# GPUnet: networking abstractions for GPU programs

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### What

A socket API for programs running on GPU

## Why

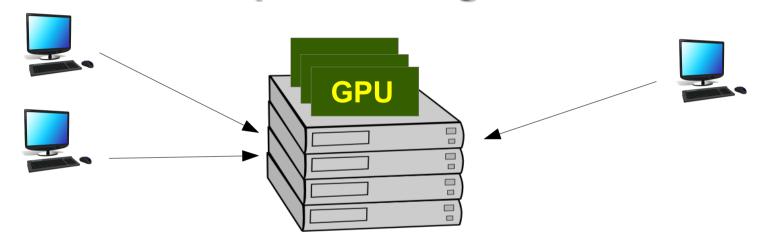
GPU-accelerated servers are hard to build

### Results

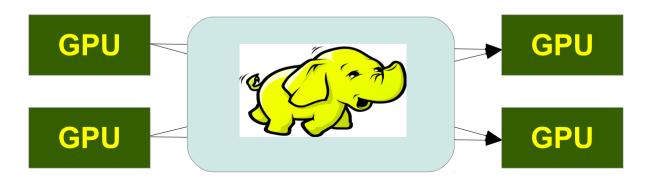
GPU vs. CPU 50% throughput, 60% latency, ½ LOC

## Motivation: GPU-accelerated networking applications

#### **Data processing server**



#### **MapReduce**



# Recent GPU-accelerated networking applications

SSLShader (Jang 2011), GPU MapReduce (Stuart 2011), Deep Neural Networks (Coates 2013), Dandelion (Rossbach 2013), Rhythm (Agrawal 2014) ...

# Recent GPU-accelerated networking applications

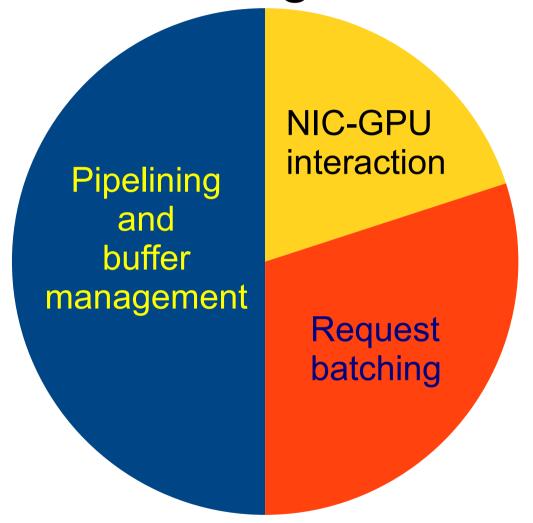
SSLShader (Jang 2011), GPU MapReduce (Stuart 2011), Deep Neural Networks (Coates 2013), Dandelion (Rossbach 2013), Rhythm (Agrawal 2014) ...

## required heroic efforts



Mark Silberstein - EE, Technion

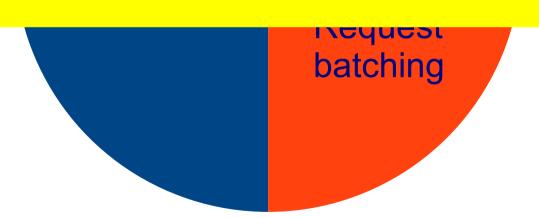
# GPU-accelerated networking apps: Recurring themes



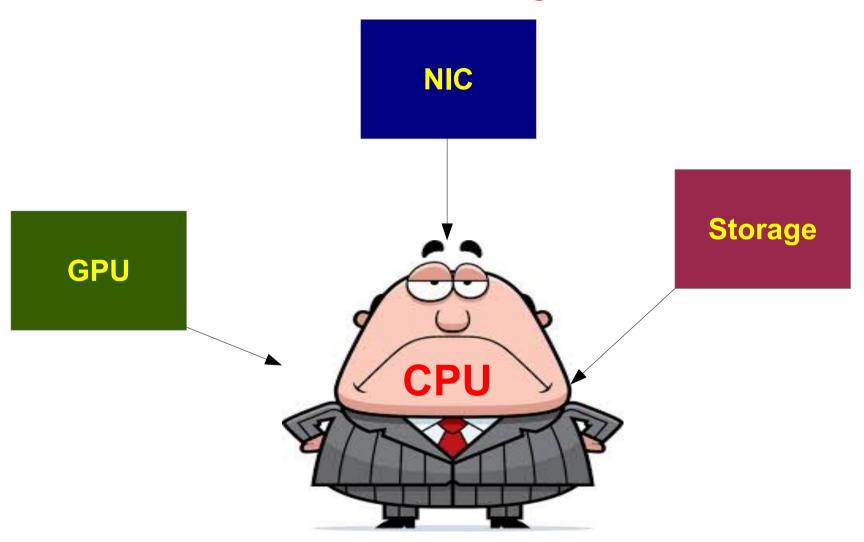
# GPU-accelerated networking apps: Recurring themes



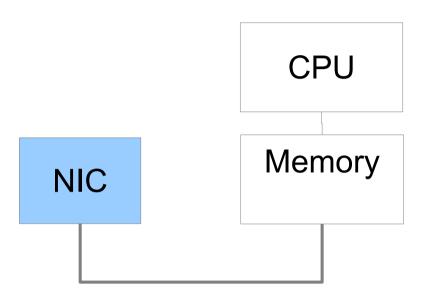
We will sidestep these problems



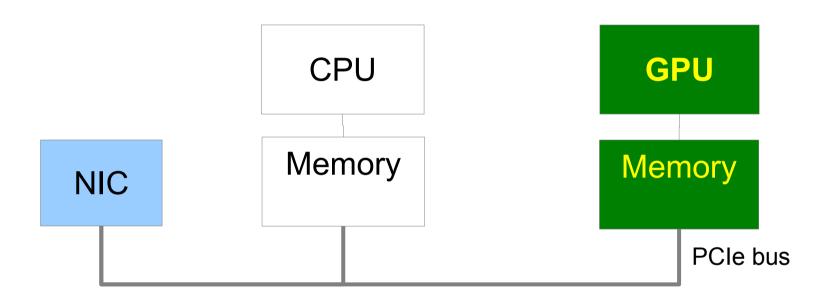
# The **real** problem: CPU is the **only** boss



