

# Deploying AI at Meta Scale

BUILDING DEVELOPER-FRIENDLY HIGH-PERFORMANCE SCALABLE SYSTEMS FOR AI + PYTORCH

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Community Statistics



**3.65B**

people use at least one of our services  
monthly, approximately



**2.91B**

monthly active users  
on Facebook



**700M**

people use Augmented Reality  
across our apps and devices monthly



Snapshot of Data Center Footprint

>100

Data centers

~50M

Square feet

100%

Renewable energy

New Albany, Ohio

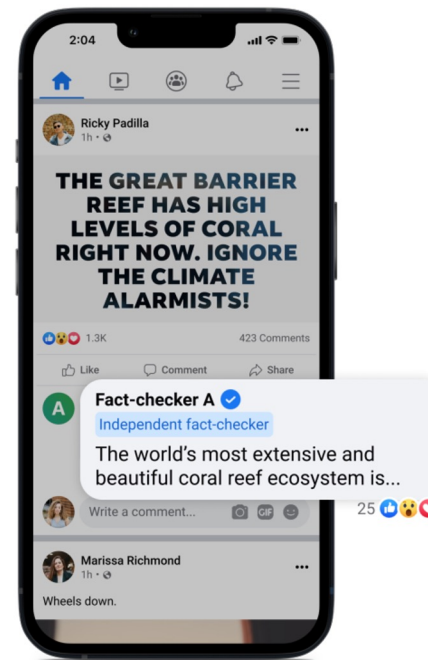


How we use AI...

AI AT META

## Content Understanding

- Identify and eliminate inappropriate content before being viewed
- Leverages several model types, including computer vision, image classification, and natural language processing



**~250M**  
Violating Content (Q2)

**95%+**  
Actioned on by AI

AI AT META

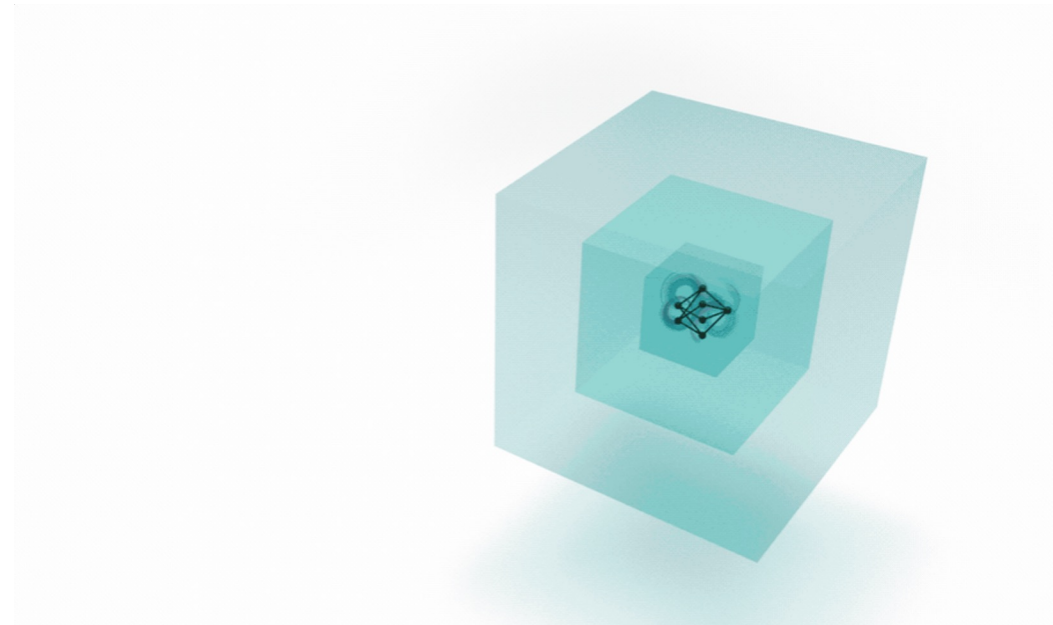
# Natural Language Processing

- Open Pretrained Transformer (OPT-175B)
- XMLR

Sources:

<https://ai.facebook.com/research/no-language-left-behind/>

<https://ai.facebook.com/blog/democratizing-access-to-large-scale-language-models-with-opt-175b/>



