

Rotation Invariant Learning on Contours

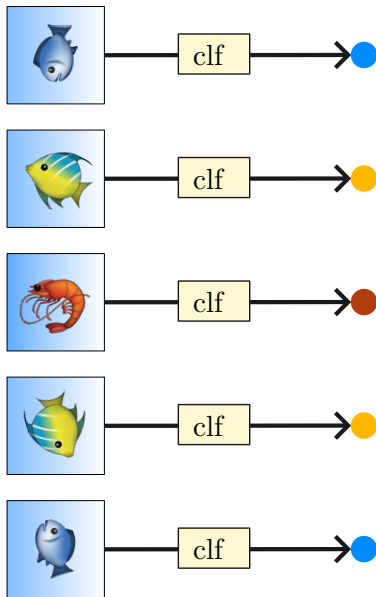
Plankton Workshop – IMR

Odin Hoff Gardå

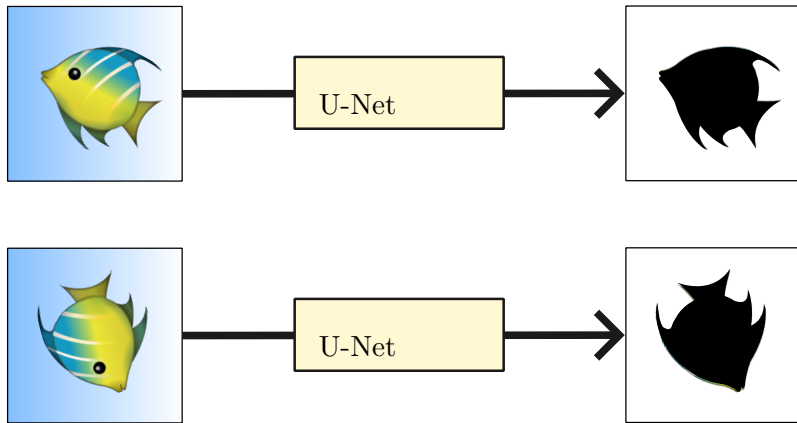
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June 3, 2025

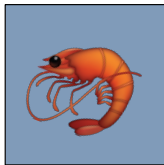
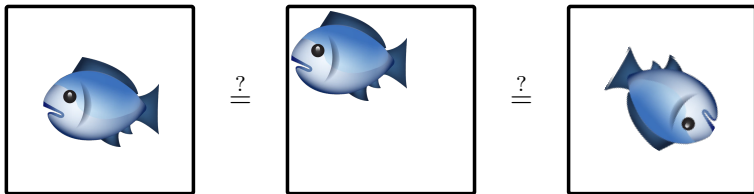
Rotation Invariance Visualized



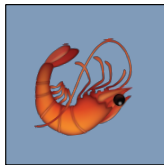
Rotation Equivariance Visualized



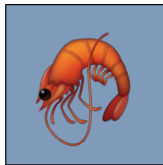
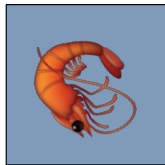
Geometric Priors



train



learn

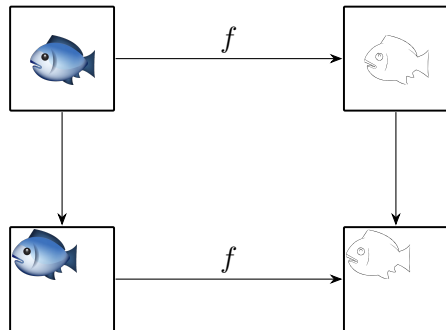


Geometric Priors

You already use them!

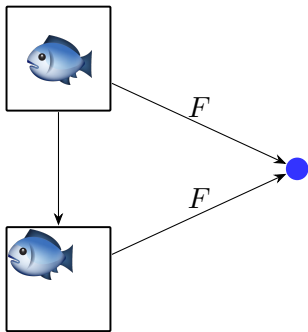
Example: CNNs (Translation Equivariance)

Let f be a convolutional layer.