## Example: CPU server

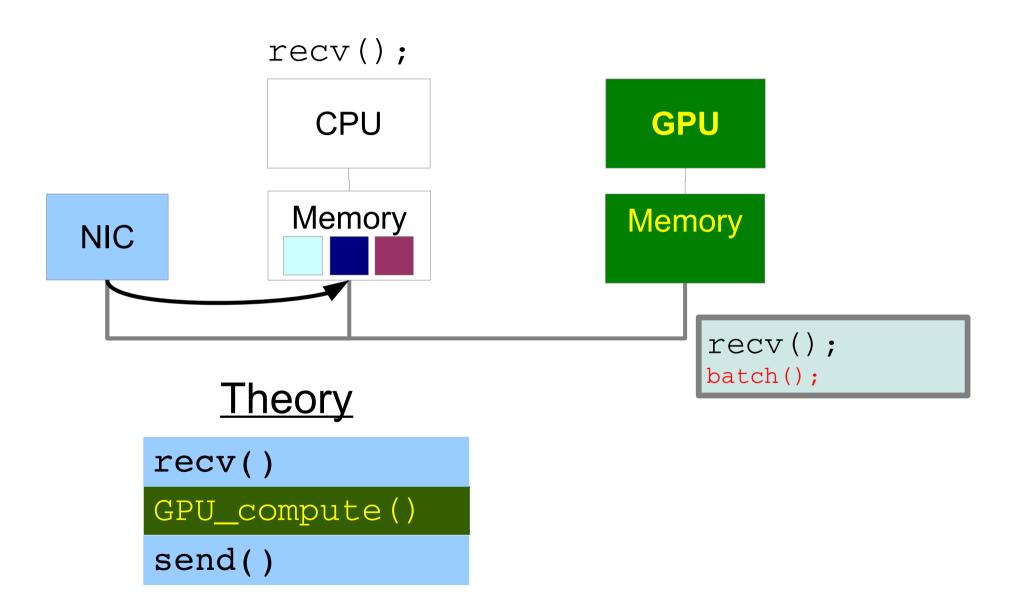


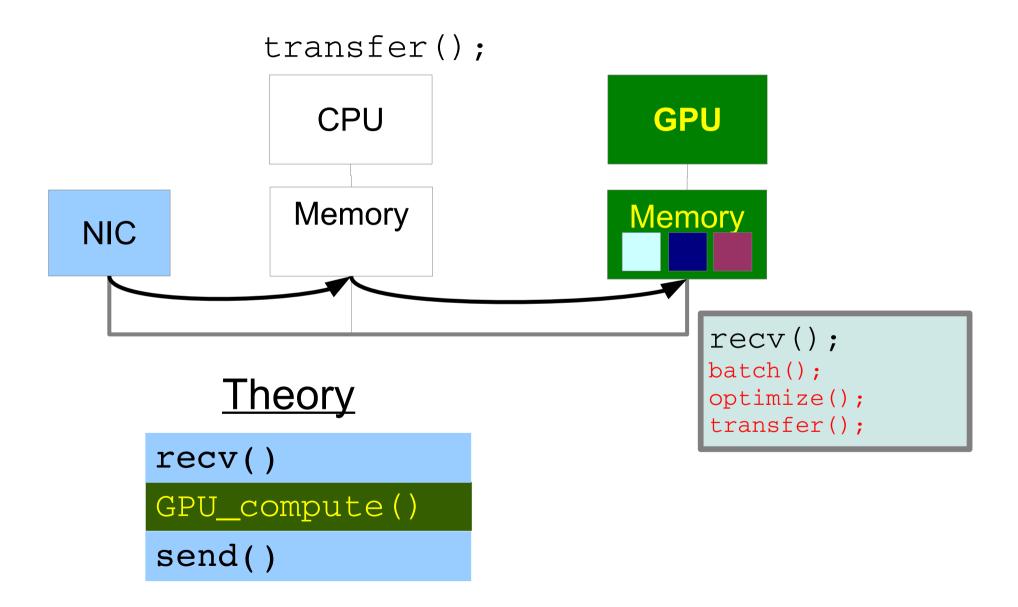
```
recv()
compute()
send()
```

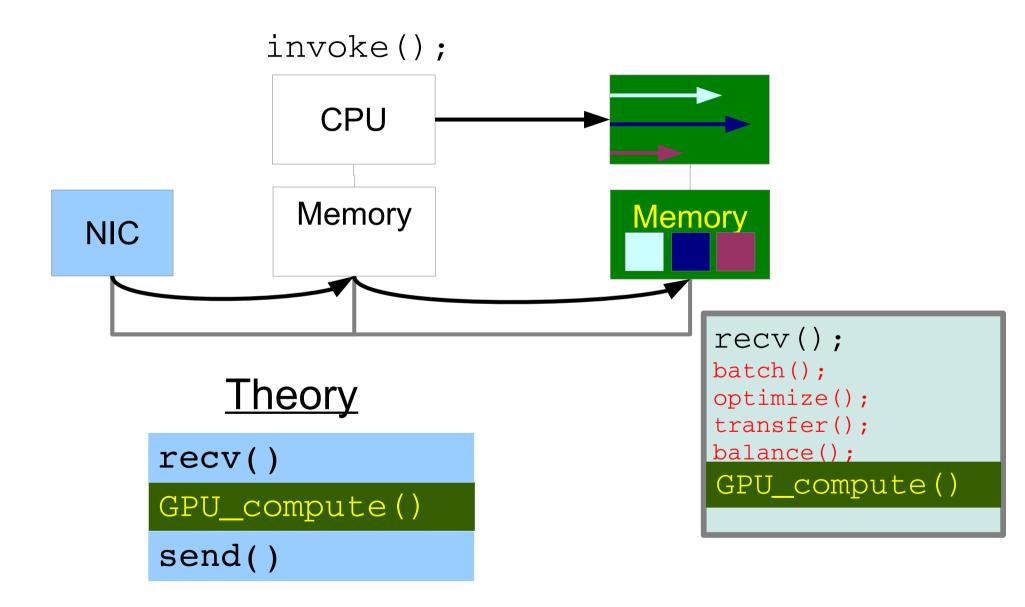


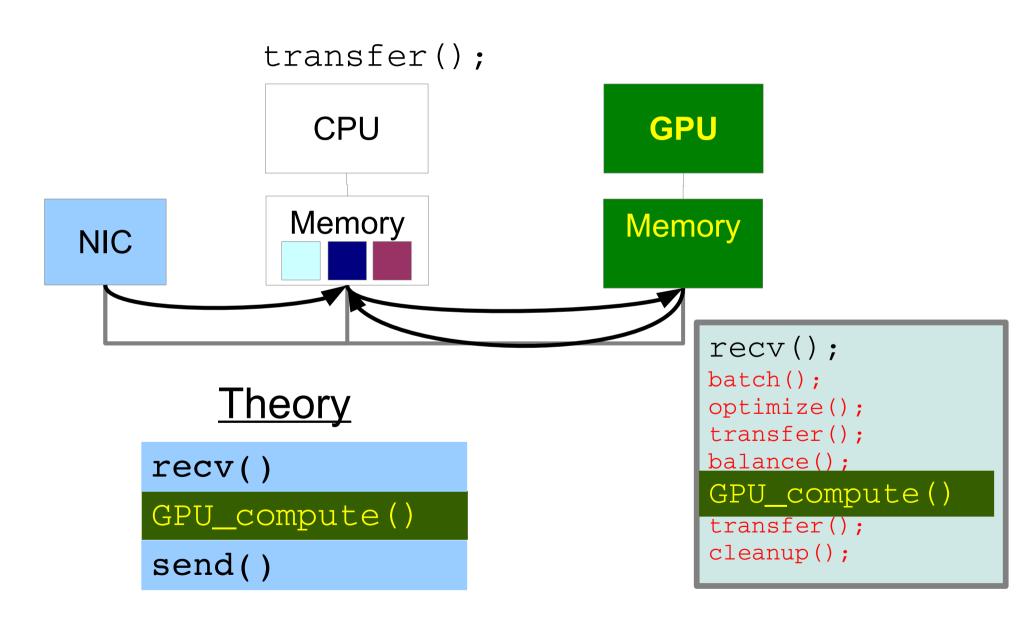
### **Theory**

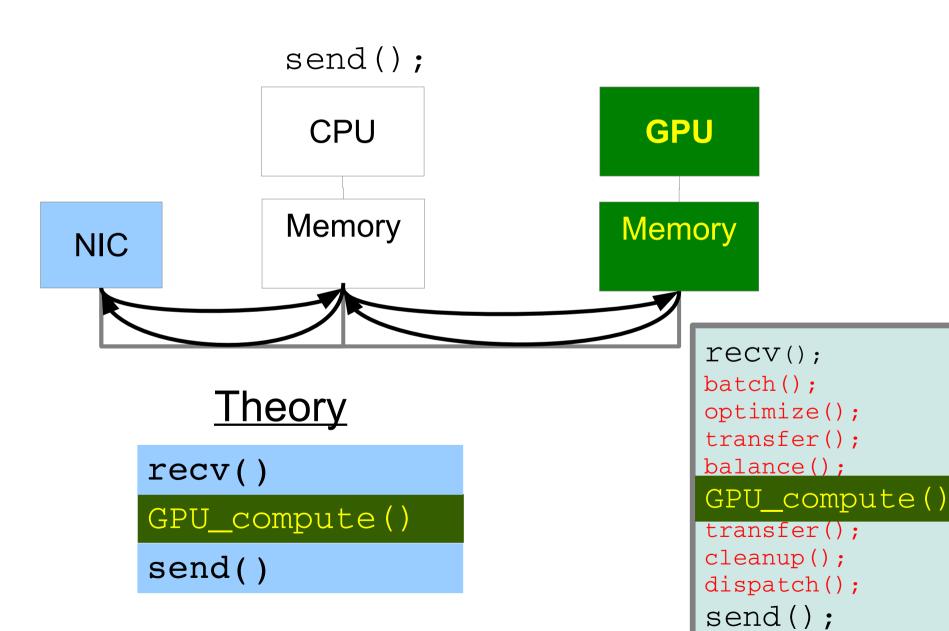
```
recv()
GPU_compute()
send()
```





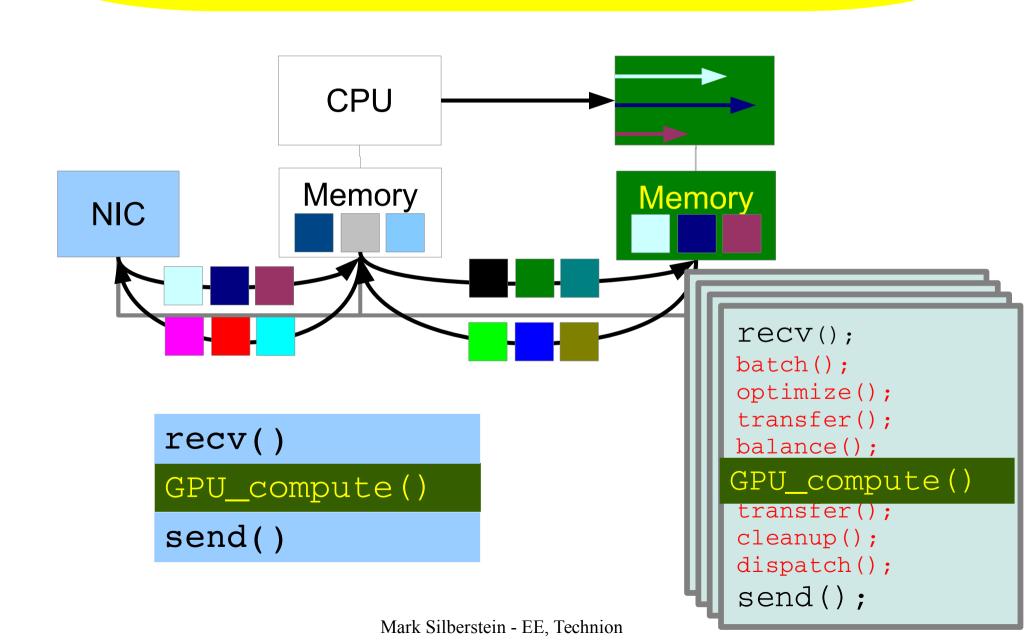






### Aggressive pipelining

Double buffering, asynchrony, multithreading



## This code is for a CPU to manage a GPU

```
batch();
optimize();