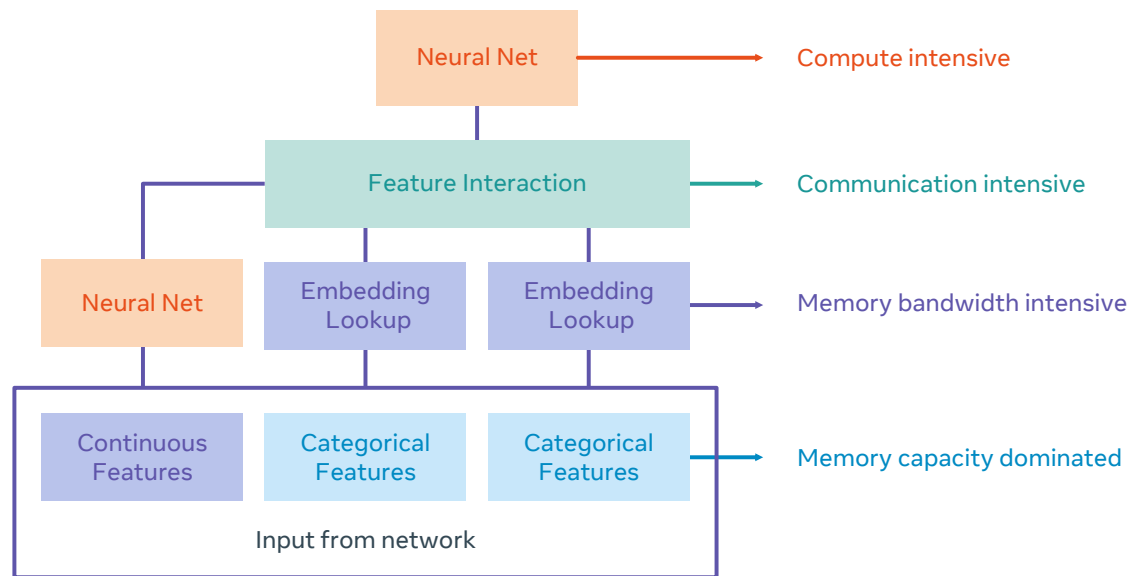
AI AT META

# Recommendation & Personalization

- Deep Learning Recommendation Model (DLRM)
- TBSM, DHEN

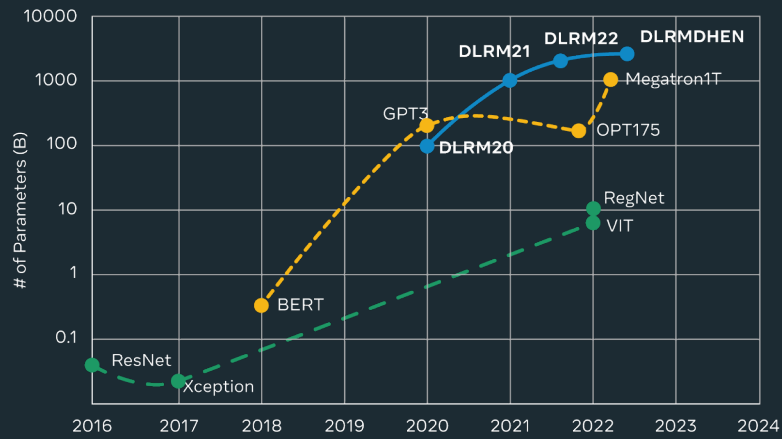
Source:

<https://ai.facebook.com/blog/dlrm-an-advanced-open-source-deep-learning-recommendation-model/>

