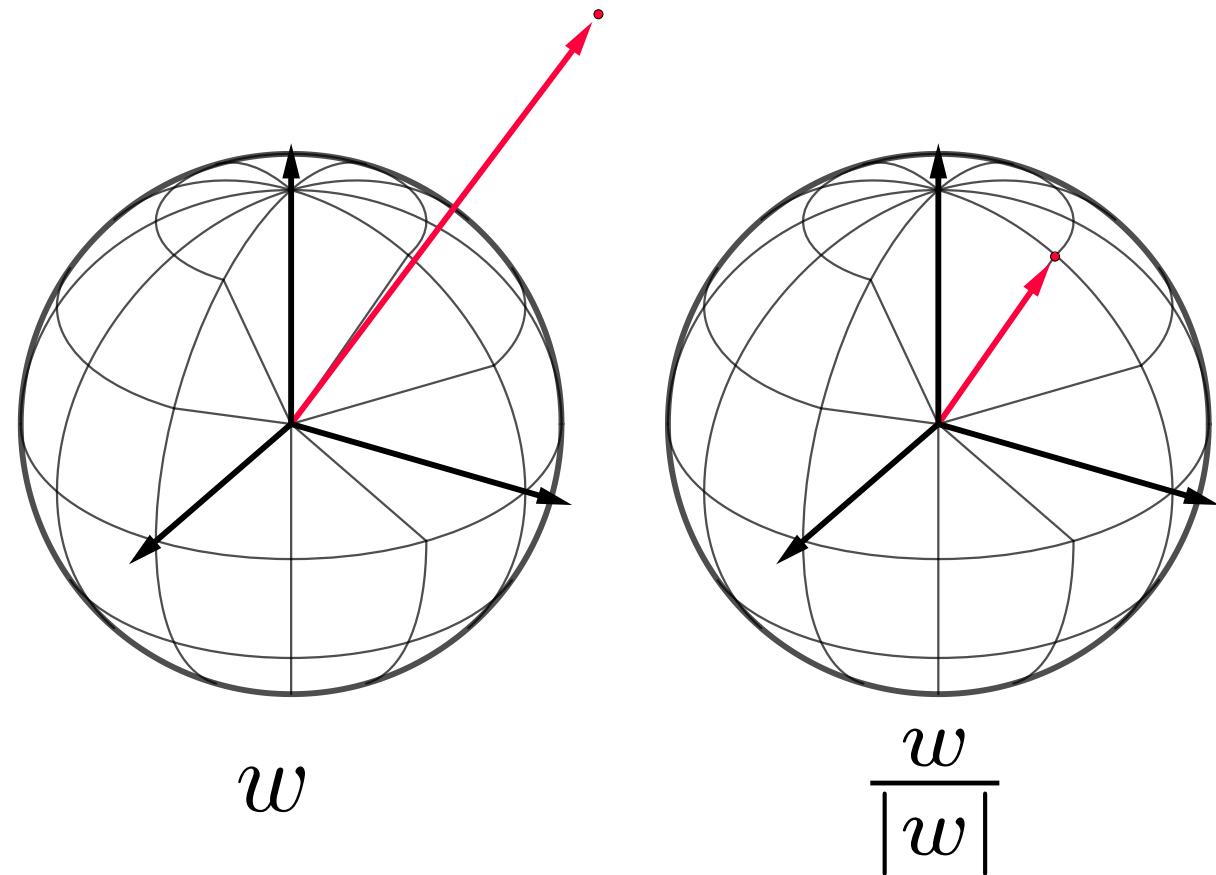


# Deep Networks on Toroids: Removing Symmetries Reveals the Structure of Flat Regions in the Landscape Geometry

*Fabrizio Pittorino, Antonio Ferraro, Gabriele Perugini,  
Christoph Feinauer, Carlo Baldassi, Riccardo Zecchina*