transfer();
balance();
GPU_compute()
transfer();
cleanup();
dispatch();
```

## GPUs are not co-processors

GPUs are peer-processors

## They need I/O abstractions

File system I/O – [GPUfs ASPLOS13] Network I/O – this work

# GPUnet: socket API for GPUs Application view

```
node0.technion.ac.il

GPU native server

socket(AF_INET,SOCK_STREAM);
listen(:2340)

GPUnet
```

Network

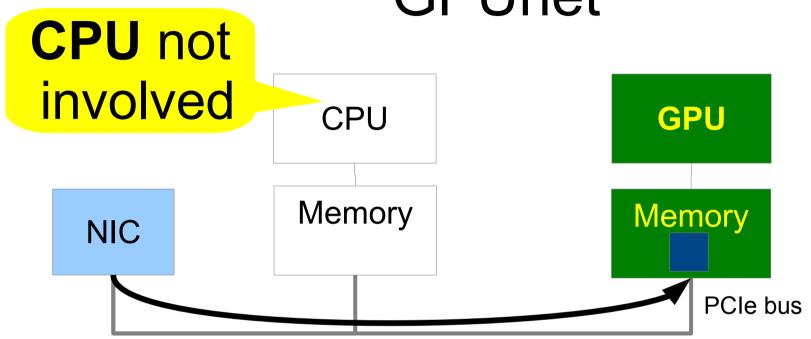
#### **CPU** client

```
socket(AF_INET, SOCK_STREAM);
connect("node0:2340")
```