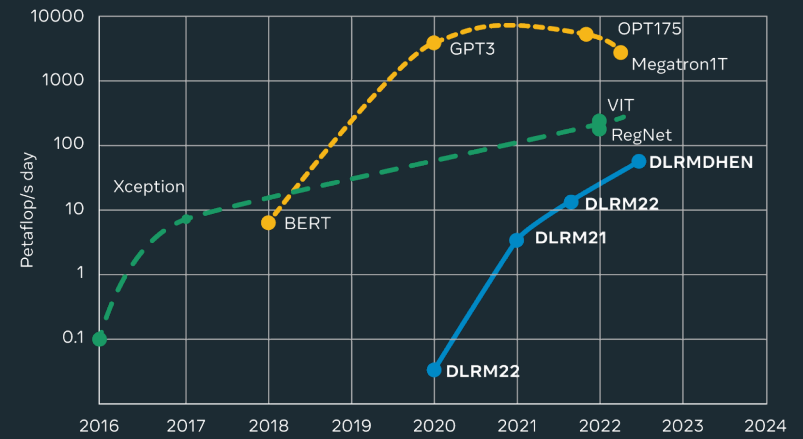


## DEEP LEARNING WORKLOADS - CHARACTERISTICS

### SIZE



### COMPUTE





AI IS POWERING EVERYTHING WE DO: AI-related statistics



**6B**

Training images



**20B+**

Translations per day



**200T+**

Predictions per day

How we develop AI...