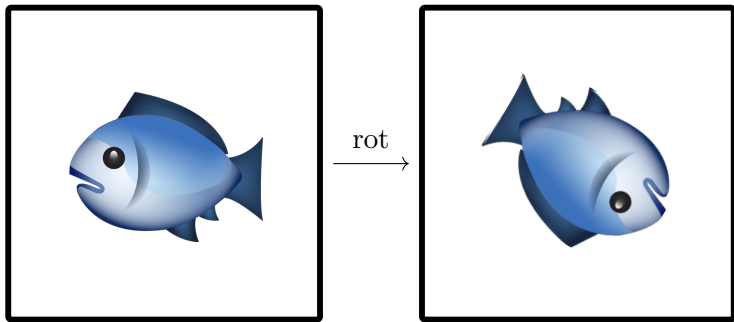


Example: CNNs (Translation Invariance)

Let F be a "fully convolutional" neural network.

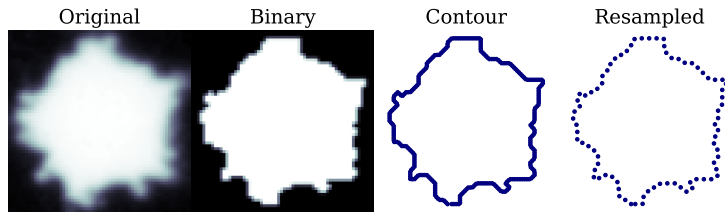


What About Rotations?



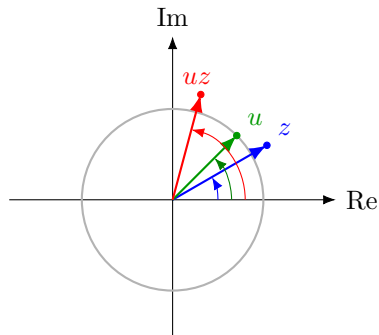
$$\text{conv}(\text{rot}(x)) \neq \text{rot}(\text{conv}(x))$$

Contours as Shape Representations



A **contour** is a sequence of points in the plane that represents the boundary of a shape.

The Complex Plane \mathbb{C}



- ▶ Complex numbers = points in the plane + multiplication.
- ▶ Rotation is multiplication with unit complex numbers.

Contours as Complex Numbers

$$\mathbb{R}^2 \cong \mathbb{C}$$

- ▶ A **contour** is a sequence (z_1, \dots, z_n) of complex numbers.
- ▶ Or a function $x: [n] \rightarrow \mathbb{C}$, where $[n] = \{1, \dots, n\}$.
- ▶ A **stack of k contours** is a function $x: [n] \rightarrow \mathbb{C}^k$.

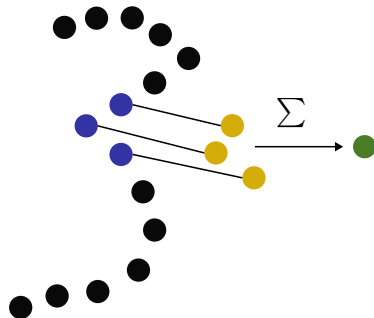
Write \mathcal{X}_n^k for the collection of all functions $x: [n] \rightarrow \mathbb{C}^k$.

Complex-valued 1D Convolution

For a filter $\phi \in \mathcal{X}_m^k$ define *circular convolution* as

$$\begin{aligned}\text{conv}_\phi: \mathcal{X}_n^k &\rightarrow \mathcal{X}_n^1 \\ x &\mapsto \phi \star x\end{aligned}$$

where $\phi \star x$ is 1D convolution of ϕ along x in the complex domain.



Equivariance: conv_ϕ is equivariant with respect to rotations and choice of starting point.

