercube need be considered, this is the **fundamental domain**. The **volume** of this domain over the entire hypercube shrinks exponentially as n grows, which appears to suggest that this symmetry is very significant for efficiency consideration. We want to quantify the notion of this efficiency.

We want to compare the **hypothesis spaces** A and B: A = unconstrained, fully-connected neural network with L layers and N total weights. The parameter space $\Theta_A = \mathbb{R}^N$. B = symmetric neural network with a special structure but its parameter space is also $\Theta_B = \mathbb{R}^M$.



It's easier to consider A and B both embedding in an ambient function space \mathcal{H} :



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