THE LANGUAGE OF AI

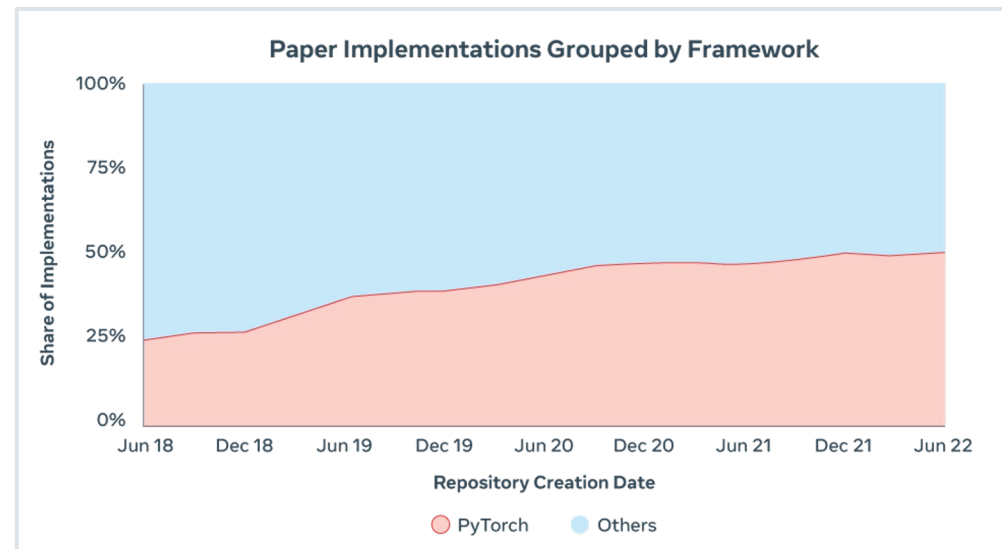


ALL IN ON AI WITH PYTORCH

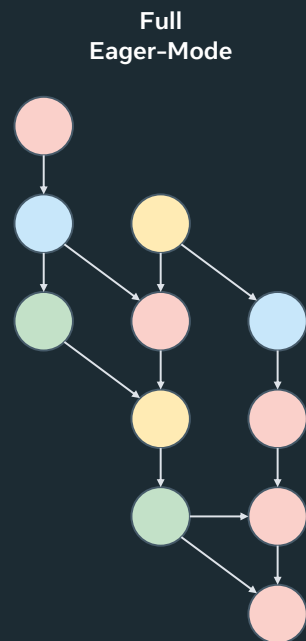
## Rapid research to production

- Benefit of deploying state-of-art models fast can be huge

Source:  
<https://paperswithcode.com/trends>



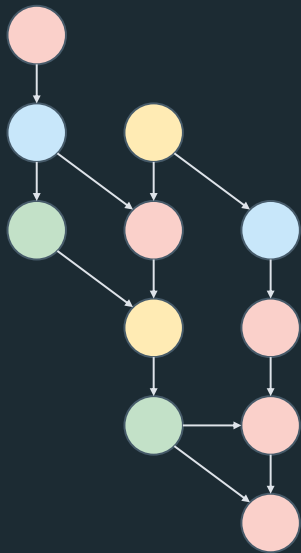
## PYTORCH: A DEVELOPER-FIRST MINDSET



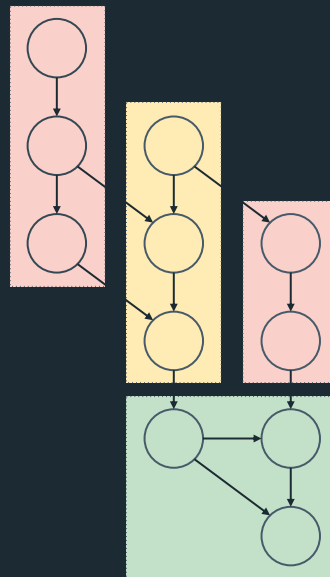
```
for epoch in range (max_epochs):
    for model_target, model_input in enumerate(training_data):
        if numpy.random.randint(100) > 90: # 10% noise
            model_target = torch.from_numpy(numpy.random.randint(2))
        model_output = dlrn(model_input)
        model_loss = torch.nn.BCELoss(model_output, model_target)
        model_loss.backward()
        optimizer.step()
        print("BCE loss " + str(model_loss))
        matplotlib.pyplot.plot(...)
```

# DEVELOPER EFFICIENCY VS. PERFORMANCE

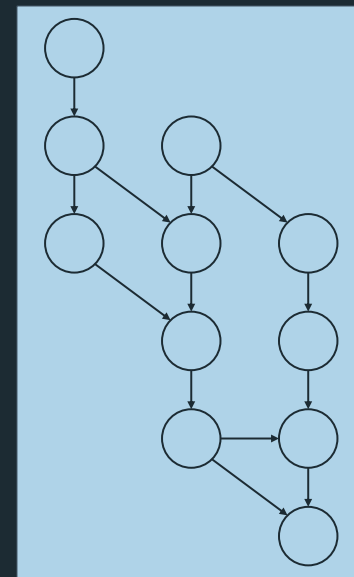
Full Eager-Mode



Partial Eager-Mode

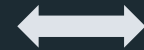


Graph Mode



Developer Efficiency

Model Performance



How we enable AI...

## OPTIMIZING THE AI SYSTEM FOR PYTORCH



### Programmable

Easy operator authoring → new compute primitives for model innovation.



### Dynamic

Fast operator launch.  
Dynamic tensor shapes, memory allocation,  
easy to prototype



