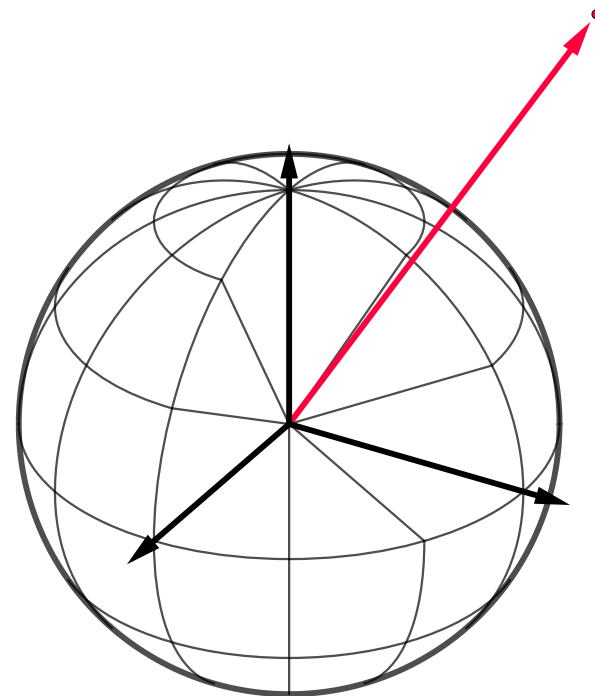


# Deep Networks on Toroids: Removing Symmetries Reveals the Structure of Flat Regions in the Landscape Geometry

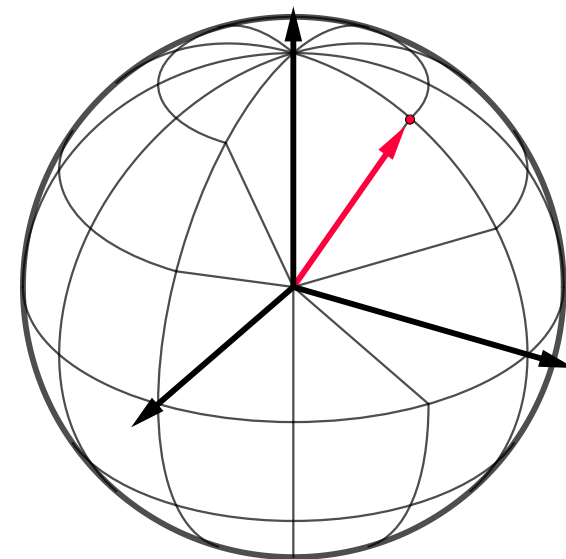
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Bocconi University, Milan, Italy

ICML 2022 - Short Presentation



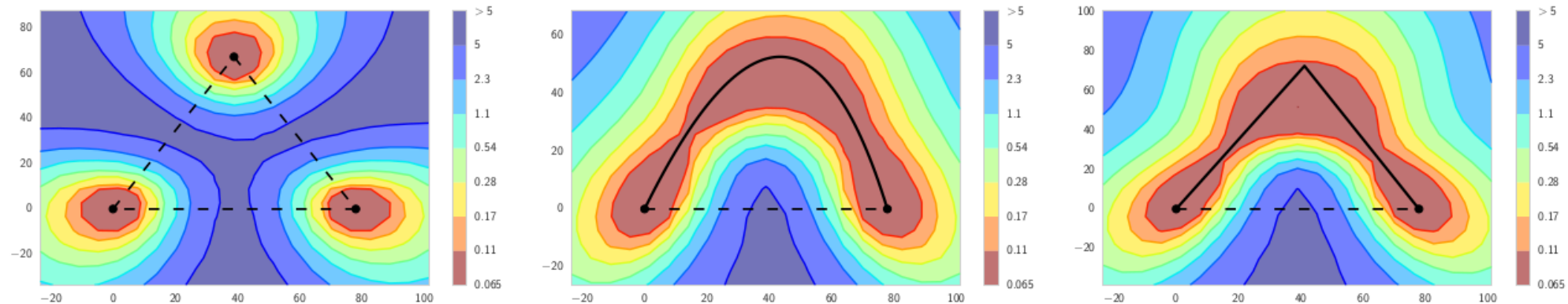
$w$



$\frac{w}{|w|}$

# Neural Network Landscapes

- can be explored in terms of the *loss* or *error* space
- determine the dynamics of gradient descent
- exhibit non-trivial symmetries
- points at the same height can have drastically different generalization properties, connected to **flatness**
- minima found independently can often be connected with relatively simple paths



Plot from Garipov, T., Izmailov, P., Podoprikin, D., Vetrov, D. P., & Wilson, A. G. (2018). *Loss surfaces, mode connectivity, and fast ensembling of dnns. Advances in neural information processing systems.*

