

# Accelerating DNN Training Through Joint Optimization of Algebraic Transformations and Parallelization

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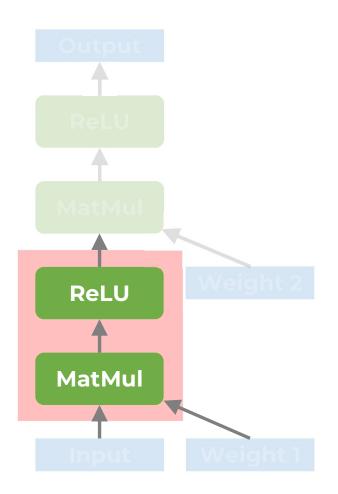
# **Unity**

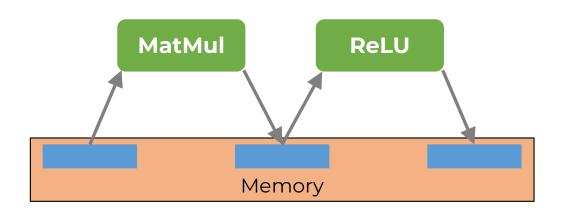
Accelerating DNN Training Through Joint Optimization of Algebraic Transformations and Parallelization

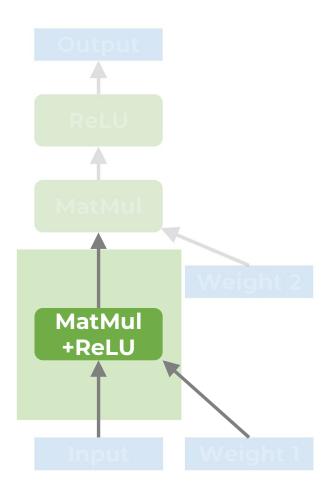
# 1. Algebraic Transformations

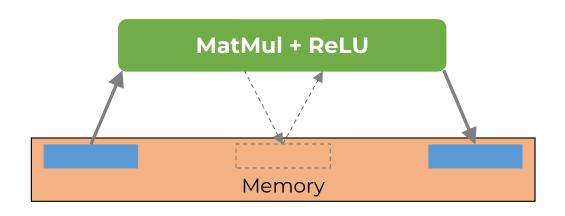
# 2. Parallelization

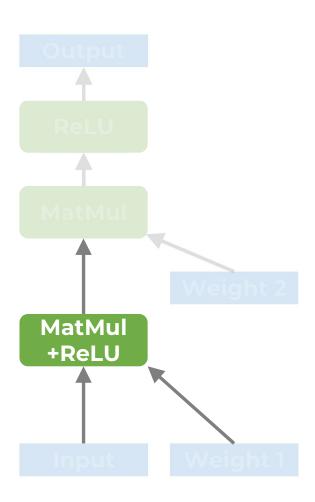
# 1. Algebraic Transformations



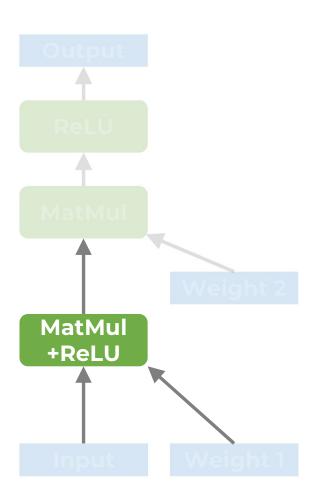








# Operator Fusion



# Operator Fusion Operator Splitting

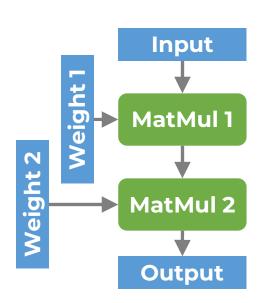
Operator Fusion
Operator Splitting
Operator Reordering

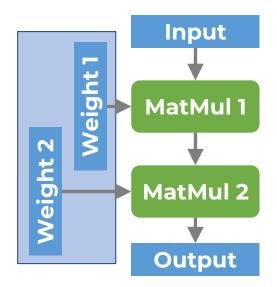
Operator Fusion
Operator Splitting
Operator Reordering

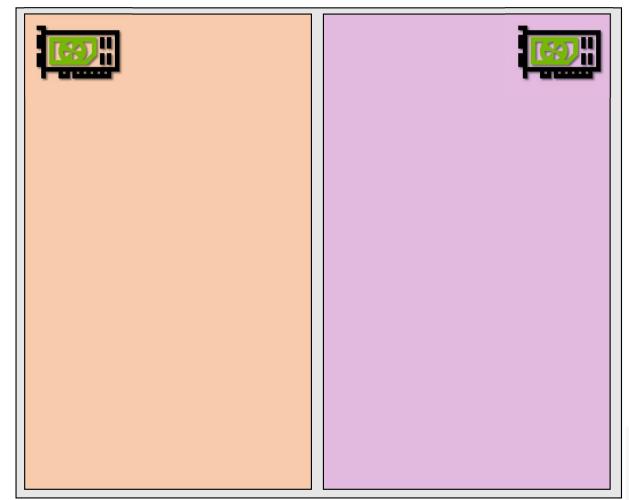
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# 1. Algebraic Transformations