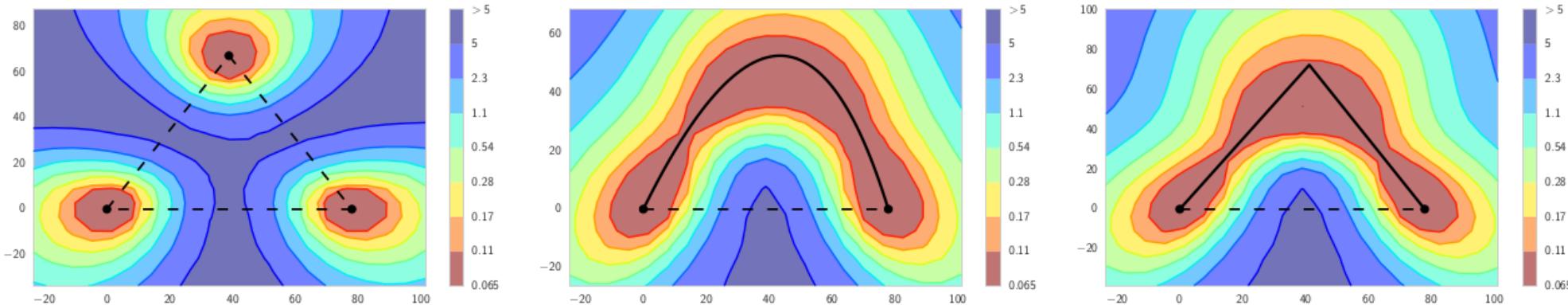
Bocconi University, Milan, Italy

ICML 2022 - Short Presentation



# 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., Podoprikhin, 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

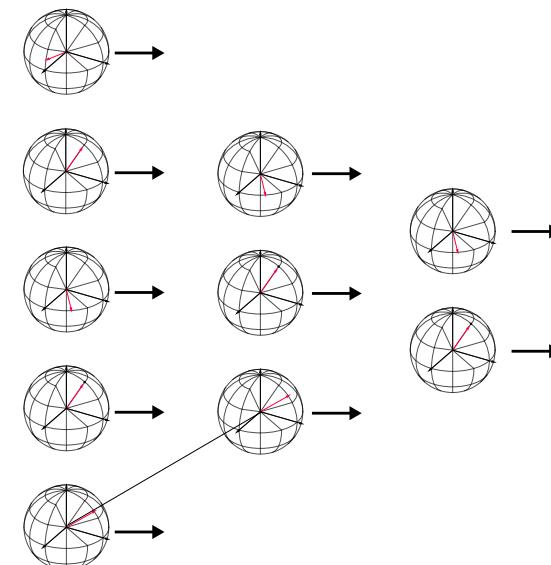
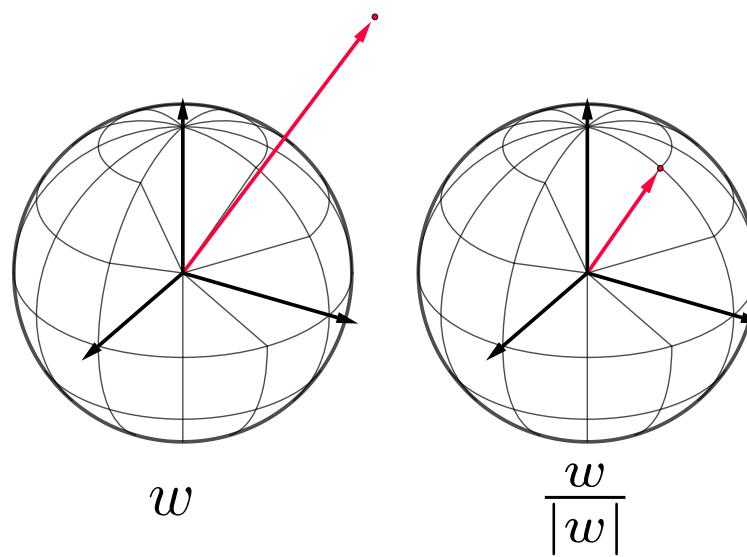
# Symmetries