Activation Functions

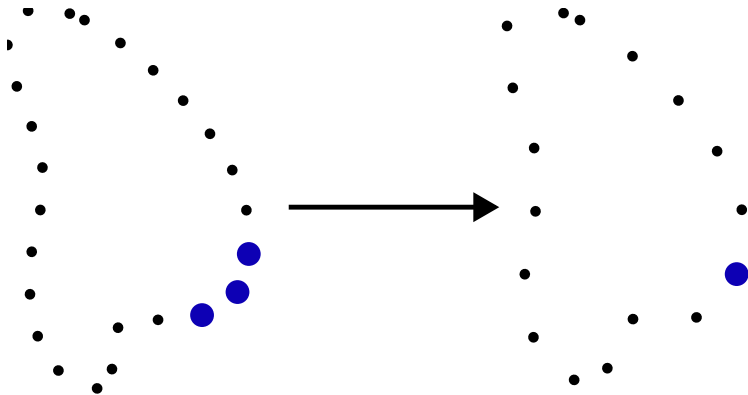
A function $f: \mathbb{C} \rightarrow \mathbb{C}$ is rotation equivariant if and only if $f(z) = g(|z|)z$ for some $g: [0, \infty) \rightarrow \mathbb{C}$.

Examples:

- ▶ Siglog: $(|z| + 1)^{-1}z$.
- ▶ Amplitude-phase-type: $\tanh(|z|)z|z|^{-1}$.
- ▶ ModReLU: $\text{ReLU}(|z| + b)z|z|^{-1}$ with learnable $b \in \mathbb{R}$.

Local Spatial Pooling (Coarsening)

$$\mathcal{X}_n^k \rightarrow \mathcal{X}_m^k \quad (m \leq n)$$



Reduce the number of points.

Global Pooling (Invariant Layer)

Rotation invariant function $\mathcal{X}_n^k \rightarrow \mathbb{R}^k$.

For example, any function depending only on radii such as mean or max, or a learnable function.

Plankton Classification Dataset

5 classes from Mesozooplankton→Animalia→Arthropoda
→Crustacea→Copepoda→Calanoida.



Class	#
Temoridae/Temora/Temora spp	257
Metridinidae/Metridia_late	271
Clausocalanidae/Microcalanus_late	421
Centropagidae/Centropages_late	345
Calanidae/Calanus_late	644
Total	1938

Split 50/50 stratified. Image size 64×64 , contour length 128.

Plankton Results

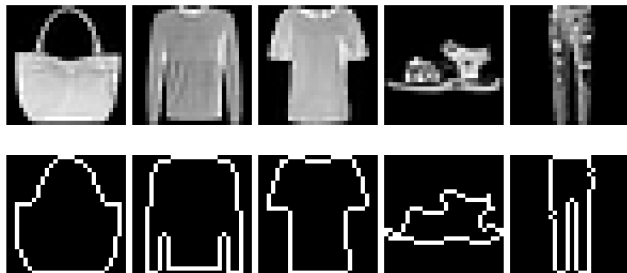
Model	Accuracy (test)	# Params
CNN	0.642 ± 0.022	$\approx 114\text{k}$
CNN+Aug	0.896 ± 0.004	$\approx 114\text{k}$
Ours	0.879 ± 0.008	$\approx 100\text{k}$
Ours+RH	0.905 ± 0.012	$\approx 101\text{k}$

Table: Classification metrics on dataset. Average over 10 runs.

- ▶ **CNN** Standard CNN on gray-scale images. With and without rotation augmentation.
- ▶ **Ours** Complex-valued 1d convolutions and global pooling.
- ▶ **Ours+RH** With a simple rotation-invariant texture descriptor (radial histogram).

Fashion MNIST Contours

Contours based on the Fashion MNIST dataset.



Model	Accuracy	# Params
CNN+Aug	0.860 ± 0.001	$\approx 294\text{k}$
Ours	0.878 ± 0.001	$\approx 77\text{k}$

Table: Performance on Fashion MNIST. Average over 30 runs.

The End

Questions?