## Our work

- studies the error landscape
- ... after removing symmetries
- ... around solutions found with different algorithms (entropic algorithms and standard algorithms)
- ... in networks with continuous and binary weights

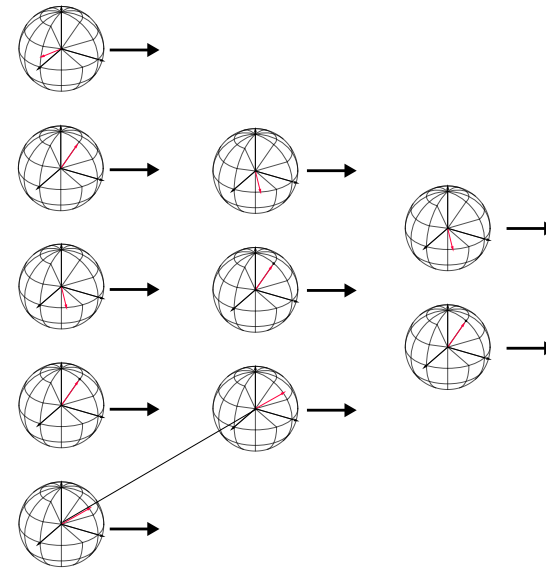
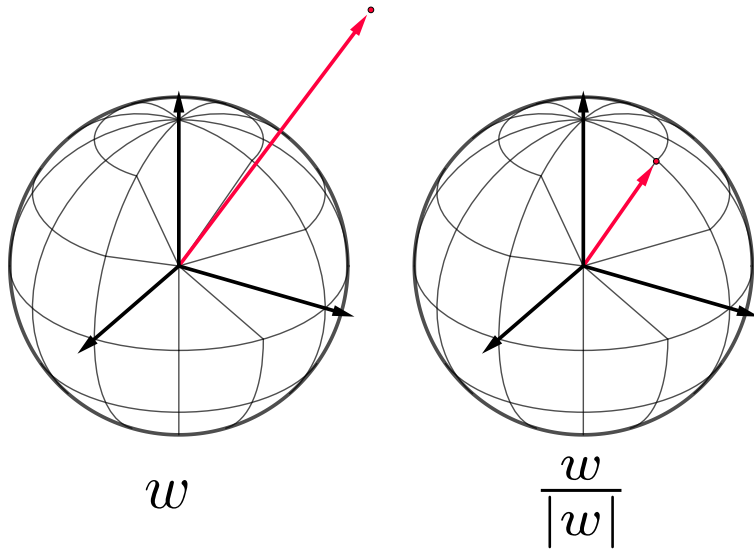
# Symmetries

## Weight-Norm

For the *ReLU* activation function we have  $f(\alpha x) = \alpha f(x)$  for  $\alpha \geq 0$  and therefore

$$f(\vec{w} \cdot \vec{x}) = |\vec{w}| f\left(\frac{\vec{w}}{|\vec{w}|} \cdot \vec{x}\right),$$

which means we can *push* the weight norms to the next layer.



**Every neuron becomes a hypersphere with norm=1**

**The complete network is a product of hyperspheres**

[The last layer precedes a `argmax` and since we are only interested in the error we can normalize it *globally* with a positive factor]

