



2. Problem Characteristics of NNs

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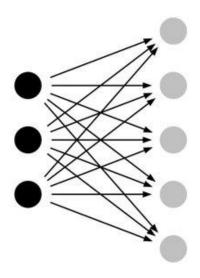






Content

- 1. Algorithmic Description
- 2. Possible Optimizations



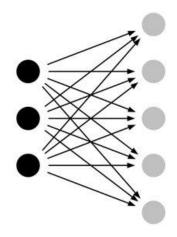




Pseudocode Description

Fully connected layer

```
for (b = 0 to B-1) // batch
for (n = 0 to N-1) // neuron
for (i = 0 to I-1) // input
```

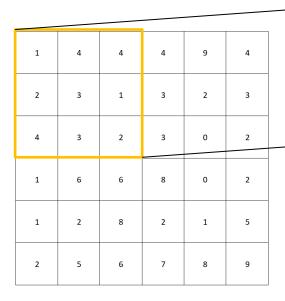






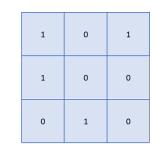
Working with Filters

2D Example



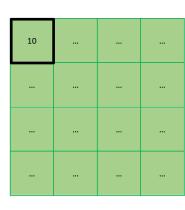
1	4	4
2	3	1
4	3	2

Part of Input



*

Filter Kernel



Output

Input

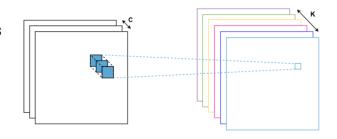




Pseudocode Description

Convolutional layer

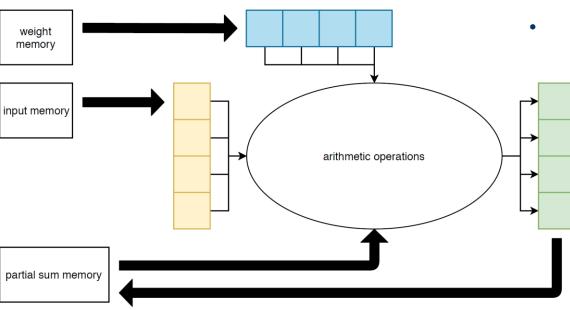
```
for (b = 0 to B-1) // batch for (k = 0 to K-1) // output channels for (c = 0 to C-1) // input channels for (x = 0 to X-1) // input columns for (y = 0 to Y-1) // input rows for (fx = 0 to FX-1) // filter columns for (fy = 0 to FY-1) // filter rows
```







Processing



- What data is used in one cycle?
- What data is used in consecutive cycles?







Weights



Outputs





Things can be Optimized

Fully connected layer

```
for (b = 0 to B-1) // batch
for (n = 0 to N-1) // neuron
for (i = 0 to I-1) // input
```



Convolutional layer

```
for (b = 0 to B-1) // batch for (k = 0 to K-1) // output channels for (c = 0 to C-1) // input channels for (x = 0 to X-1) // input columns for (y = 0 to Y-1) // input rows for (fx = 0 to FX-1) // filter columns for (fy = 0 to FY-1) // filter rows
```

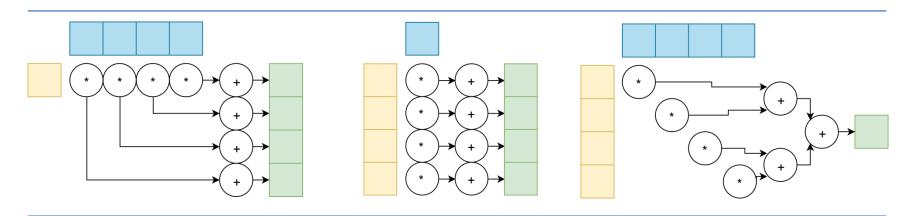


Data stationarity





Data Parallelism



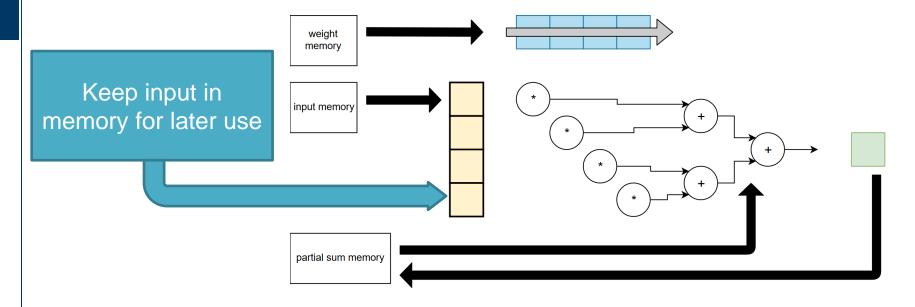
Input Parallel	Weight Parallel	Output Parallel
Reduce input bandwidth	Reduce weight bandwidth	Reduce output bandwidth





Data Stationarity

Input Stationarity







Data Stationarity

Weight Stationarity

Keep weights in memory for later use