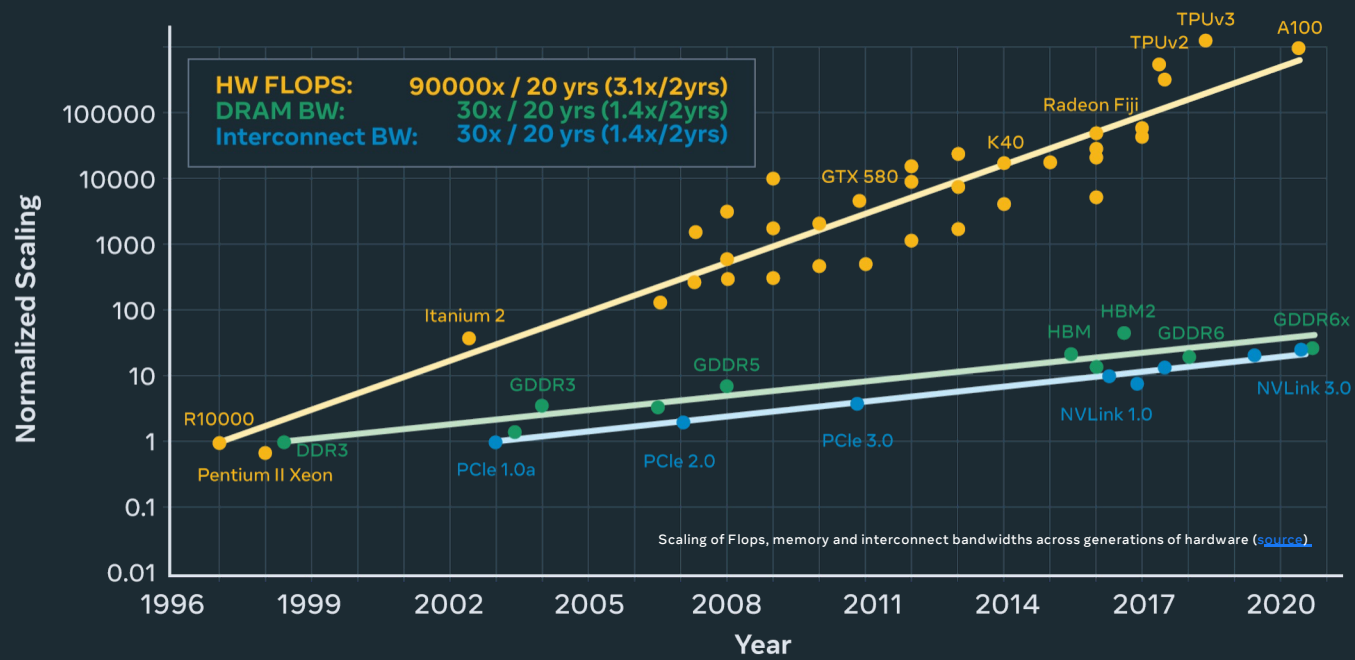
### Scalable & Tunable

Future-proof.  
Balance between  
compute, memory, network.

