

# The CoRa Tensor Compiler: Compilation for Ragged Tensors With Minimal Padding

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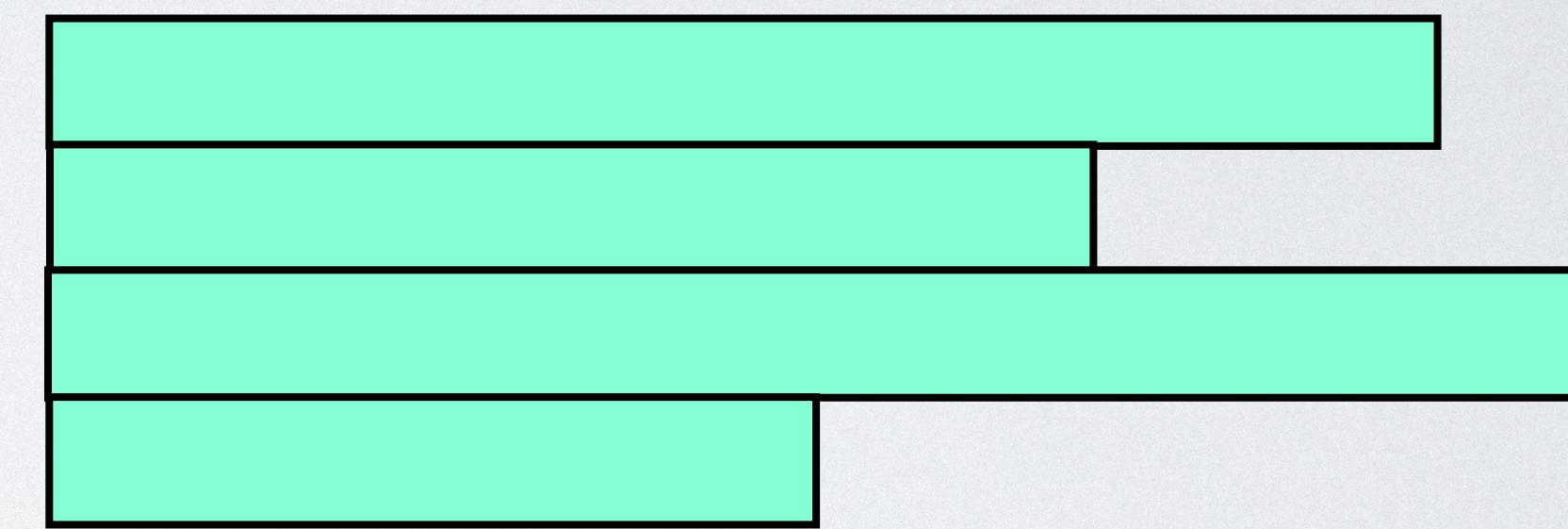
<sup>2</sup>OctoML

# Ragged Tensors in Deep Learning

- Natural language processing

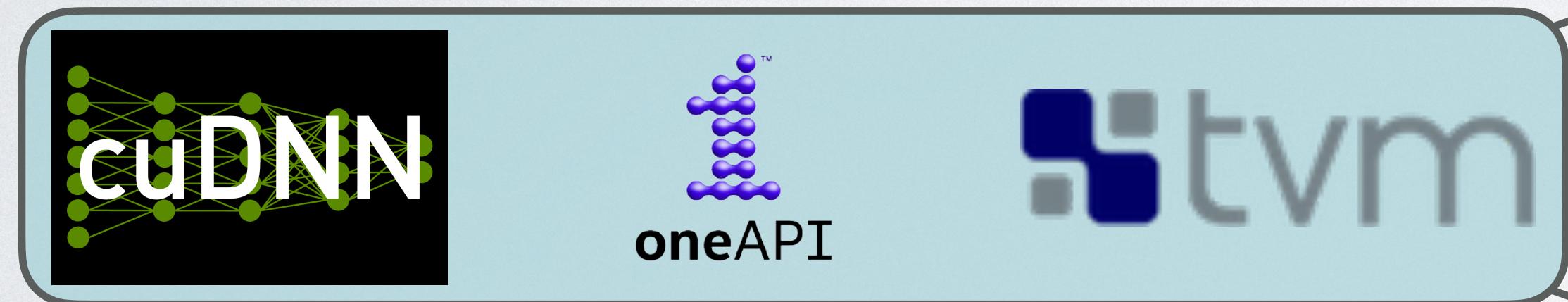
```
input_batch = [  
    3 [Dogs, bark, .],  
    5 [Maine, is, a, state, .],  
    4 [The, song, rocks, !],  
    | [Hello]  
]
```