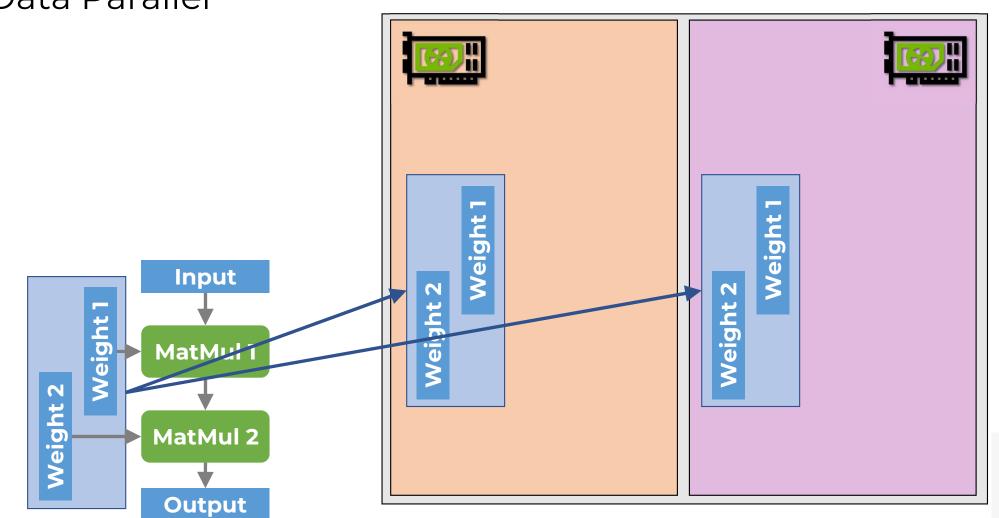
2. Parallelization



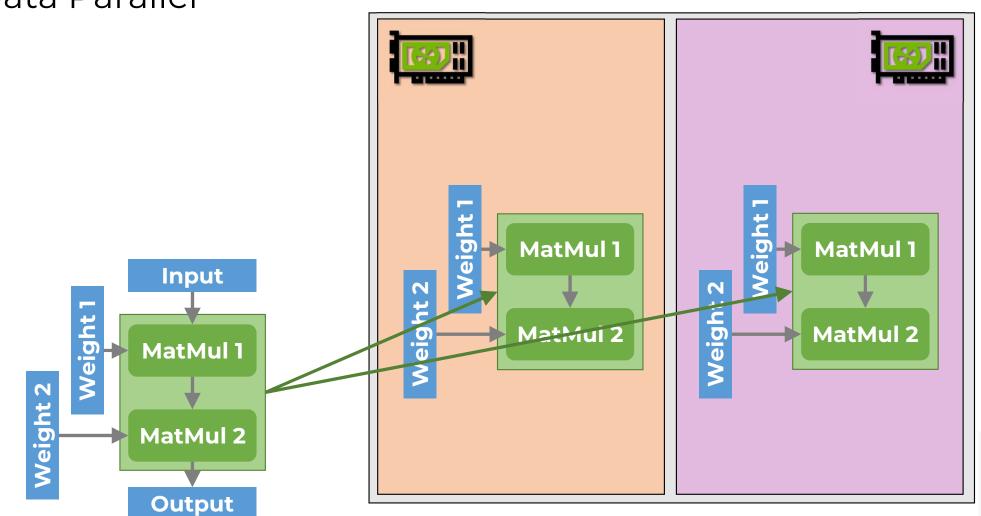




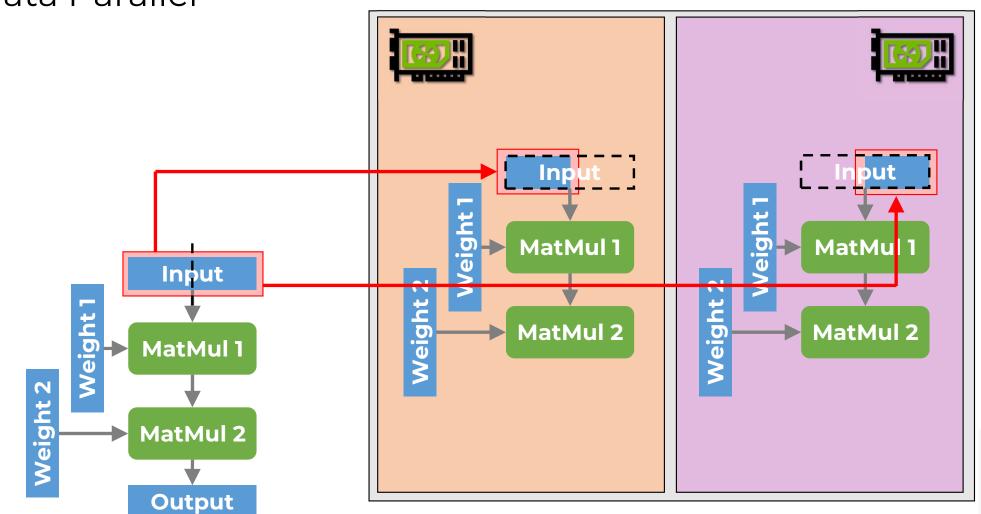
0 5 /



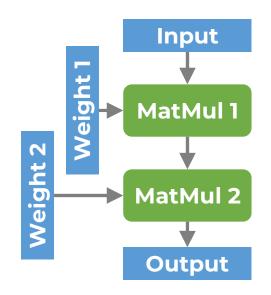
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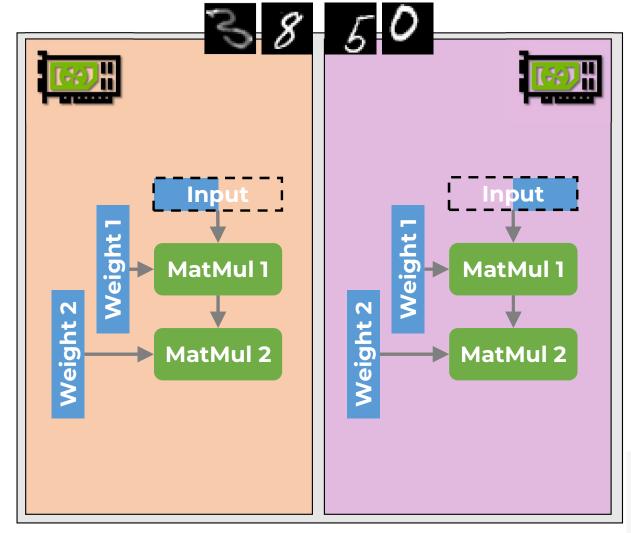


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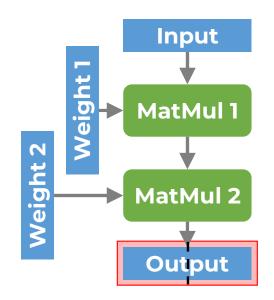


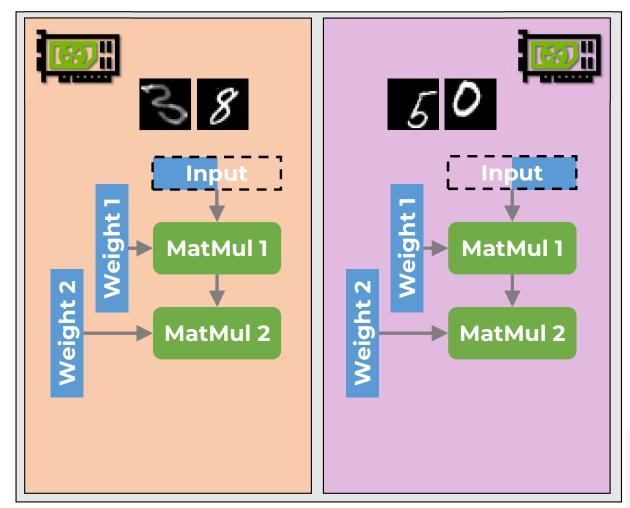
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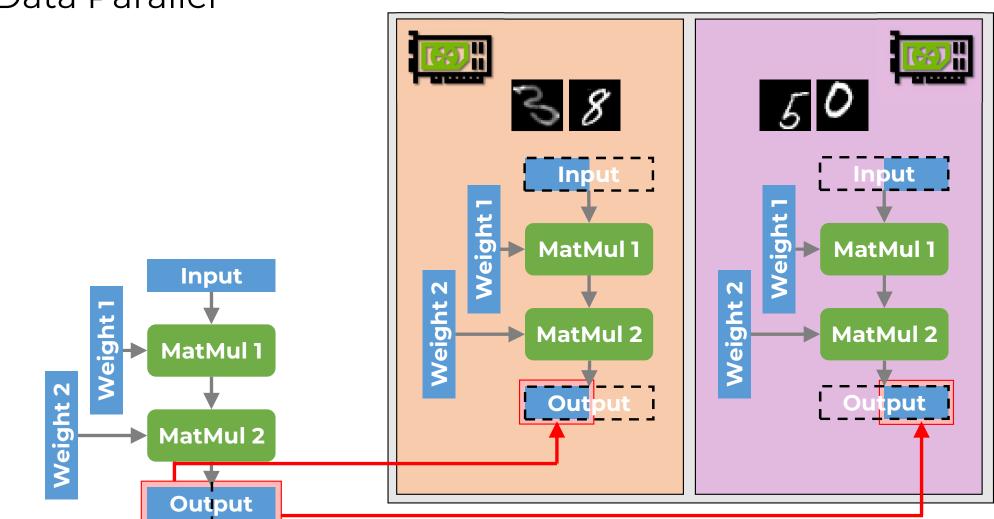