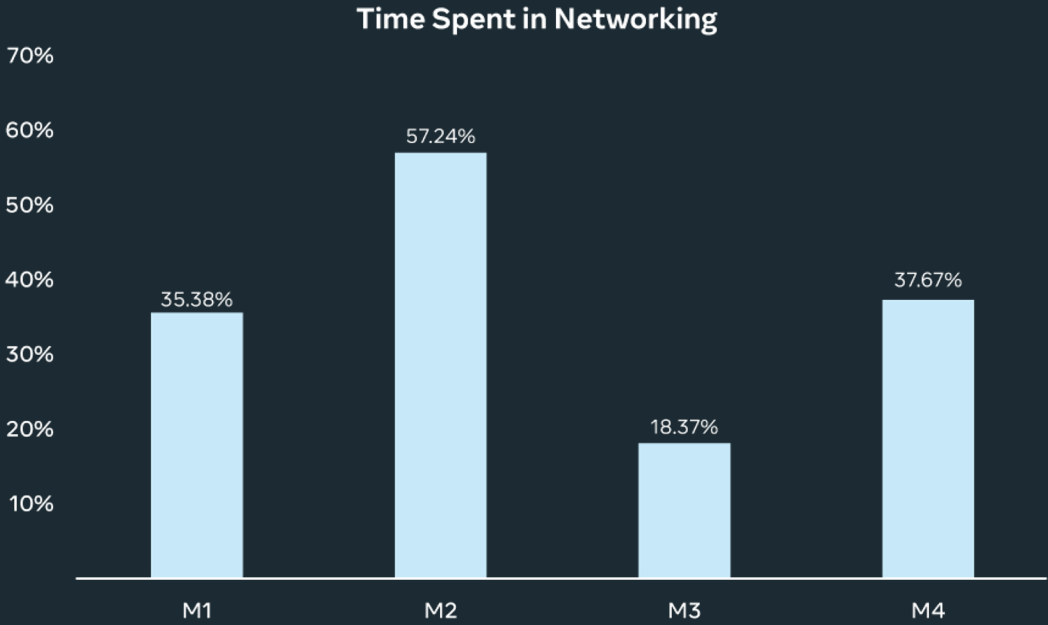


NETWORK IS LAGGING ON EVOLUTION CURVE

Scaling of Peak hardware FLOPS, and Memory/Interconnect Bandwidth



NETWORK I/O IS KEY FOR RECOMMENDATION WORKLOADS

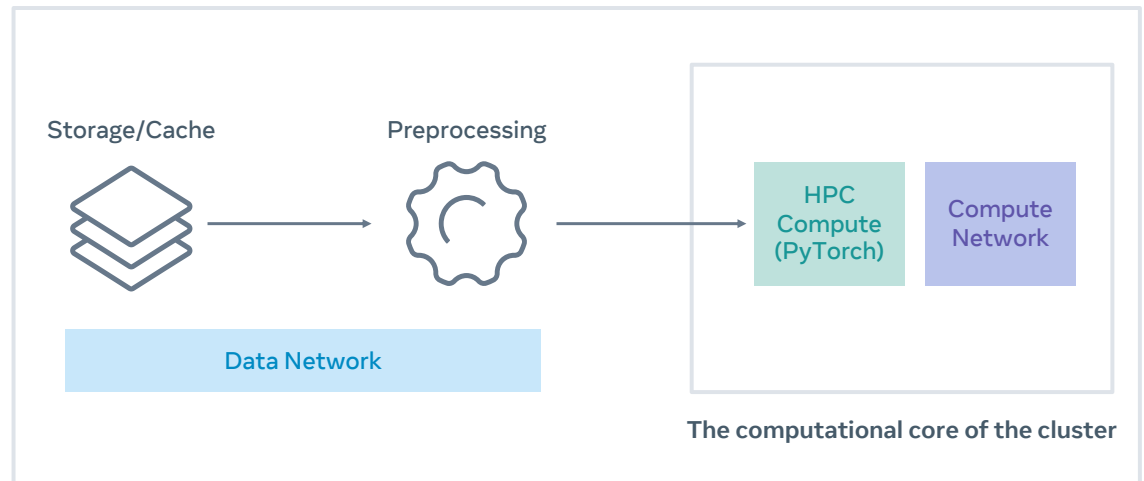


Ranking requires high injection & bisection bandwidth

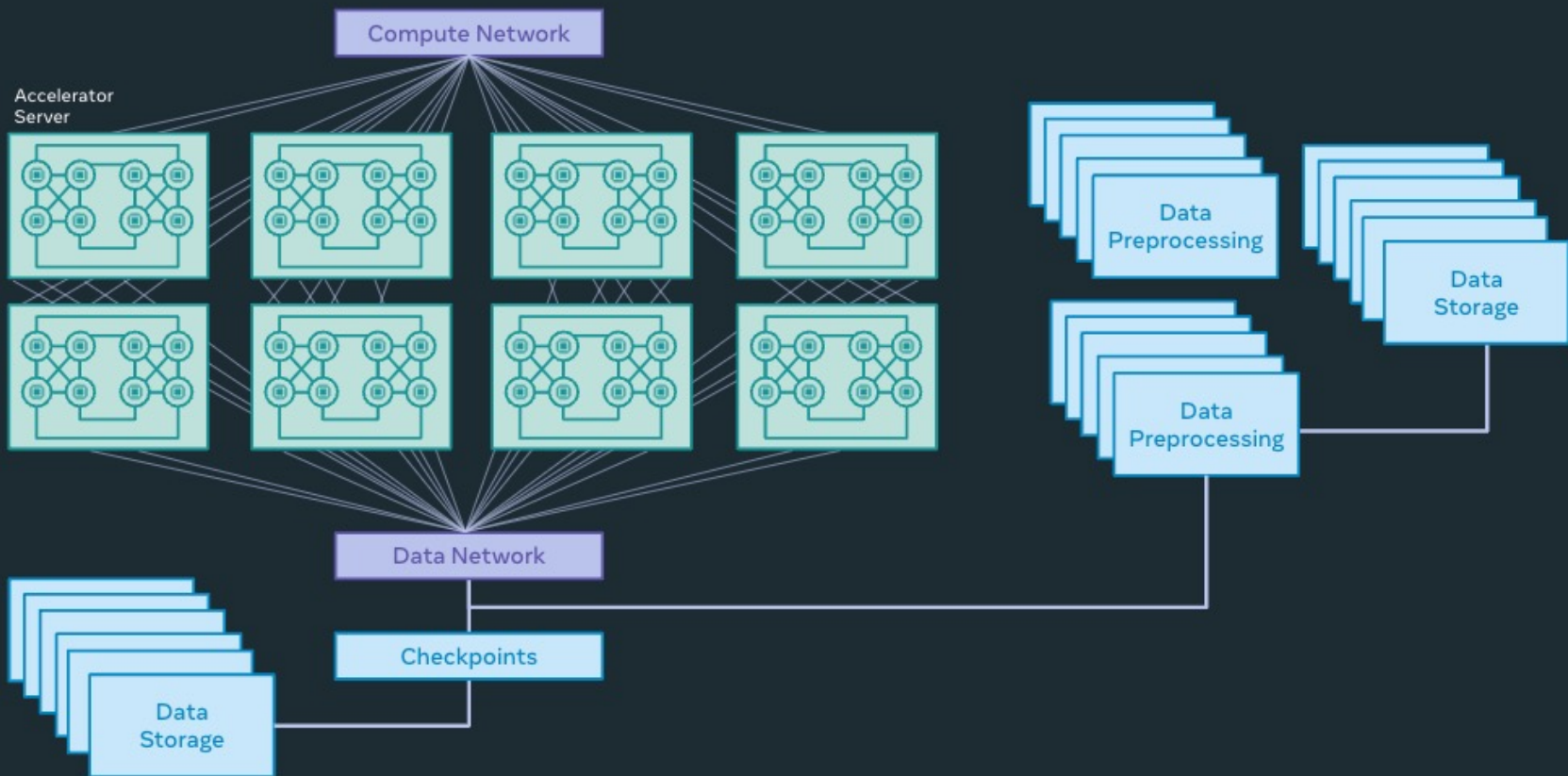
## WHAT MAKES AN AI TRAINING CLUSTER?