- Image processing

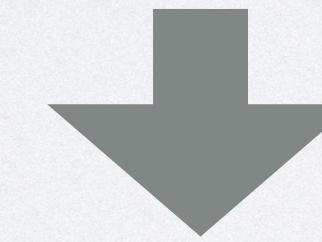


Ragged Tensor

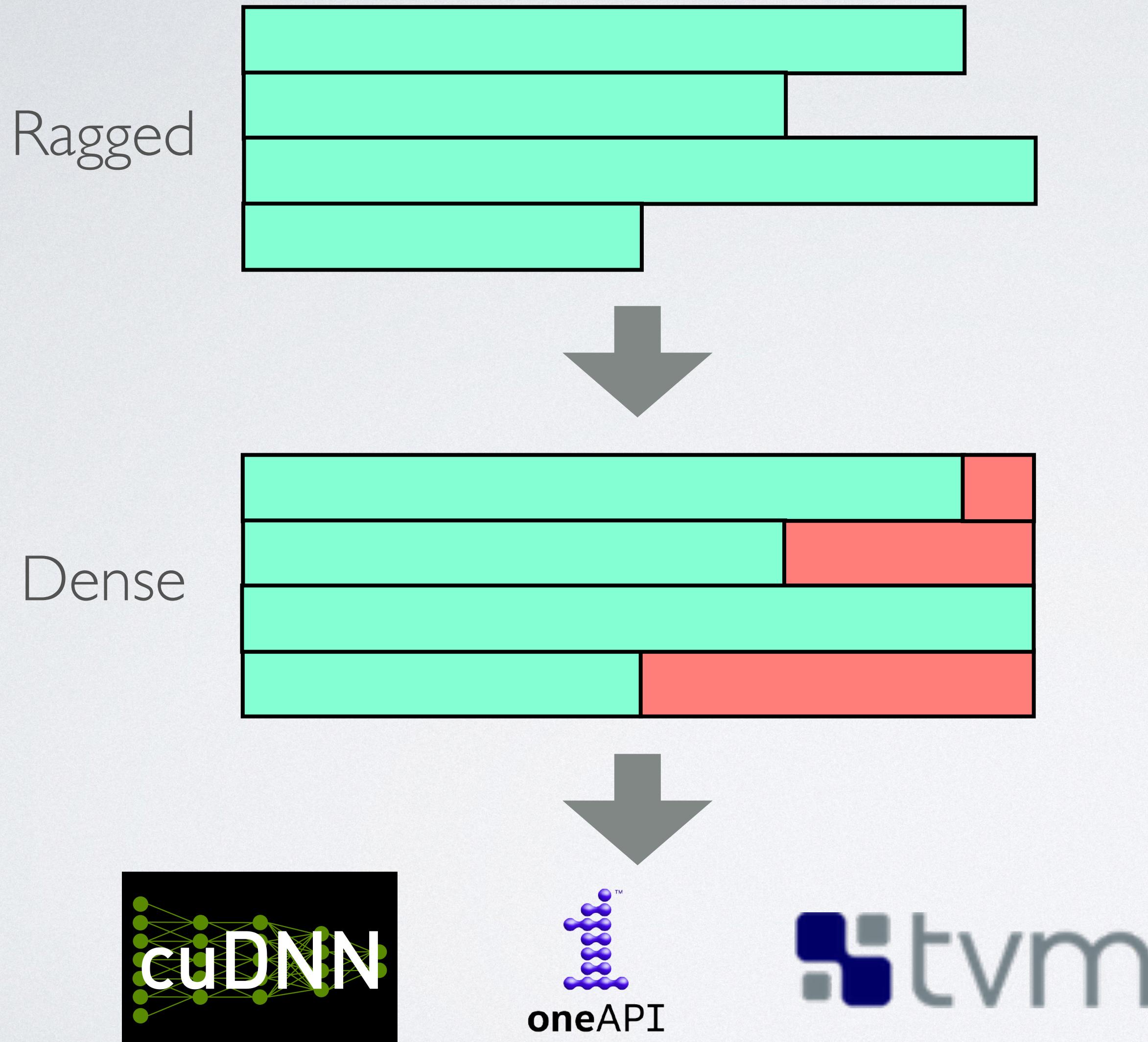
# Limited Support for Ragged Tensor Operators