#### GPU *native* client

```
socket(AF_INET,SOCK_STREAM);
connect("node0:2340");
GPUnet
```

# GPU-accelerated server with GPUnet



```
recv()
GPU_compute()
send()
```

# GPU-accelerated server with GPUnet

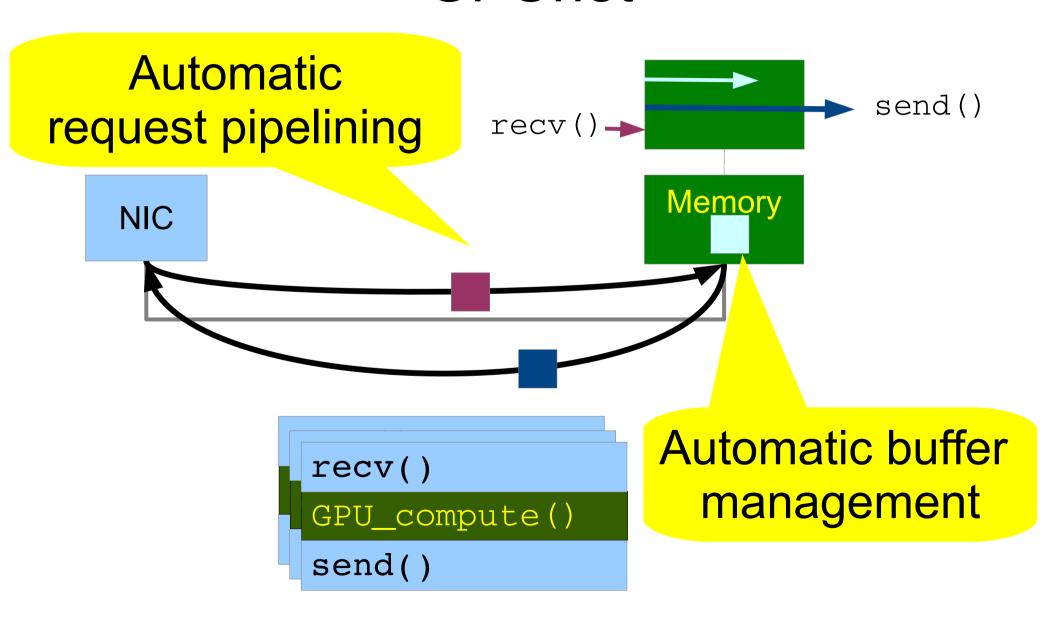


```
recv()
GPU_compute()
send()
```

## GPU-accelerated server with **GPUnet**

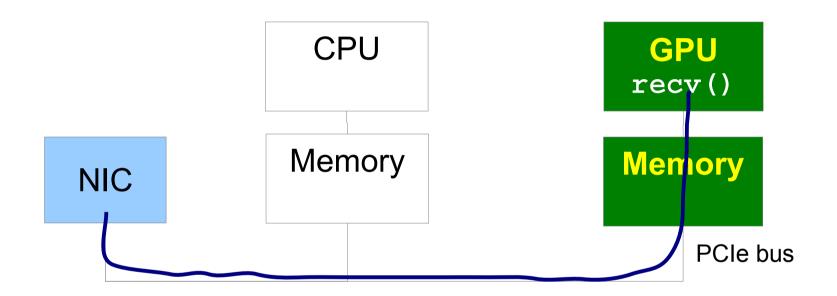
