

Estimation of normalization coefficient with a CNN

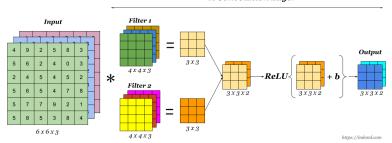




Layers and channels



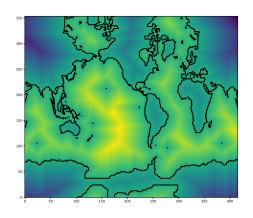
#### A Convolution Layer



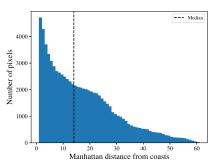
One convolution layer

### Distance from coast





Sign distance map



Distribution of ocean pixels by distance from the coast

Layers



- ▶ 1 input layer:
  - ightharpoonup 3 imes 3 convolution
  - 4 input channels, 64 output channels
  - Batch normalization
  - Activation function: Exponential Linear Unit (ELU)
- ▶ 8 hidden layers:
  - $ightharpoonup 3 \times 3$  convolution
  - ▶ 64 input channels, 64 output channels
  - Batch normalization
  - ELU
- ▶ 1 output layer
  - $ightharpoonup 3 \times 3$  convolution
  - 64 input channels, 1 output channel
- Skip connections



### Final architecture

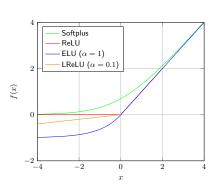


- ▶ Loss function:  $\mathbb{E}[\tilde{\varepsilon}]$  with  $\tilde{\varepsilon}$  defined component wise by  $\tilde{\varepsilon}_i = (\frac{\hat{\gamma}_i^2 \gamma_i^2}{\gamma_i^2})^2$ ,  $\forall i \in \text{Ocean cells}$
- $\triangleright 2.99 \times 10^5$  trainable parameters
- $ightharpoonup 5 imes 10^4$  epochs (number of passes on the entire training dataset)





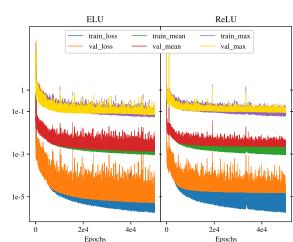
$$\begin{aligned} \operatorname{ReLU}(x) &= \begin{cases} x, & \text{if } x > 0 \\ 0, & \text{if } x \leq 0 \end{cases} \\ \operatorname{LReLU}(x) &= \begin{cases} x, & \text{if } x > 0 \\ \alpha x, & \text{if } x \leq 0 \end{cases} \\ \operatorname{ELU}(x) &= \begin{cases} x, & \text{if } x > 0 \\ \alpha (e^x - 1), & \text{if } x \leq 0 \end{cases} \\ \operatorname{Softplus}(x) &= \log(1 + e^x) \end{aligned}$$



Different activation functions

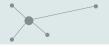


- ► ELU and Softplus gave better results than ReLU
- The best results were obtained using ELU
- ► LReLU was worse

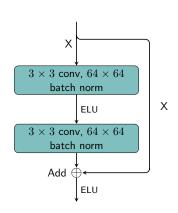


Difference in convergence using ELU and ReLU as activation functions

# Skip connections



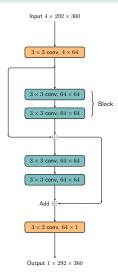
- ► 10 layers with the same input/output layers
- ▶ 8 hidden layers → 4 blocks with 2 layers each
- Each block adds its input to its output



Block details



# Architecture representation

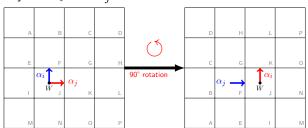


Architecture representation with 2 blocks and skip connections

## Data augmentation



- Data augmentation aims to artificially increase the amount of data by transforming existing data
- We tried **rotations** and **flips**. Further modifications are necessary for  $\alpha_i$  and  $\alpha_j$

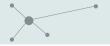


Representation of the 90° rotation

But it did not work



# The results on toy model



### Randomized $(10^4)$ :

- For one 1 sample:
- ► Mean absolute relative error: 1.1 %
- ► Max absolute relative error: 5.6 %
- ▶ Quantile 99,99 of the relative error: 5.4 %

### CNN:

- ► For one 90 samples:
- ► Mean absolute relative error: 0.65 %
- ► Max absolute relative error: 2.81 %
- ▶ Quantile 99,99 of the relative error: 1.90 %

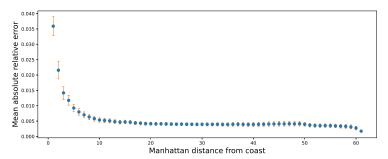


► For one 190 samples:

▶ Mean absolute relative error: 0.38 %

► Max absolute relative error: 4.27 %

▶ Quantile 99,99 of the relative error: 1.87 %



Mean absolute relative error in relation to the distance to the coast

# Computation cost



- ▶ On kraken: gpu partition (Tesla V100-PCIE-16GB)
- ▶ Training duration: 9 hours to 16 hours  $(2.99 \times 10^5 \text{ to } 7.45 \times 10^5 \text{ trainable parameters})$ . But it only has to be performed **once**.
- ► Computation speed on 1 sample: 1 to 2 seconds

# Conclusion and perspectives



- ► The CNN provides convincing results and is better than the randomization method in the context of the experiment
- ▶ We thought to perform the training on sections of the image
- Expand to the 3-dimensional case

Thank you for your attention!