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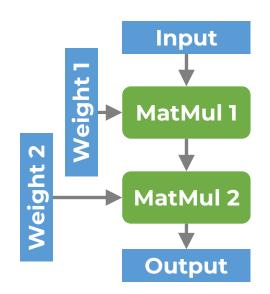


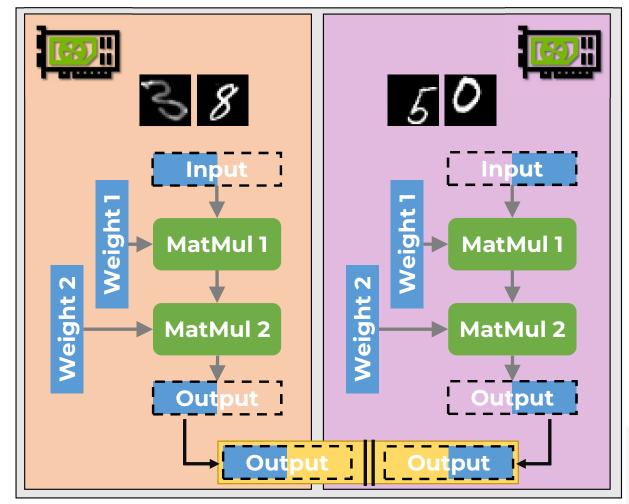


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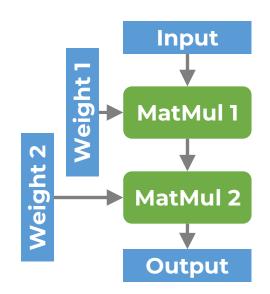


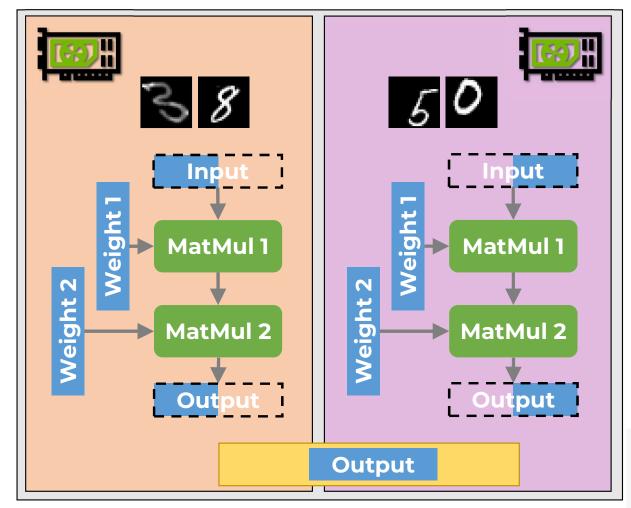
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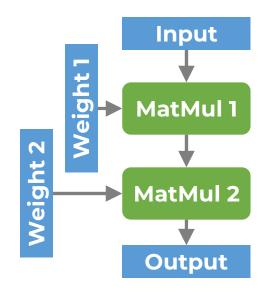


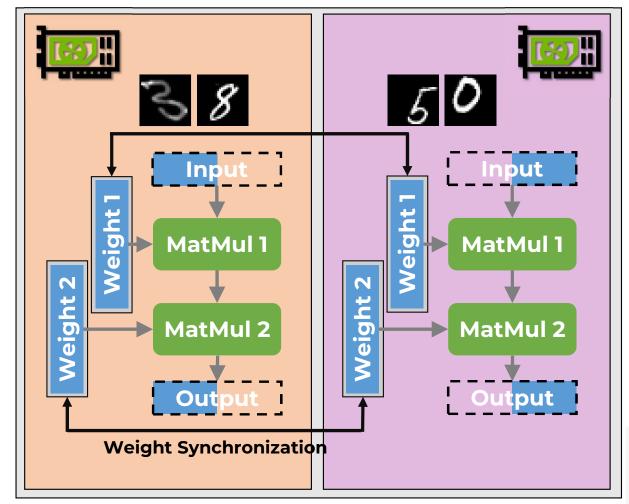
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### Model Parallel

# Data Parallel Model Parallel

#### **Attribute Parallel**

Operation	Parallelizable Dimensions		
	(S)ample	(A)ttribute	(P)arameter
1D pooling	sample	length, channel	
1D convolution	sample	length	channel
2D convolution	sample	height, width	channel
Matrix multiplication	sample		channel

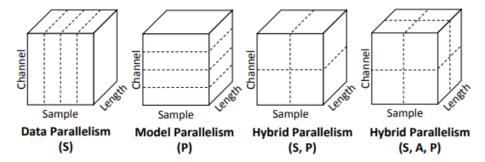


Figure 3: Example parallelization configurations for 1D convolution. Dashed lines show partitioning the tensor.

Data Parallel Model Parallel Attribute Parallel

Reduction Parallel

Data Parallel
Model Parallel
Attribute Parallel
Reduction Parallel

Parameter Parallel

Data Parallel
Model Parallel
Attribute Parallel
Reduction Parallel
Parameter Parallel
Pipeline Parallel

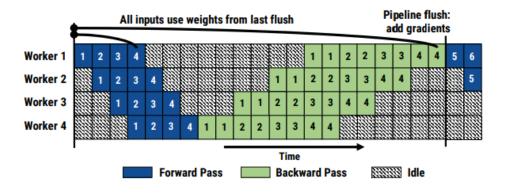


Figure 3: GPipe's inter-batch parallelism approach. Frequent pipeline flushes lead to increased idle time.

Data Parallel
Model Parallel
Attribute Parallel
Reduction Parallel
Parameter Parallel
Pipeline Parallel
...

Parallelization

#### Algebraic Transformations

Operator Fusion
Operator Splitting
Operator Reordering

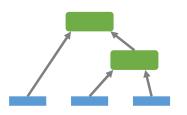
#### Auto-Parallelization

```
FlexFlow [MLSys 19]
      Tofu [EuroSys 19]
PipeDream [SOSP 19]
  automap [arXiv 19]
    Whale [arXiv 21]
      Alpa [OSDI 22]
```

## Algebraic Optimizers

```
MetaFlow [MLSys 19]
TASO [SOSP 19]
PET [OSDI 21]
Tensat [MLSys 21]

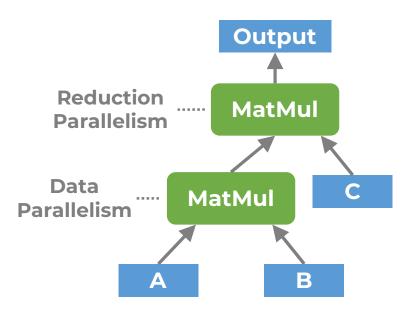
•••
```



#### Auto-Parallelization

#### Algebraic Optimizer



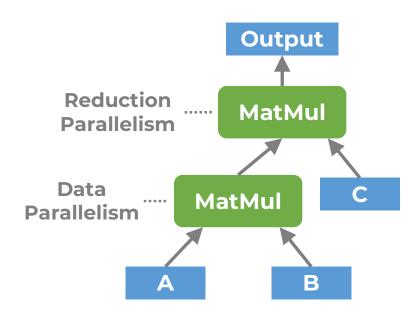


"annotated computation graph"