No request batching send() recv() **Memory** NIC recv() GPU\_compute() send()

# GPU-accelerated server with GPUnet



# Building a socket abstraction for GPUs

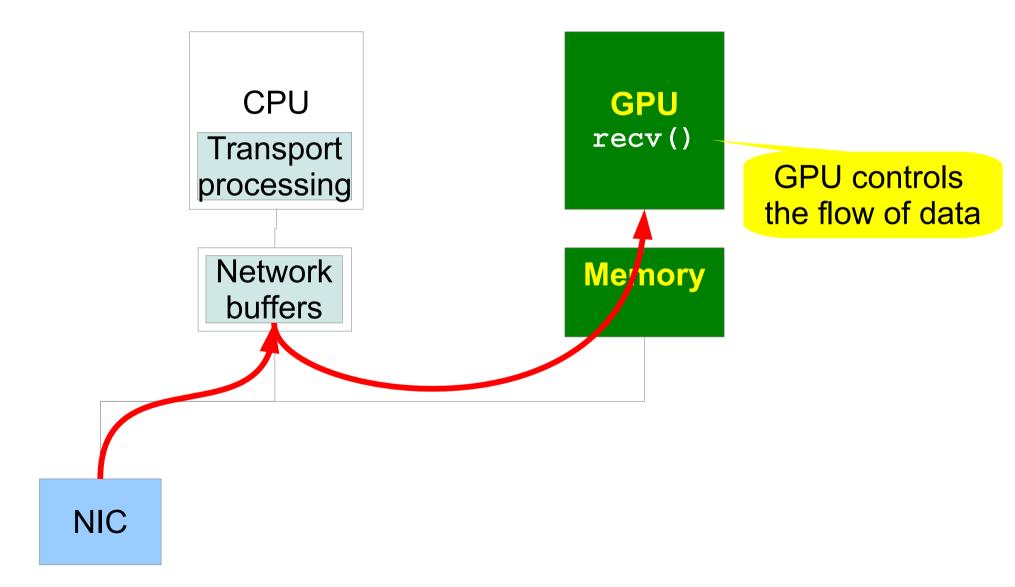
### Goals



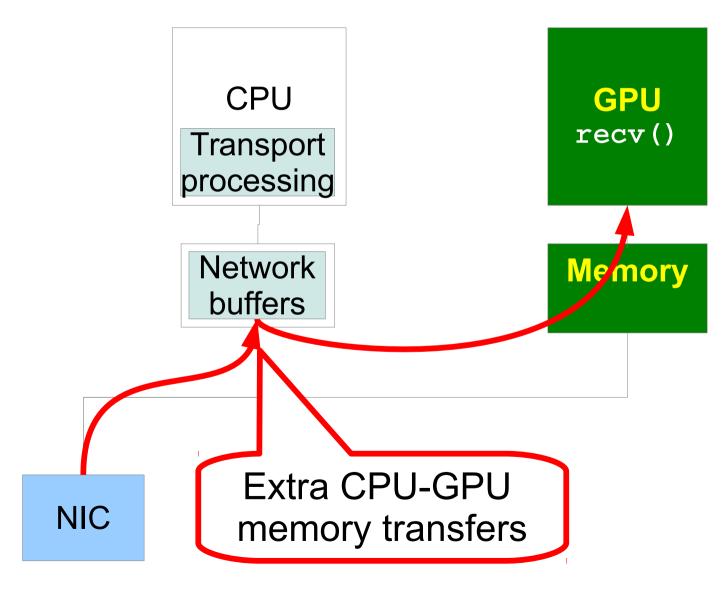
Simplicity
Reliable streaming
abstraction for GPUs

Performance NIC → GPU data path optimizations

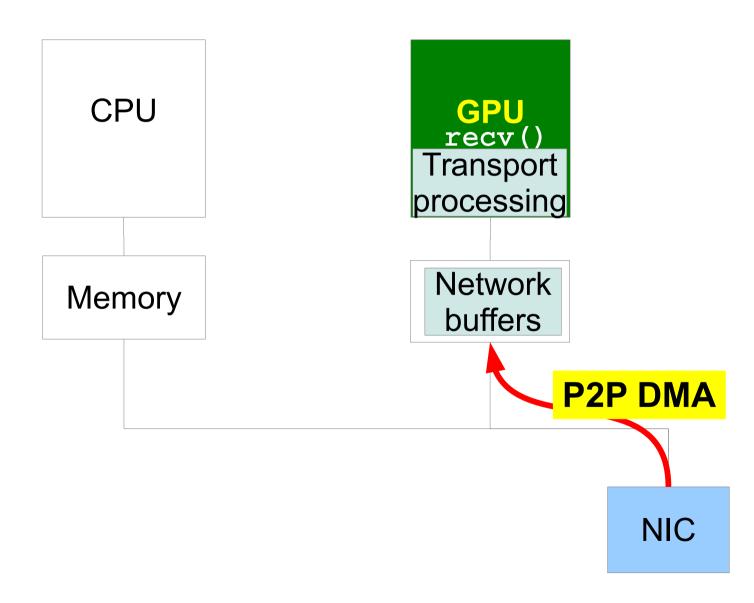
# Design option 1: Transport layer processing on CPU