- Limited support for ragged tensors
- Extensive support for dense tensors

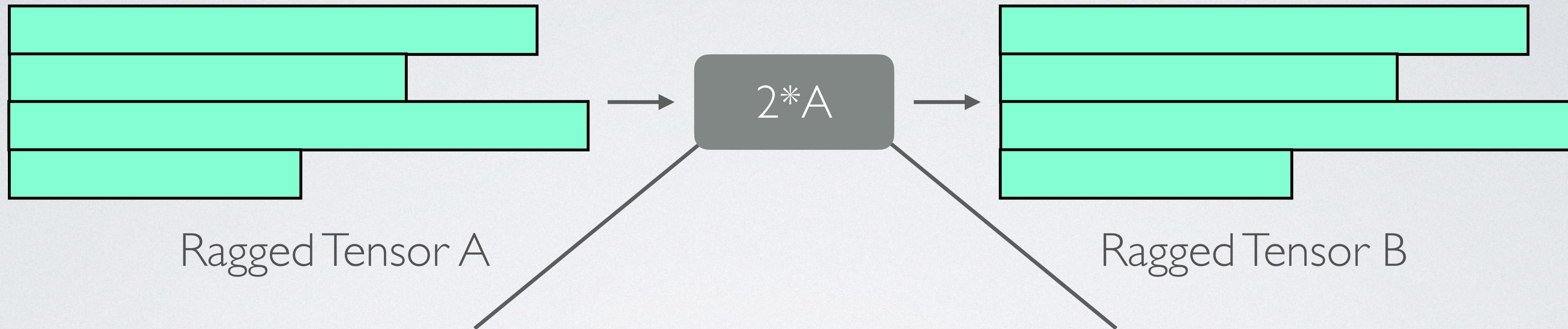


# Padding Leads to Wasted Computation



1.07 - 2.41X wasted  
computation for a  
transformer encoder layer!

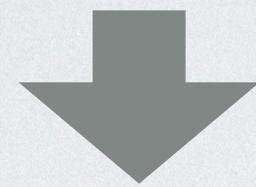
# Ragged Computations Are Similar to Dense Computations



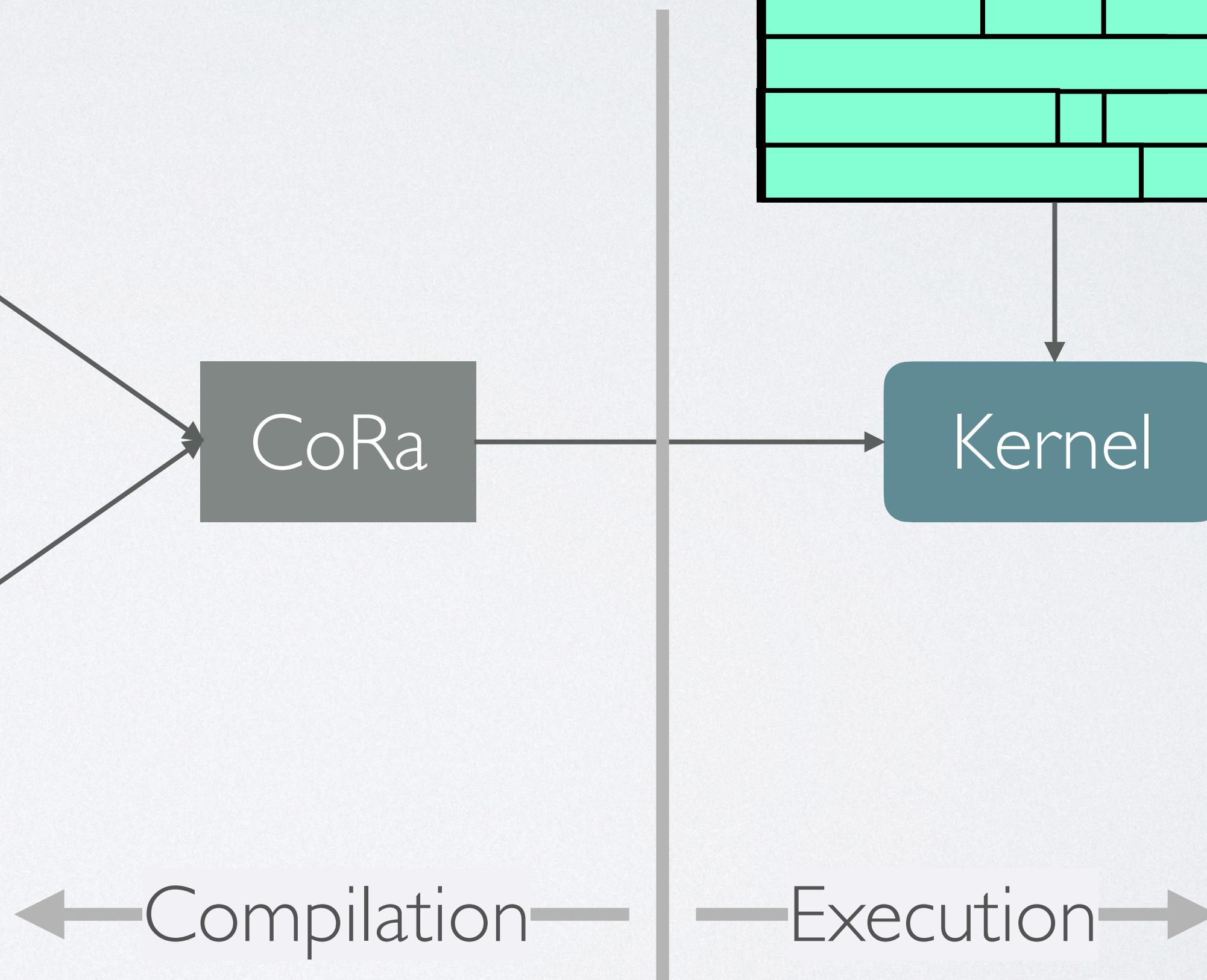
```
for i in 0:32:  
  for j in 0:s(i):  
    B[i, j] = 2*A[i, j]
```