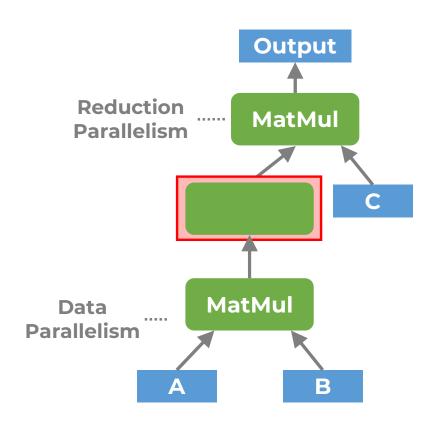


Auto-Parallelization



#### Algebraic Optimizer





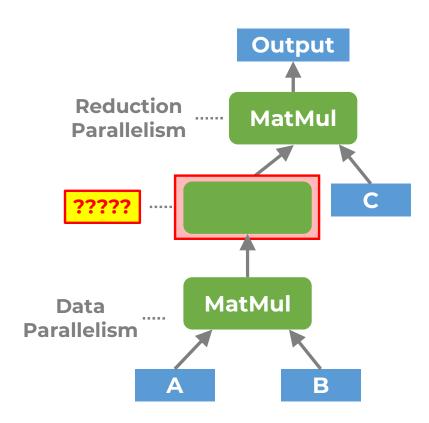


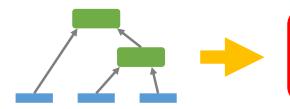
Auto-Parallelization



#### Algebraic Optimizer







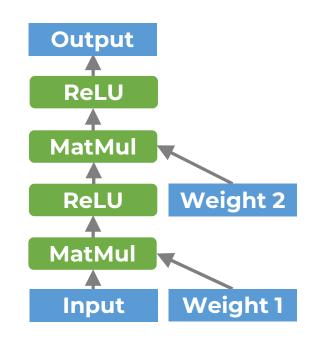








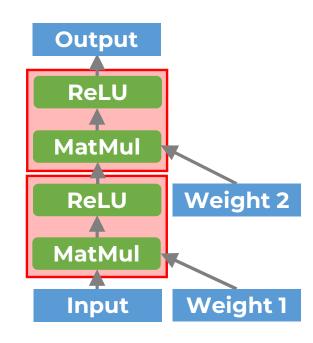








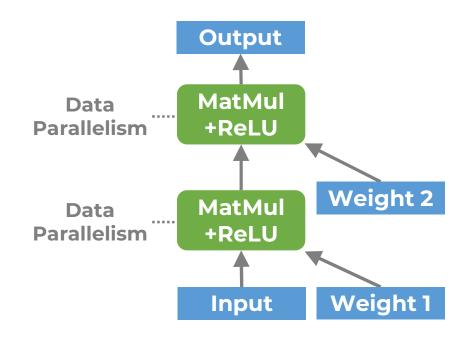


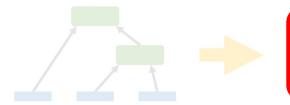








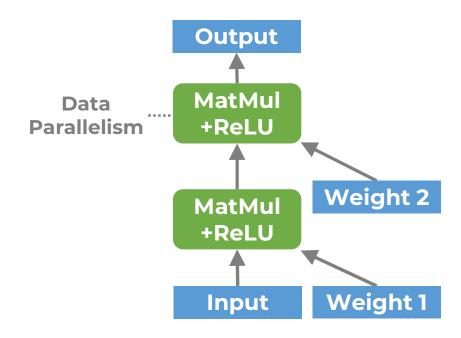


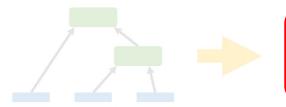






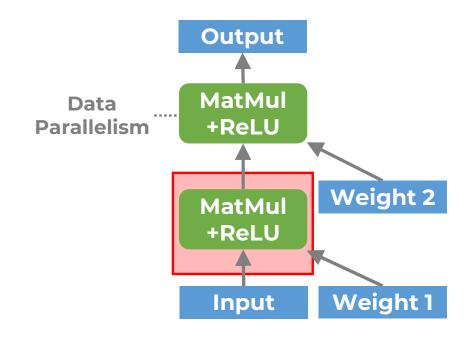


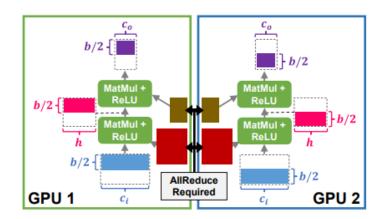


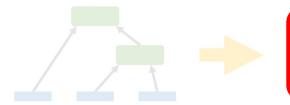








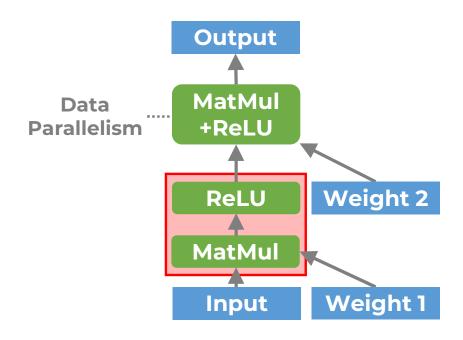






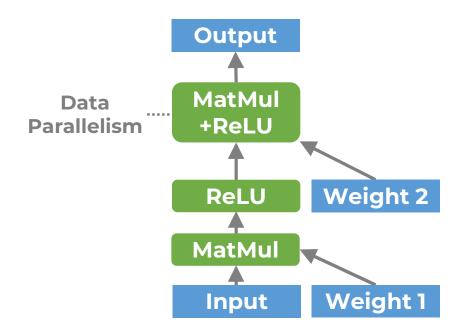










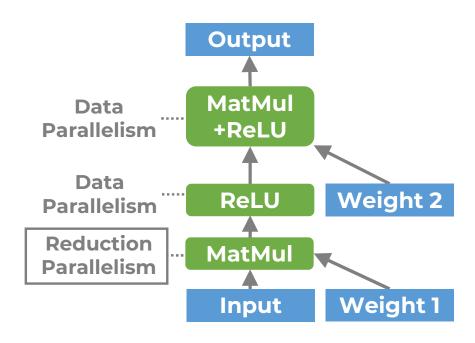










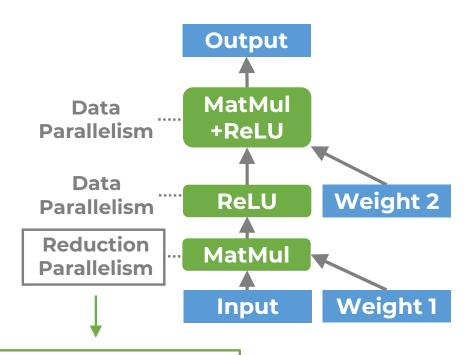


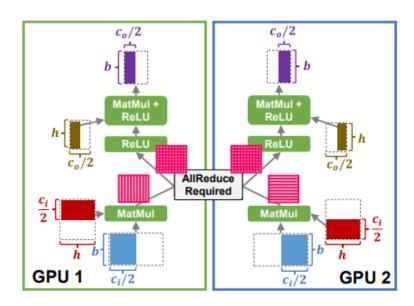




Auto-Parallelization







pprox 6 imes less communication!



### 1. Representation

2.