- Storage Cache
- Data ingestion
- Compute Nodes
- Compute fabric

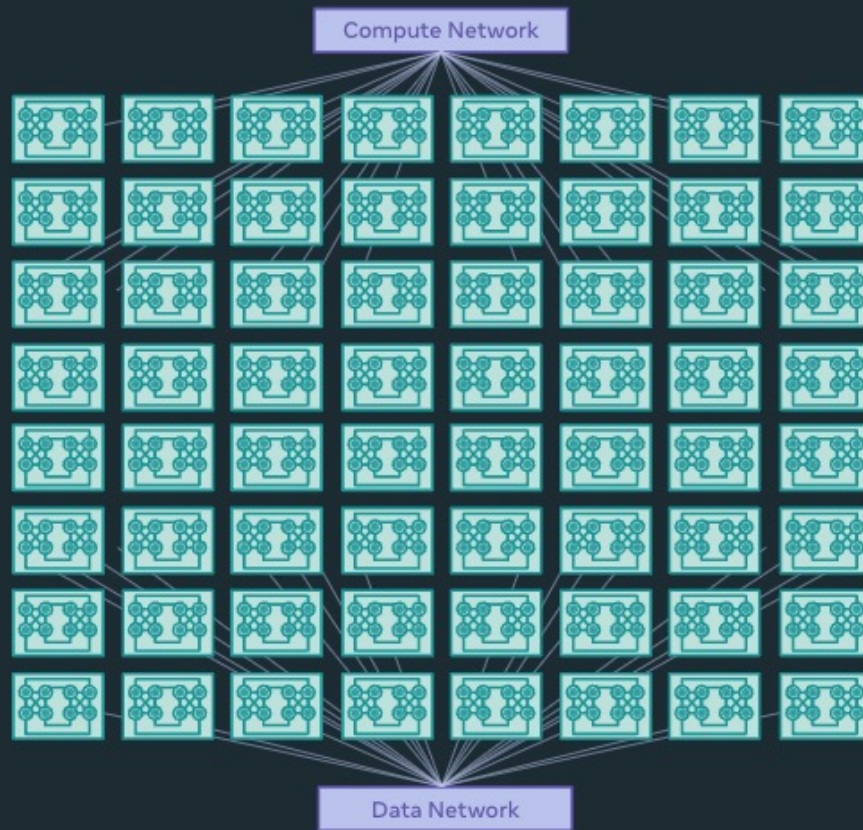
## PyTorch AI Training Cluster



# THE PYTORCH AI TRAINING CLUSTER



THE PYTORCH AI TRAINING CLUSTER OF THE FUTURE (2025+)