## Weight-Norm

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

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

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

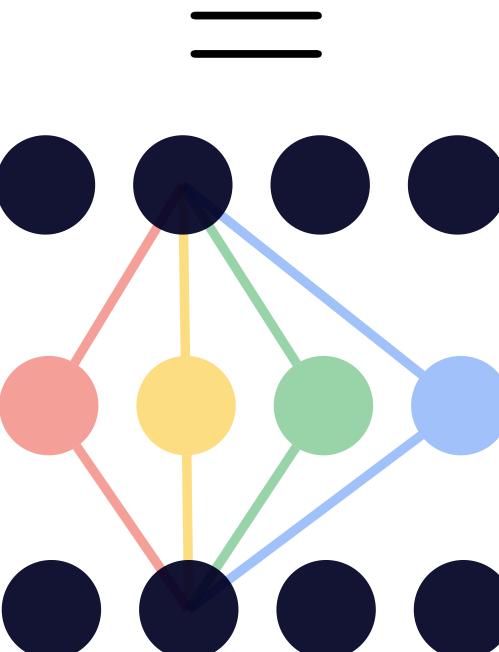
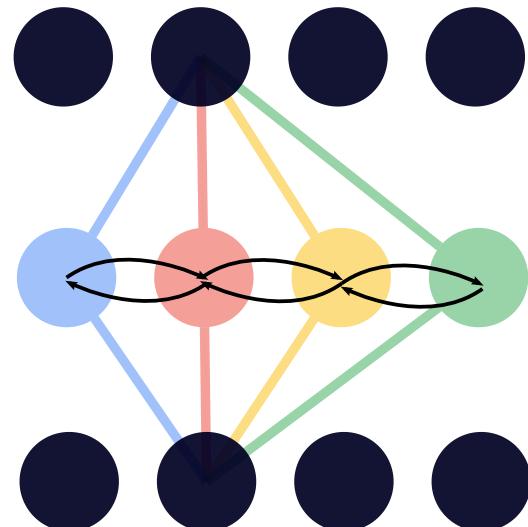
## Permutation

- 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

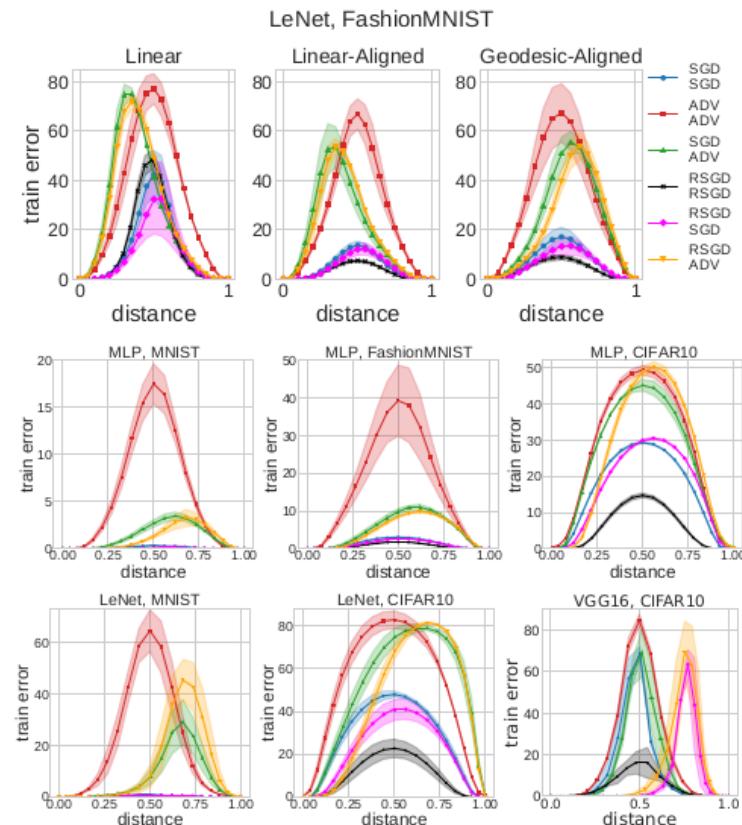
## Simple Algorithm for Matching

```
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] ,  $\pi$  ` )
end for
```

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

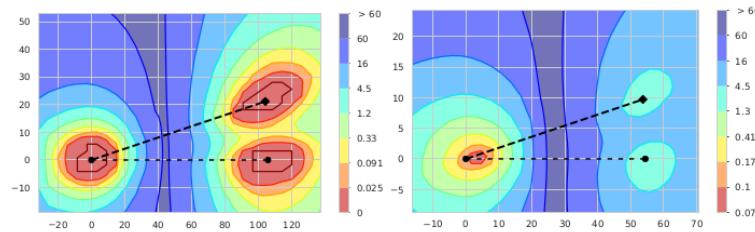


# Mode Connectivity



- 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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