# Symmetries

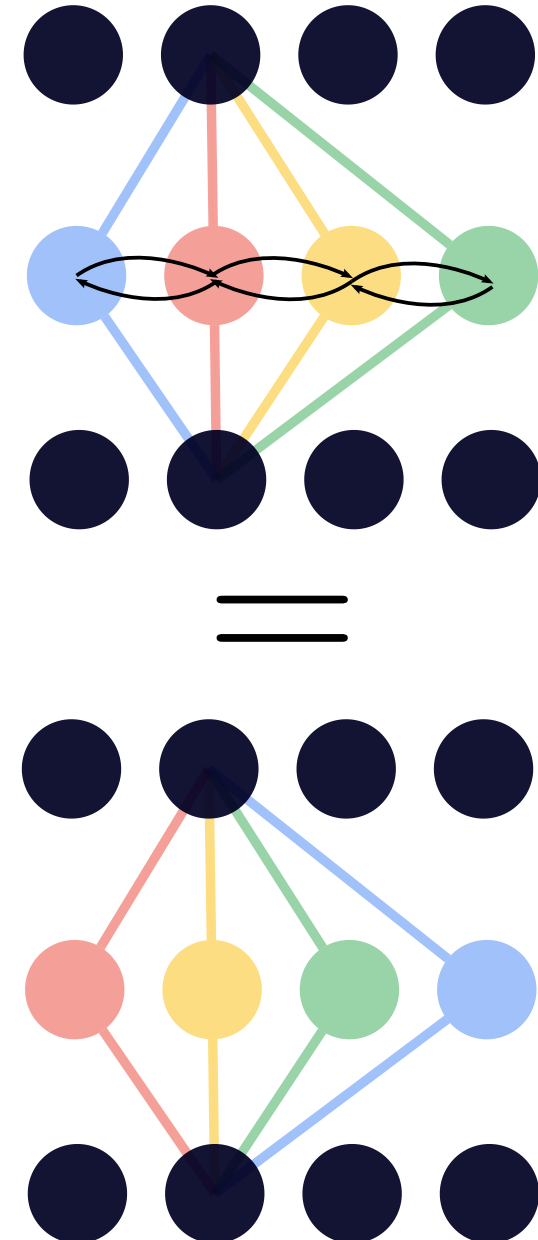
## Permutation

- Neurons within a layer can be exchanged as long as incoming and outgoing connections are taken care of
- The same goes for kernels in convolutional layers
- Networks need to be *aligned* before comparing them

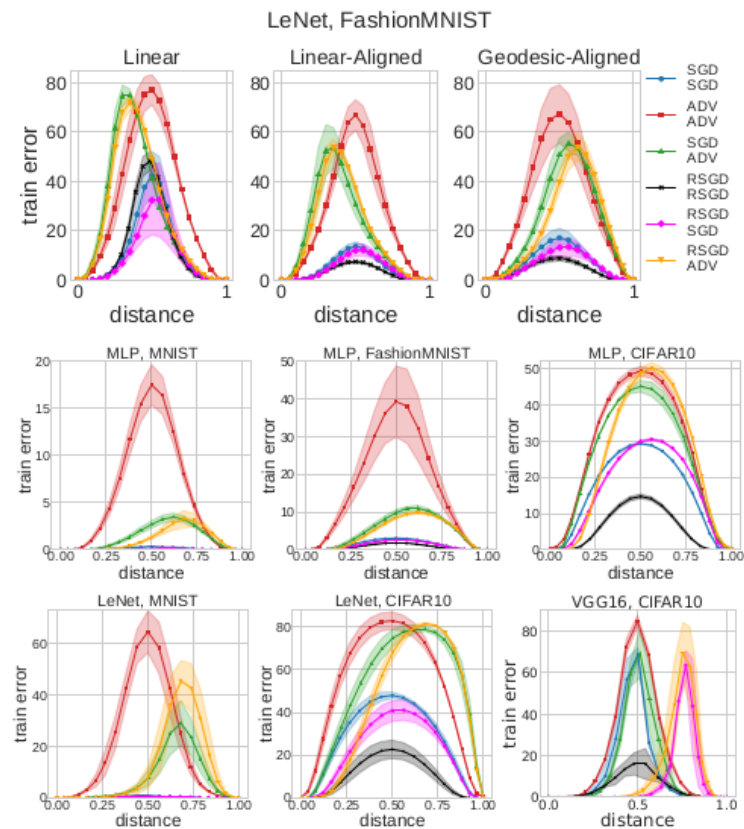
## Simple Algorithm for Matching

```
Input: Two normalized NNs with parameters A[1..L], B[1..L] and L layers
for l = 1 to L - 1 do
   $\pi$  = Match(A[l], B[l])
  PermutePrev(B[l],  $\pi$ )
  PermuteNext(B[l+1],  $\pi^{-1}$ )
end for
```

**Match** takes two sets of parameter vectors and solves a weighted bipartite graph matching problem using cosine similarities between vectors as weights.

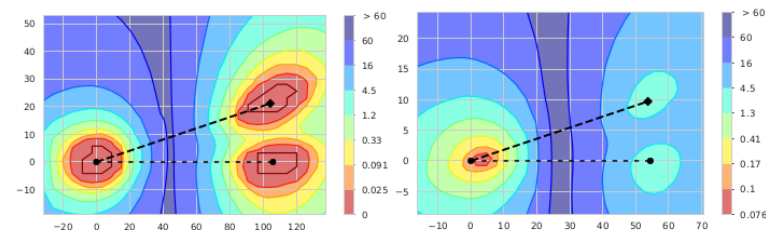


# Mode Connectivity



- SGD: Stochastic Gradient Descent
- RSGD: *Replicated SGD* (finds flatter minima)
- ADV: Adversarial Initialization
- Linear: Straight path between minima
- Geodesic: Path on normalized sub-manifold

## VGG16 on Cifar10



- Left Panel: Unnormalized
- Right Panel: Normalized
- Left Points: RSGD (finds flatter minima)
- Right Points: unaligned/aligned SGD with adversarial initialization

**Difference is only visible *after* symmetry removal**

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# **Deep Networks on Toroids: Removing Symmetries Reveals the Structure of Flat Regions in the Landscape Geometry**

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