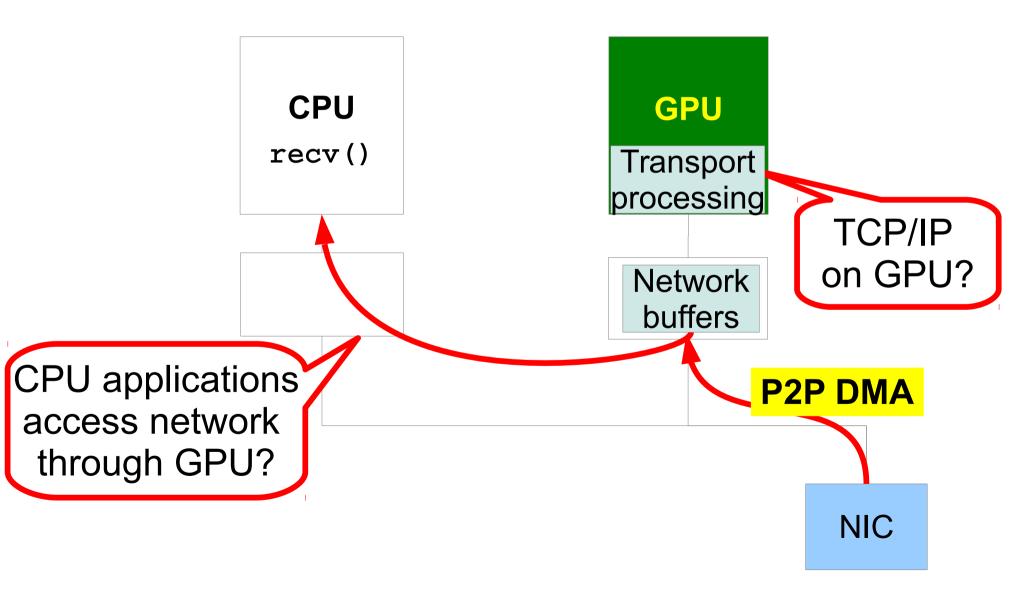
# Design option 1: Transport layer processing on CPU



# Design option 2: Transport layer processing on GPU



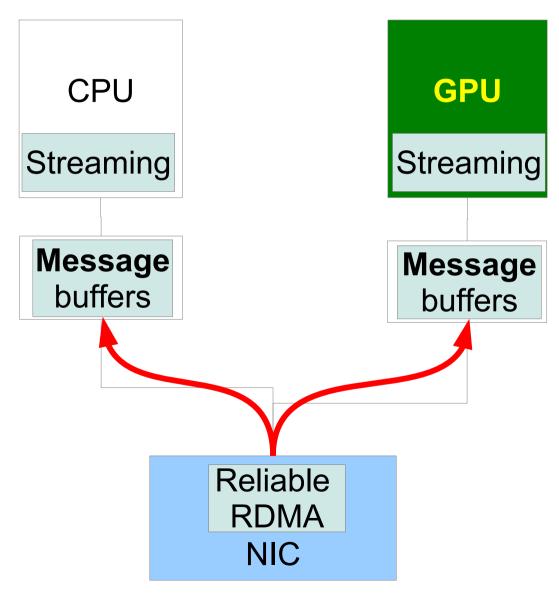
# Design option 2: Transport layer processing on GPU



### Not CPU, Not GPU

We need help from NIC hardware

# RDMA: offloading transport layer processing to NIC



## **GPUnet layers**

**GPU Socket API** 

Reliable in-order streaming

Reliable channel

RDMA Transports *Infiniband* 

Non-RDMA Transports
UNIX Domain Socket, TCP/IP

## **GPUnet layers**

Simplicity

**GPU Socket API** 

Reliable in-order streaming

Reliable channel

RDMA Transports *Infiniband* 

**GPU** 

Non-RDMA Transports
UNIX Domain Socket, TCP/IP

NIC CPU

Performance

## See the paper for

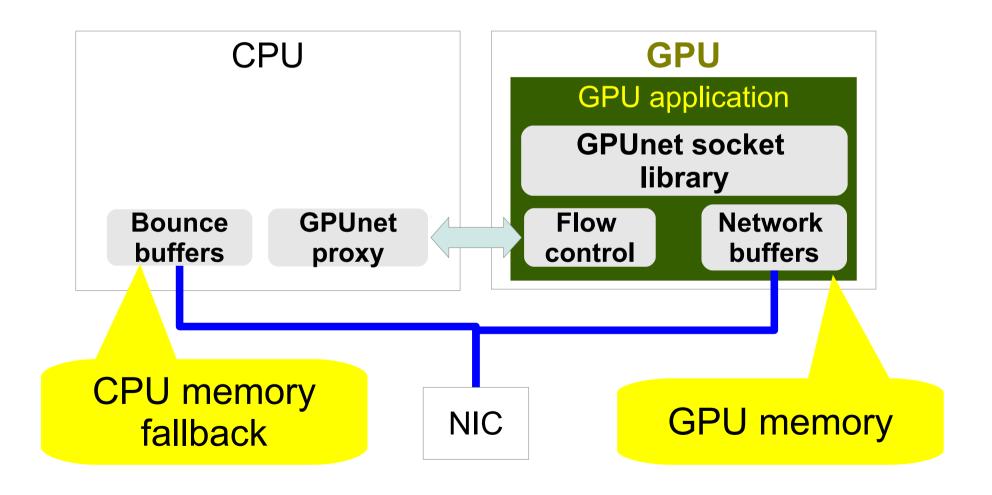
- Coalesced API calls
- Latency-optimized GPU-CPU flow control
- Memory management
- Bounce buffers
- Non-RDMA support
- GPU performance optimizations