# CoRa: a Tensor Compiler for Ragged Tensors

```
for i in 0:32:  
    for j in 0:s(i):  
        B[i,j] = 2*A[i,j]
```

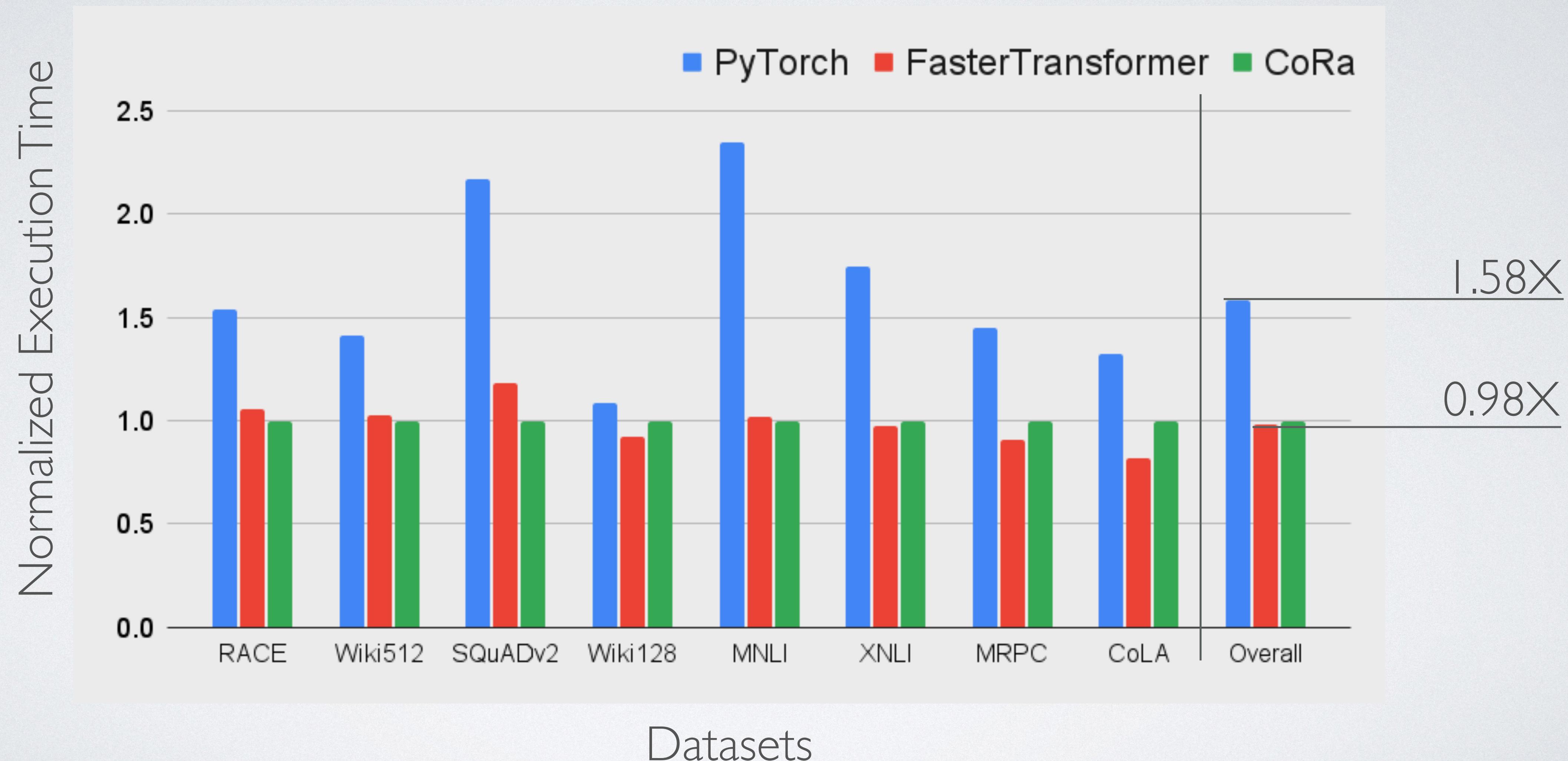


```
loop_exts = [32, lambda b: s(b)]  
A = input_tensor(loop_exts)  
O = compute(loop_exts,  
           lambda i,j: 2*A[i,j])  
  
i, j = O.axis  
s.padding(j, 32)  
jo, ji = s.split(j, nparts=32)  
s.bind(i, 'blockIdx.x')  
s.bind(jo, 'threadIdx.x')
```

