## **Optimization of Hybrid Dealiasing Convolutions**

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#### **Overview**

# The convolution theorem provides a method for

efficiently computing convolutions using Fast Fourier Transforms (FFTs), with the caveat that the convolution is periodic. For two arrays f and g that have period M the convolution is defined as

$$(f * g)_l = \sum_{j=0}^{l} f_j \ g_{l-j} \ \forall l \in \{0, \dots, M-1\}$$

Many applications such as nonlinear PDEs require a linear convolution, which can be computed using the convolution theorem, provided that the data is sufficiently padded with zeros so we can avoid aliases. > Implicit dealiasing provides an alternative to explicit dealiasing. To maximize performance, we use hybrid padding: the padding is chosen to be fully implicit, fully explicit, or a combination of the two.

#### **Problem**

- Find a rectangle such that I can use its parameters to estimate a large square/rectangles parameters.
- > Also create a more optimized and efficient hyperparameter tuning algorithm called *experience* to find the optimal parameters.

Which rectangle's parameters can be used to estimate the square's parameters?



#### References

Noel Murasko and John C. Bowman. Hybrid Dealiasing of Complex Convolutions. Submitted to SIAM J. Sci. Computer, 2023.

### Algorithm

- We generated the data by looking only at square sizes of the form  $2^a 3^b 5^c 7^d$  where *a*, *b*, *c*, *d* are integers. For each square we look over all the rectangles by varying the height (Ly) but keeping the same length (Lx).
- > We then use the optimizer to get the results for each rectangle and cache all the data.
- Lastly we visualize and analyze the data to check for any patterns, fit Machine Learning models and come up with heuristics. Lastly we evaluate them on other datasets.



#### **Results**

- **Takeaway result 1**: 3 interesting rays of *m* values. Only need to search over those 3 rays.
- > Takeaway result 2: Simply use very skinny rectangles instead. It seems rectangles with height 32 is comparable enough.





#### **Tuning Parameters**

3 data-dependent parameters. L, M and the number C of FFTs to be computed simultaneously. Other important parameters. Size *m* of the FFTs, The number D of residues at a time, In-place (I = 1)or out-of-place (I = 0) FFTs.

### **Machine Learning**

- Results seem to be promising especially when switching to a classification task.
- Caveats include that the results are hardware dependent and does not generalize well.
- Conjecture that the predicted value (from classification) timing will be comparable to the ground truth.



#### **Future Work & Applications**

- > These heuristics will be useful in practice and can contribute to the success of the proposed hybrid dealiasing algorithm.
- We expect hybrid dealiasing to become the standard method for convolutions and are developing hybrid dealiasing for real convolutions and for Imaging and **Convolution Neural Networks.**