## 1. Representation

## 2. Scalability

# Unity

# -Representation Unity

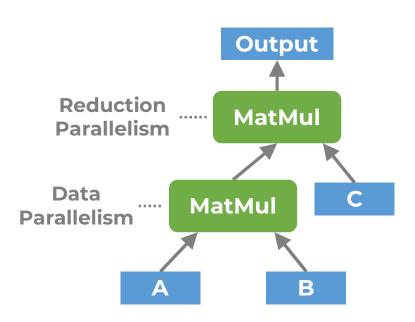
## Representation -Parallel Computation Graph (PCG) Unity

Algorithm

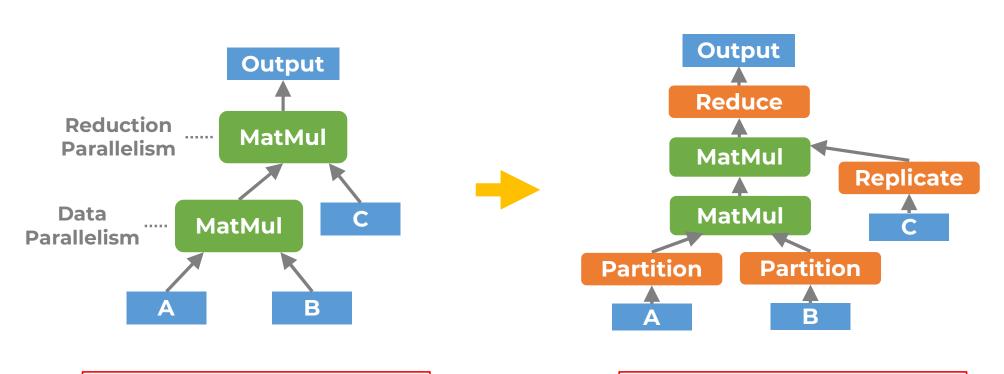
1 3 /

#### Representation

## Parallel Computation Graph (PCG)

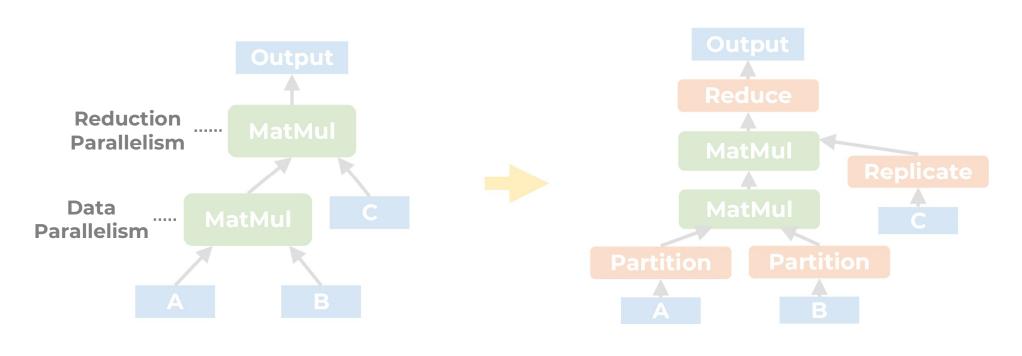


annotated computation graph



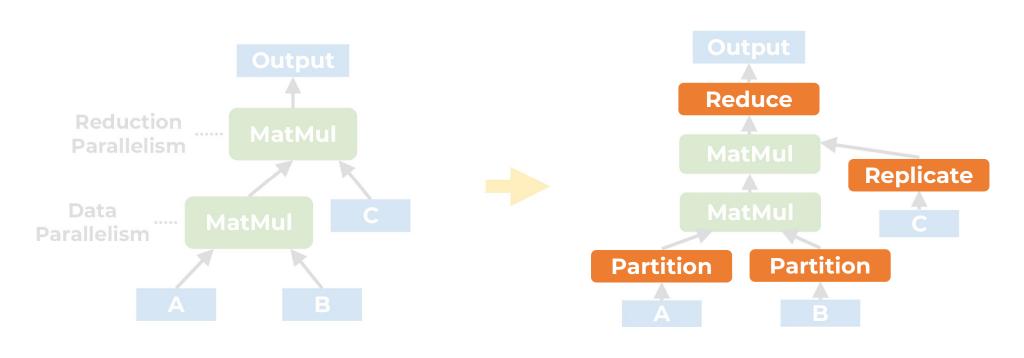
annotated computation graph

parallel computation graph (PCG)



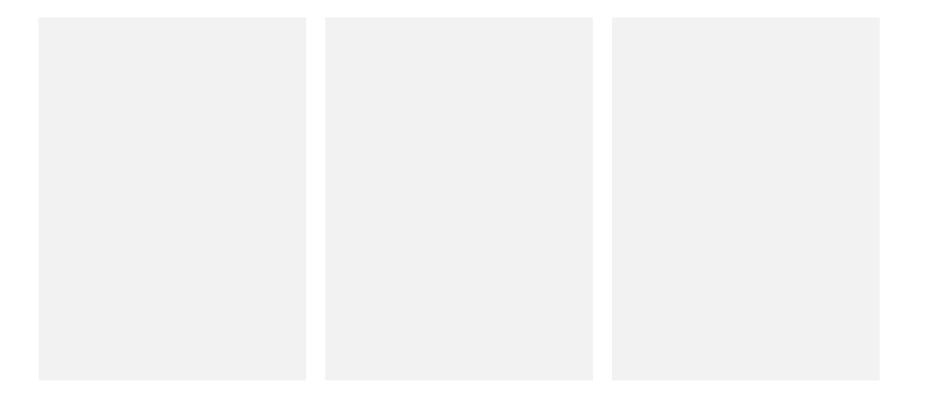
annotated computation graph

parallel computation graph (PCG)



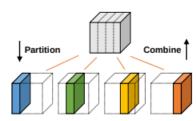
annotated computation graph

parallel computation graph (PCG)