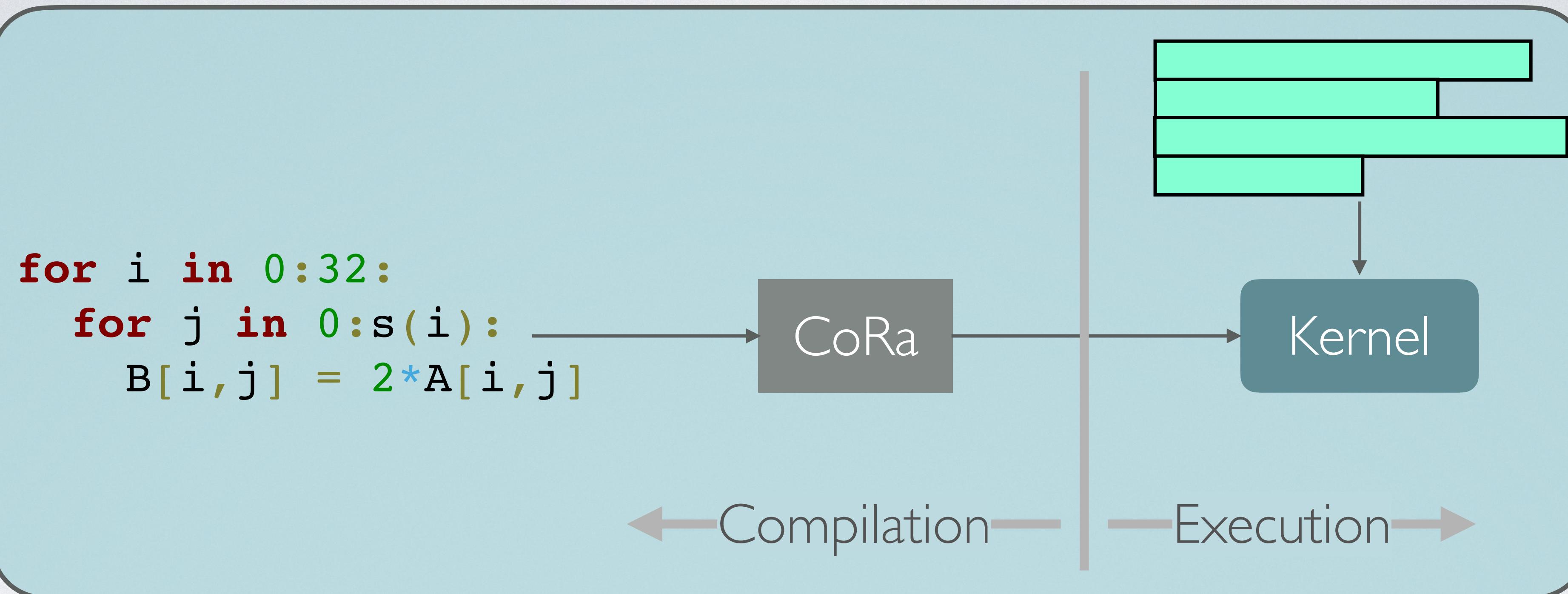


# Transformer Layer Layer Forward Latencies on Nvidia V100

Lower is better



# Conclusion



More details can be found in our [paper](#)