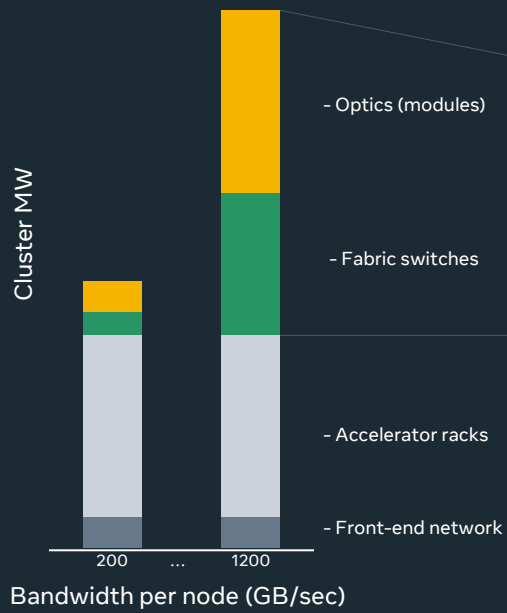
~4K  
Accelerators

~1TB/s  
of compute  
network per accelerator

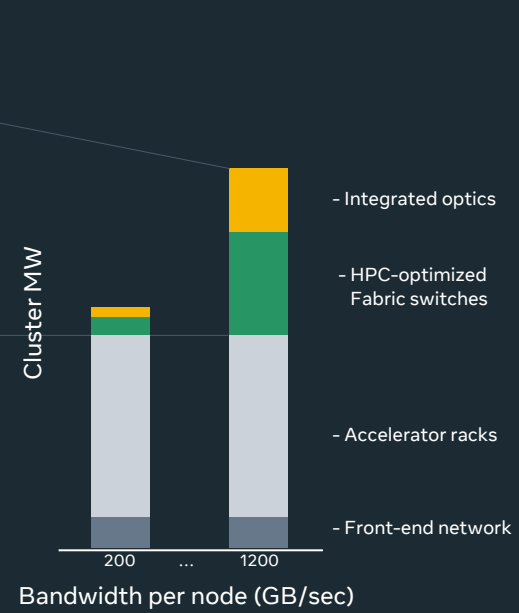
The cluster fabric is **Non-blocking**

# THE PROMISE OF OPTICAL IO

## Power breakdown for a cluster (optical modules)

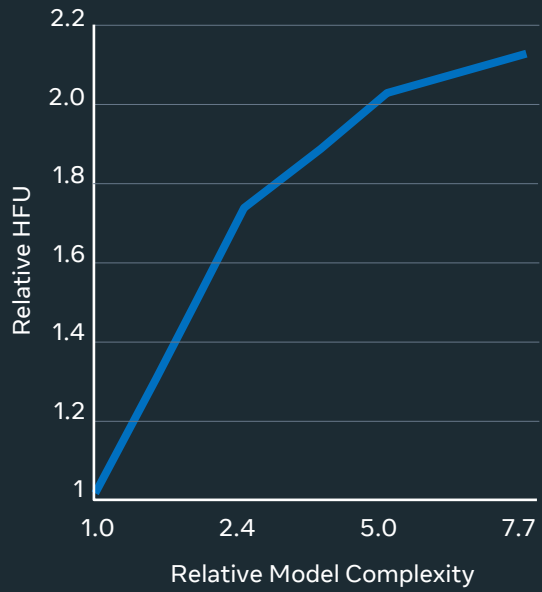


## Power breakdown for a cluster (CPO)

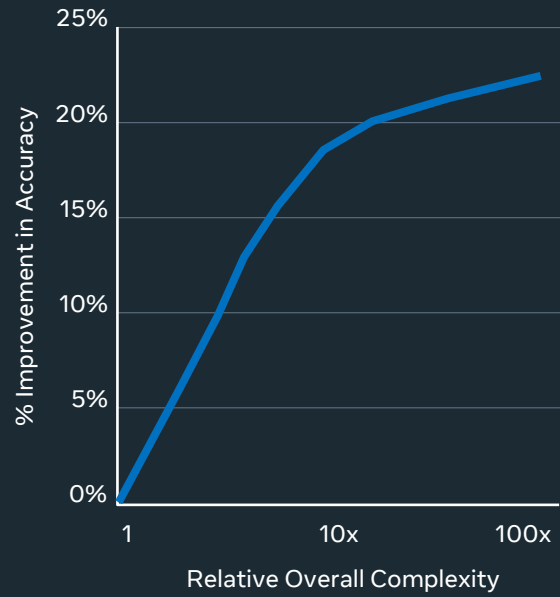


## THE BENEFIT OF A PYTORCH AI TRAINING CLUSTER

### DLRM - DHEN



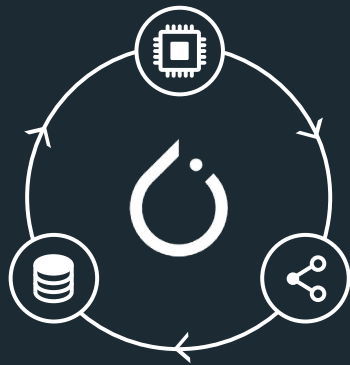
### Improving Content Relevance



**+24%**

improvement in content  
relevance over 2021-22

CONCLUSION / TAKEAWAYS



Co-design  
for PyTorch



Flexibly balance compute,  
memory, and network



Plan for  
Innovation as  
AI evolves