#### **Partition**

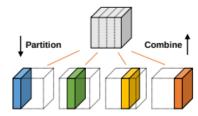
#### Combine

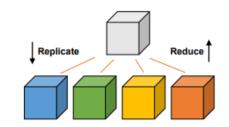


Replicate

Combine

Reduce



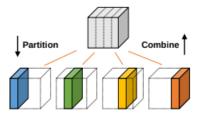


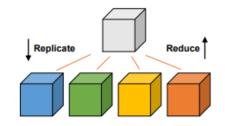
Replicate

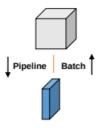
**Pipeline** 

Combine

Reduce





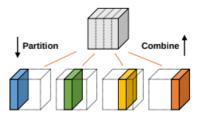


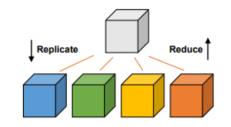
Replicate

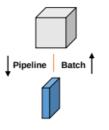
Pipeline

Combine

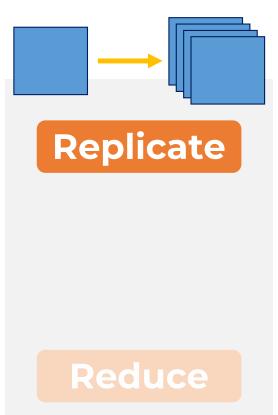
Reduce

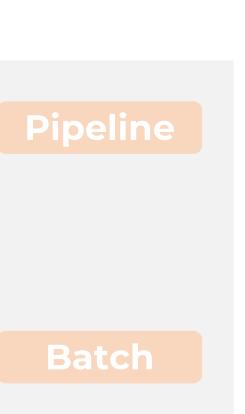






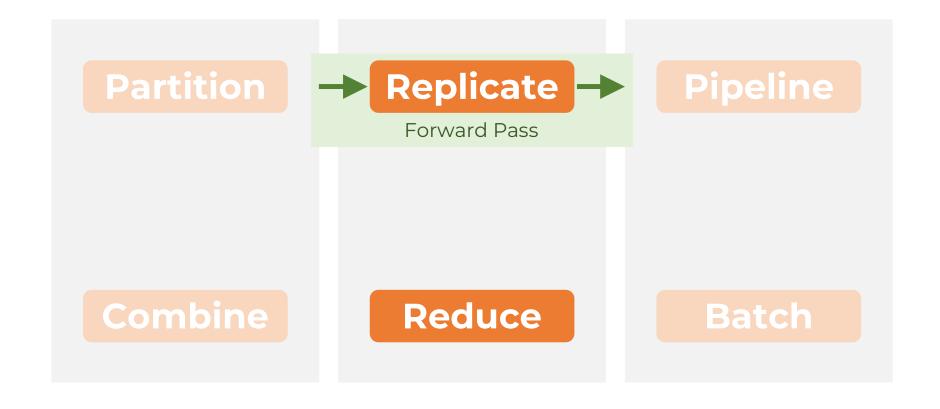


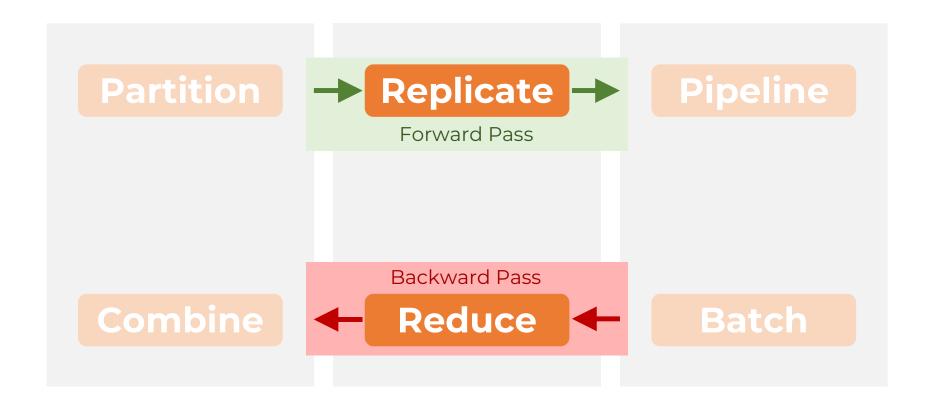




Replicate Reduce

Pipeline



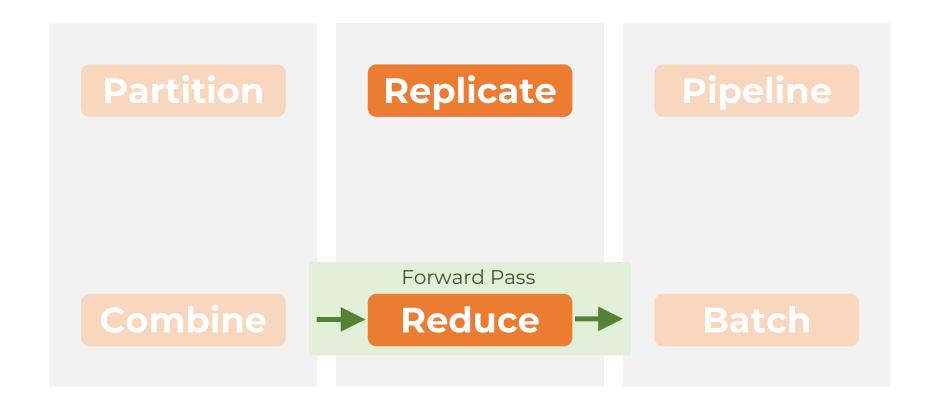


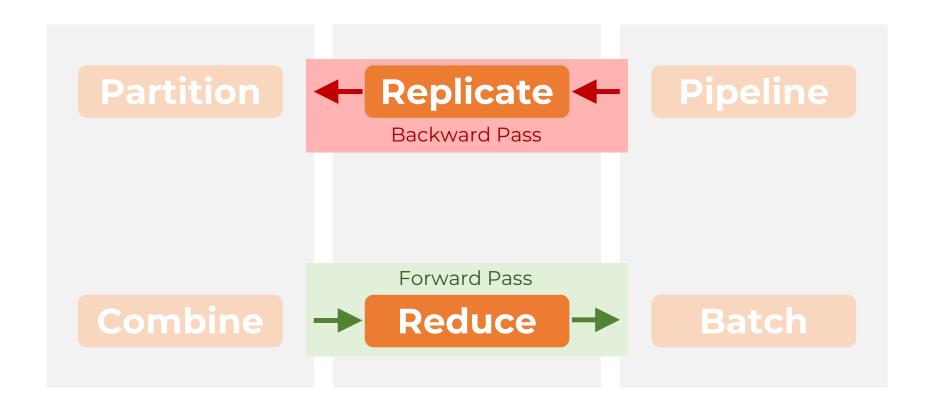
**Partition** Combine

Replicate

Pipeline

Reduce



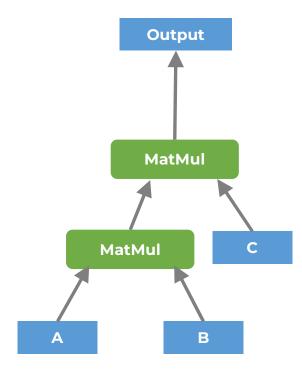


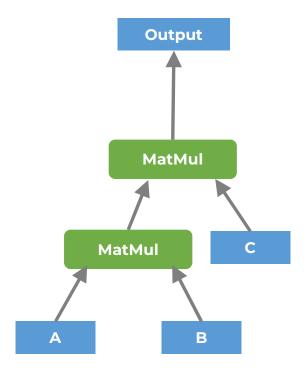
**Partition** Combine

Replicate

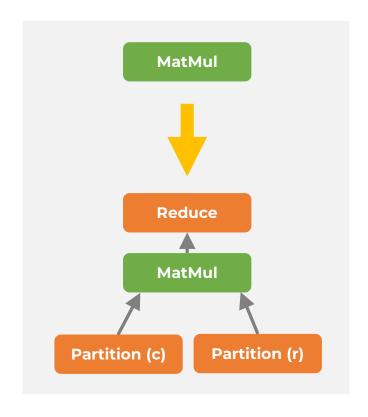
Pipeline

Reduce

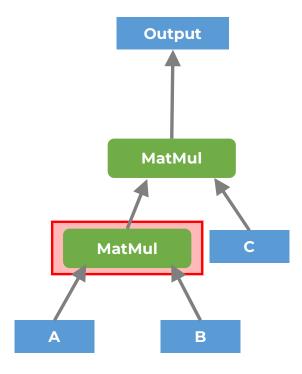




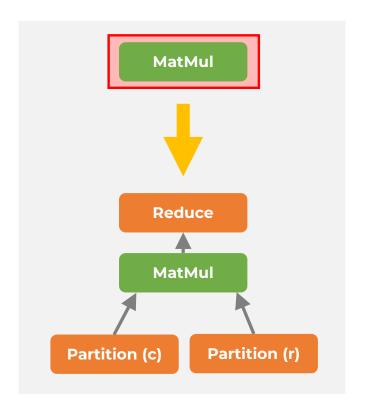
## Substitution



6

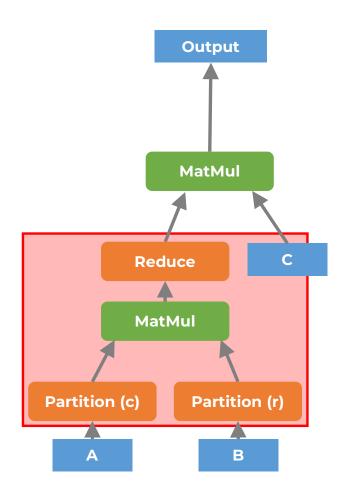


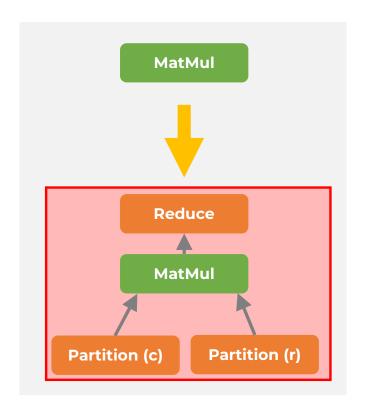
## Substitution

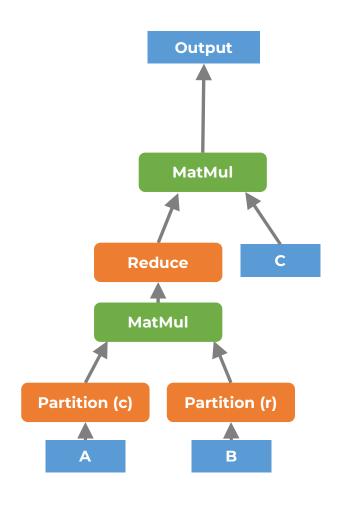


