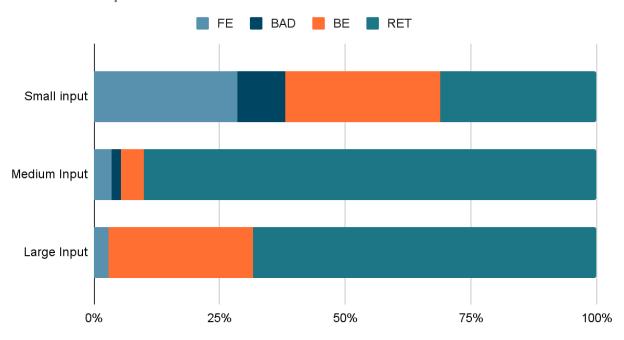
When profiling a function, I repeated the core function 10-100 times. I also profiled with 3 different input magnitudes (10, 100, 1000). I repeated the core function a lesser number of times for larger inputs and a higher number of times for smaller inputs. For all tests, I first allocated memory for each of the input variables. Then, I assigned values to them sequentially.

#### Issues in the lab

- Writing and debugging test cases was laborious as I was new to C and the process of allocating memory before using dynamically sized arrays
  - o After discovering gdb, I was able to debug seg faults much faster
- The manual process of copying the values after running each profiling was also time-consuming

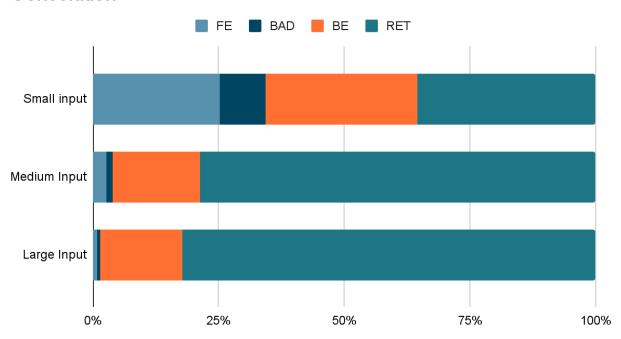
# **Matrix Multiplication**



- Implementation I ran 3 nested loops which iterated over the A row, B col, and the Acol/Brow
- Test I tested with matrix sizes of (3,3), (100,100), (1000,1000)
- Bottleneck observation -
  - As I increased the input size,
    - % share of the frontend bound decreased
    - % share of bad speculation decreased
    - % share of backend-bound decreased for the medium input and increased for the large input

- % share of retiring increased for medium input and decreased for large input
- The main bottleneck seems to be retiring

## Convolution



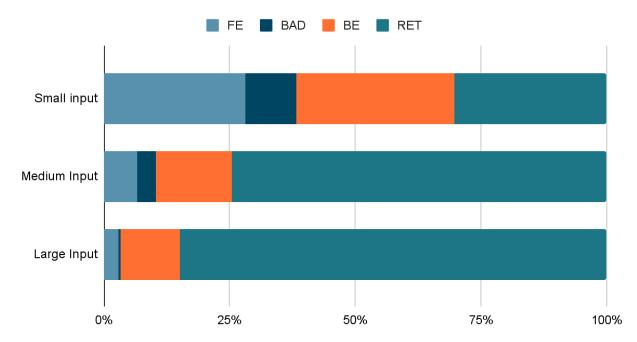
- Implementation I ran 6 nested loops. The first 3 loops iterate through each cell of the final output. The inner 3 loops loop through the row, col, and channel dimensions of the image.
- Test
  - I used the following sizes for testing
    - Channels 3
    - Filters 1
    - Kernel 2
    - Input image size 3, 100, 1000
- Bottleneck observation -
  - As I increased the input size,
    - % share of the frontend bound decreased
    - % share of bad speculation remained unchanged
    - % share of backend-bound remained unchanged
    - % share of retiring increased
  - The main bottleneck seems to be retiring

Cache data after running the largest conv output

```
Withilitation features

Withil
```

#### Linear



- Implementation I ran a loop for each row of the input function. And another inner loop for multiplying and summing the cols of each input row
- Test
  - o I used the following sizes for testing
    - I used the same dimensions as the input for both the row and col dimension of the weights
    - Input size 3, 100, 1000
- Bottleneck observation -
  - As I increased the input size,
    - % share of the frontend bound decreased

- % share of bad speculation remained unchanged
- % share of backend-bound remained unchanged
- % share of retiring increased
- The main bottleneck seems to be retiring

## Relu



- Implementation I calculated the relu function via if statements
- Test
  - o I used the following sizes for testing
    - I used the same dimensions as the input for both the row and col dimension of the weights
    - Input size 3, 100, 1000
- Bottleneck observation -
  - As I increased the input size,
    - % share of the frontend remained unchanged
    - % share of bad speculation remained unchanged
    - % share of backend-bound remained unchanged
    - % share of retiring remained unchanged
  - The main bottleneck seems to be backend-bound

#### Softmax



- Implementation I calculated the softmax by calculating the sum via 1 loop and then the softmax values via another loop
- Test
  - I used the following sizes for testing
    - I used the same dimensions as the input for both the row and col dimension of the weights
    - Input size 3, 100, 1000
- Bottleneck observation -
  - As I increased the input size,
    - % share of the front remained unchanged
    - % share of bad speculation remained unchanged
    - % share of backend-bound remained unchanged
    - % share of retiring remained unchanged
  - The main bottleneck seems to be backend-bound

#### Potential improvements

 Since most of the large input bottleneck is retiring in nature, I think the performance can be improved overall by implementing multiple threads as many of the sub-operations are independent of each other