## Implementation

- Standard API calls, blocking/nonblocking
- libGPUnet.a: AF\_INET, Streaming over Infiniband RDMA
  - Fully compatible with CPU rsocket library

 libUNIXnet.a: AF\_LOCAL: Unix Domain Sockets support for inter GPU/CPU-GPU

## Implementation

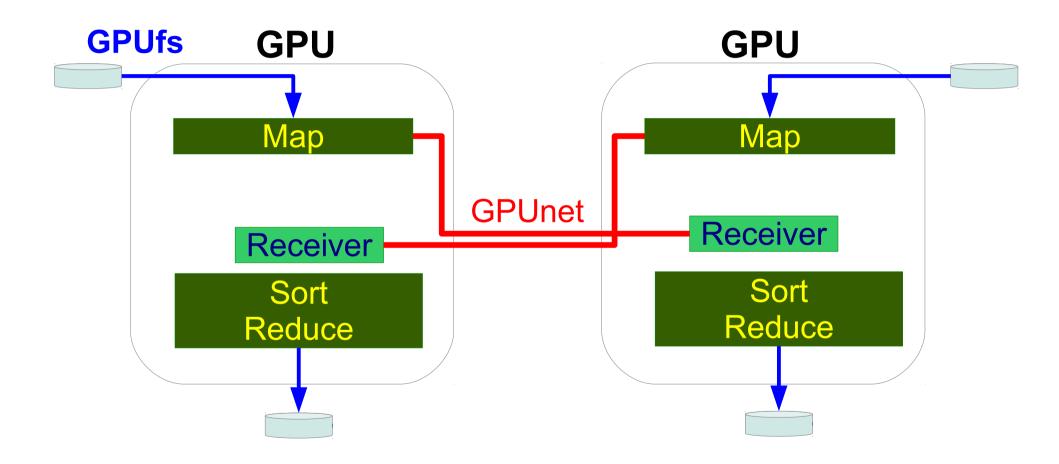


### **Evaluation**

- Analysis of GPU-native server design
  - Matrix product server
- In-GPU-memory MapReduce
- Face verification server

2x6 Intel E5-2620, NVIDIA Tesla K20Xm GPU, Mellanox Connect-IB HCA, Switch-X bridge

## In-GPU-memory MapReduce



# In-GPU-memory MapReduce: Scalability

	1 GPU (no network)	4 GPUs (GPUnet)
K-means	5.6 sec	1.6 sec ( <b>3.5</b> x)
Word-count	29.6 sec	10 sec ( <b>2.9x</b> )

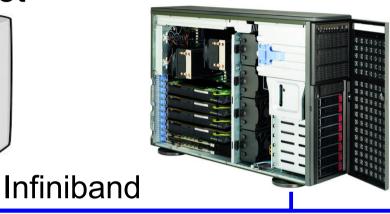
GPUnet enables scale-out for GPU – accelerated systems

### Face verification server

CPU client (unmodified) via rsocket

GPU server (GPUnet)

memcached (unmodified) via rsocket



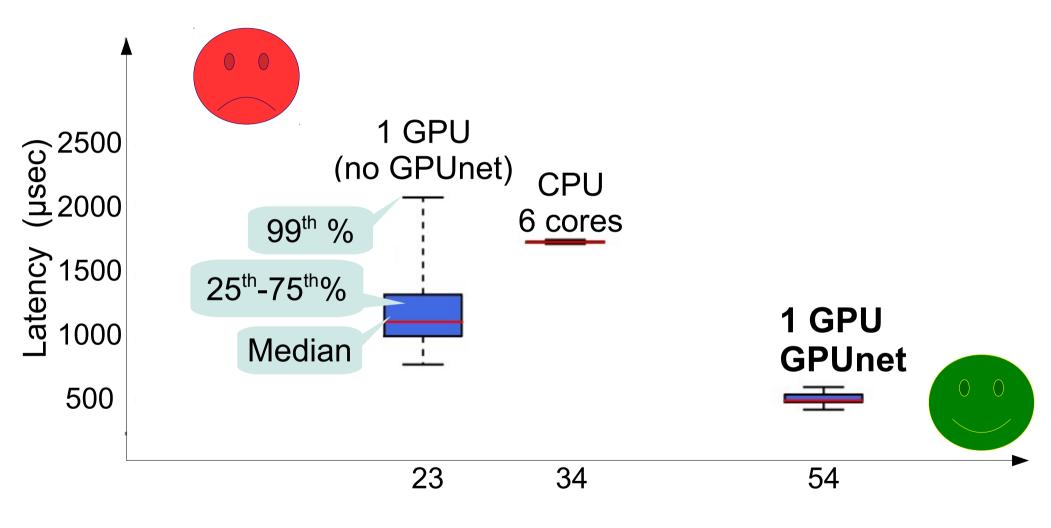




```
recv()
GPU_features()
query DB()
GPU_compare()
send()
```