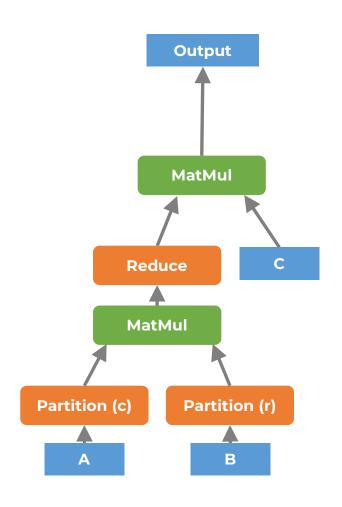
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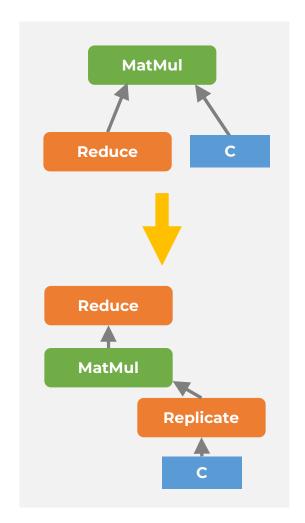
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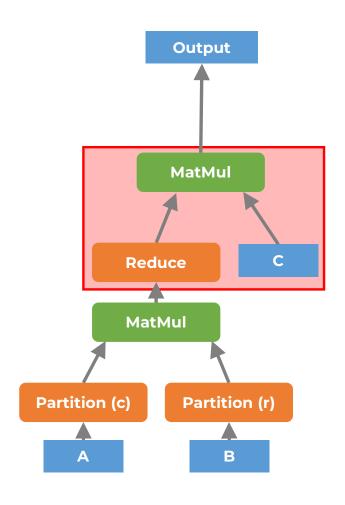


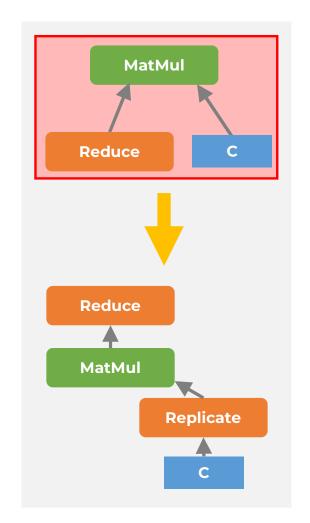


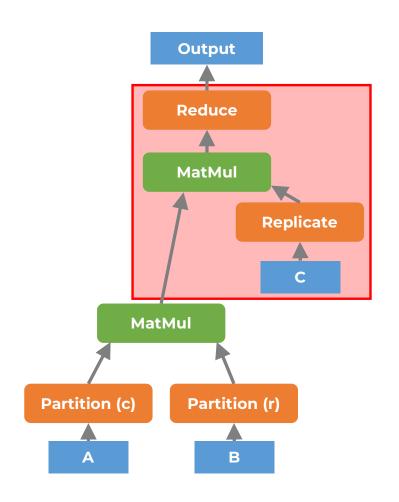


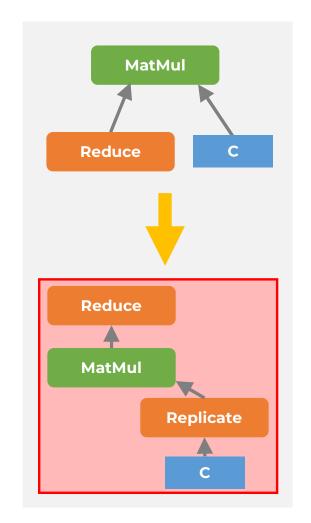


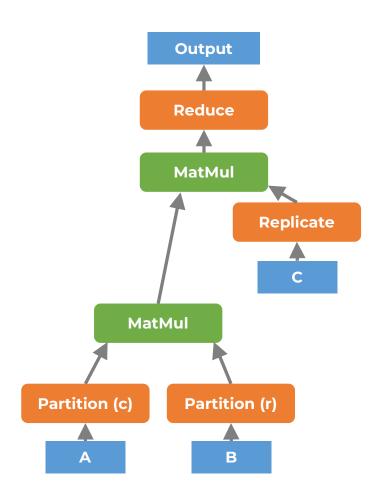












Automatically generate substitutions

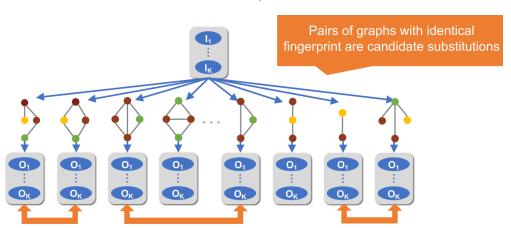
Automatically generate substitutions

New operators

Automatically generate substitutions

New operators

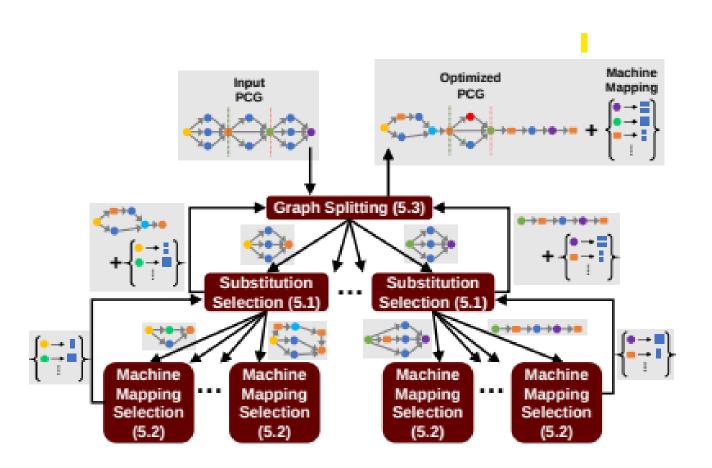
New forms of parallelism



## Explicitly represents communication

## Explicitly represents communication

## Concise



Algebraic Transformation

Parallelism Type

Parallelism Degree

**Device Mapping** 

Algebraic Transformation

Parallelism Type

Parallelism Degree

Device Mapping

Algebraic Transformation

Backtracking Search

Parallelism Type

Parallelism Degree

Device Mapping

Algebraic Transformation

Parallelism Type

Parallelism Degree

Backtracking Search



Device Mapping

Dynamic Programming

# Evaluation

## BERT-Large

(Language Modeling)

### Candle-UNO

(Precision Medicine)

MLP

(Regression)

**DLRM** 

XDL

(Recommendation)

ResNeXt-50

Inception-v3

(Computer Vision)

## BERT-Large

(Language Modeling)

#### Candle-UNO

(Precision Medicine)

MLP

(Regression)

DLRM

XDL

(Recommendation)

ResNeXt-50 Inception-v3

(Computer Vision)

#### **Baselines**





## BERT-Large

(Language Modeling)

### Candle-UNO

(Precision Medicine)

MLP

(Regression)

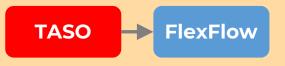
DLRM XDL

(Recommendation)

ResNeXt-50 Inception-v3

(Computer Vision)

#### **Baselines**



TASO+FlexFlow

(Sequential Optimization)

**Expert-Designed** 

or

Data Parallel



### BERT-Large

(Language Modeling)

#### Candle-UNO

(Precision Medicine)

MLP

(Regression)

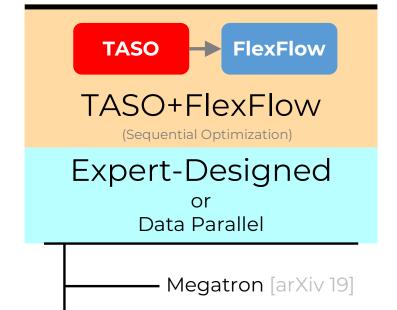
## DLRM XDL

(Recommendation)

ResNeXt-50 Inception-v3

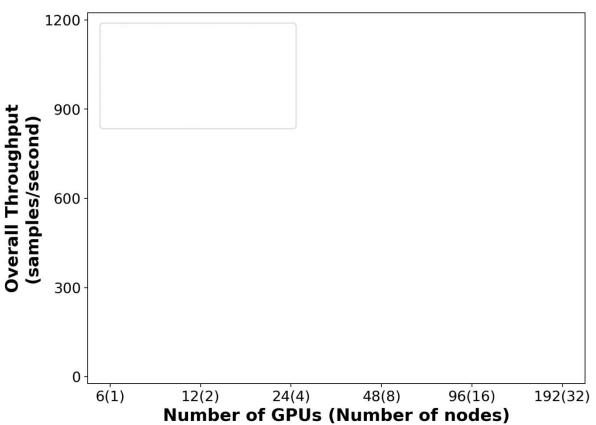
(Computer Vision)

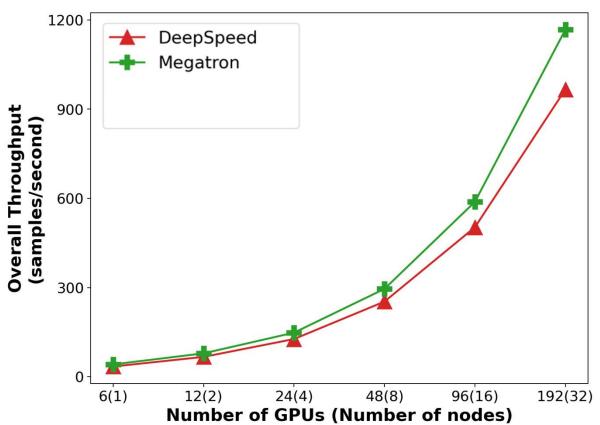
#### **Baselines**

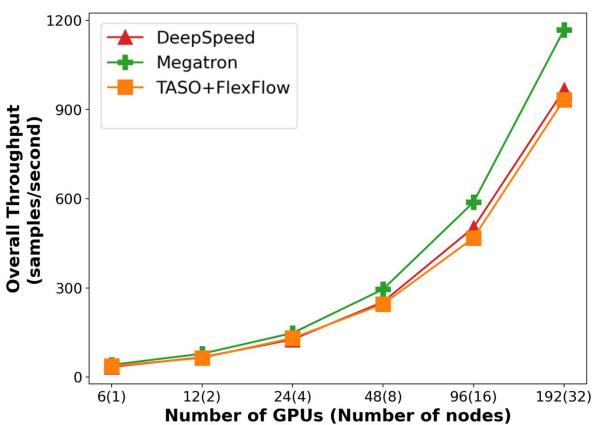


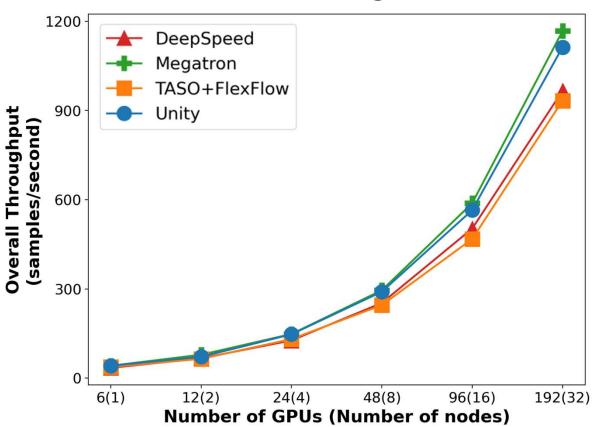
DeepSpeed [arXiv 19]



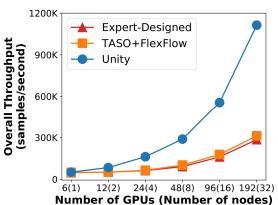




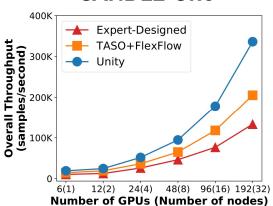




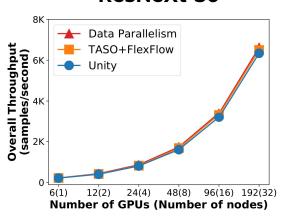
#### **DLRM**



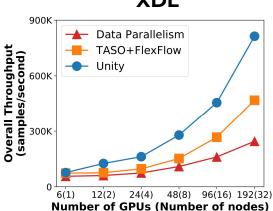
#### **CANDLE-Uno**



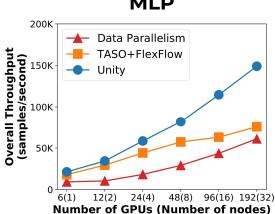
ResNeXt-50



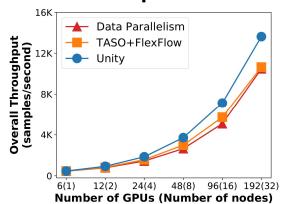
#### **XDL**



#### **MLP**



#### Inception-v3



## https://github.com/flexflow/FlexFlow

kadinzhang Added tests	for Linear operator in align/linear (#264)	488423# 19 days ago 1,339 commits	A distributed deep learning framework that supports flexible parallelization
			strategies.  II Readme  48 Apacte-2.0 license  Code of conduct  458 stars  21 watching  V 100 forks
align			
bootcamp_demo			
cmake			
conda			
			Releases 6
docker			





