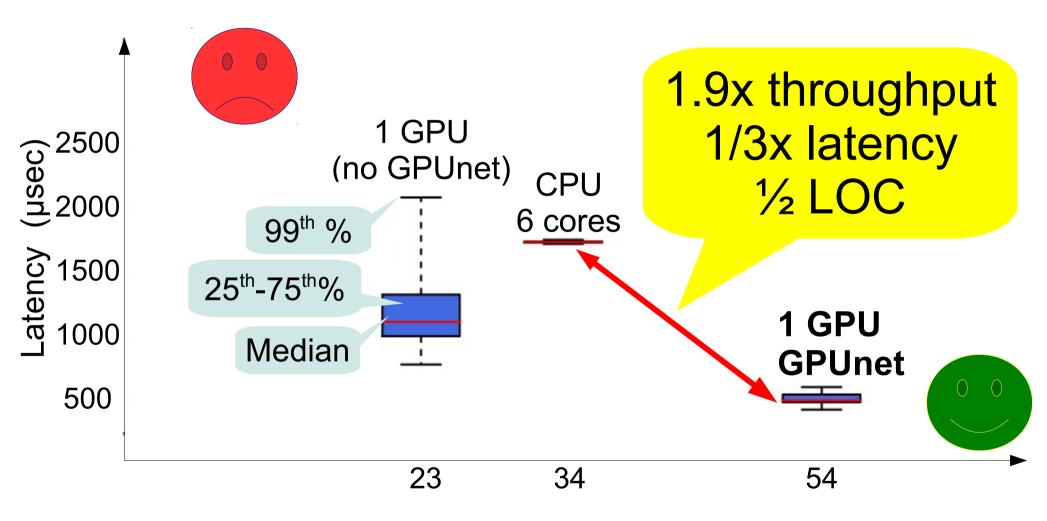


# Face verification: Different implementations



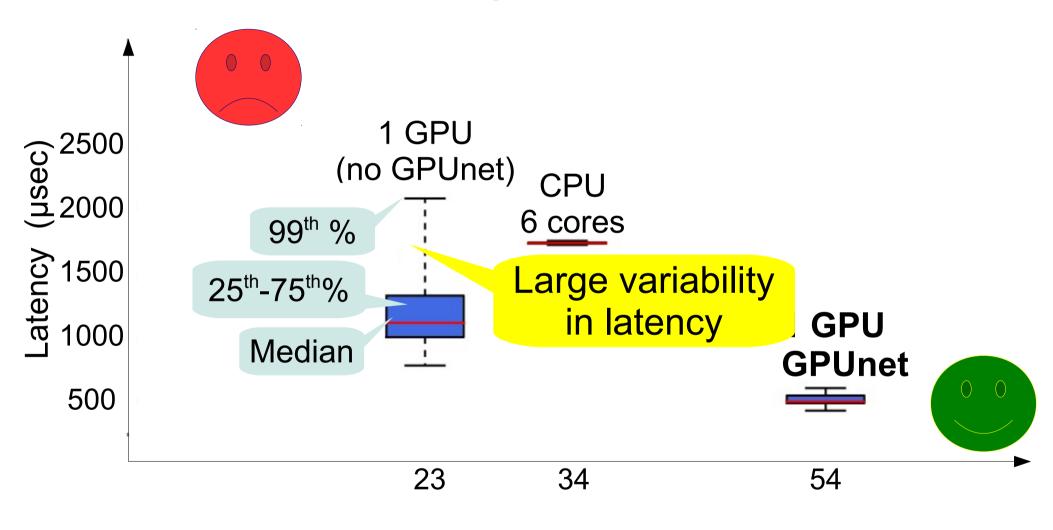
Throughput (KReq/sec)

## Face verification: Different implementations



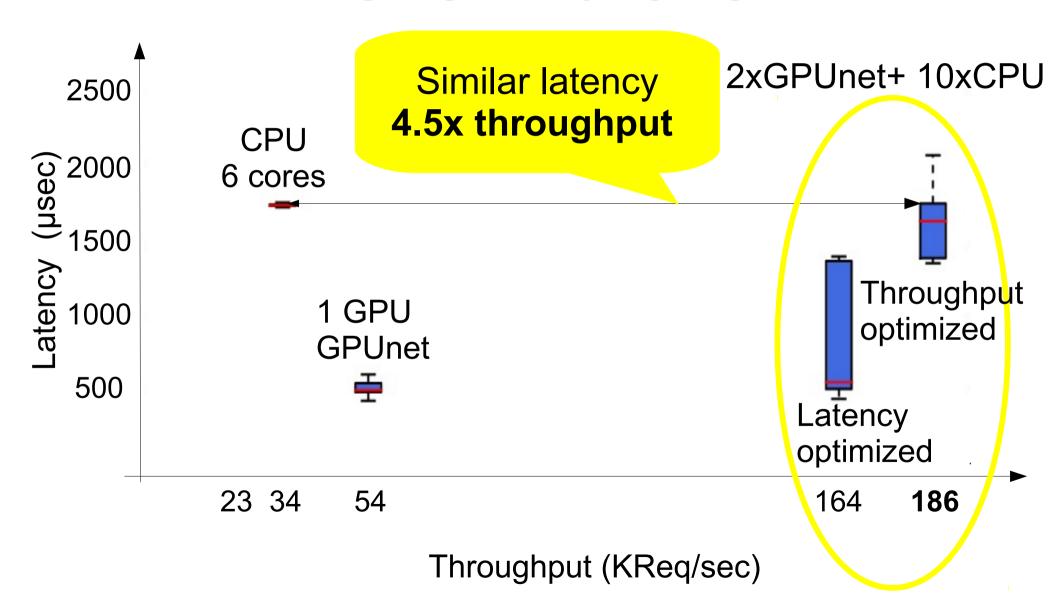
Throughput (KReq/sec)

# Face verification: Different implementations



Throughput (KReq/sec)

## Face verification on all processors 2xGPU + 10xCPU



### Set GPUs free!



GPUnet is a library providing networking abstractions for GPUs

https://github.com/ut-osa/gpunet



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