Parallelizing the Interpolation between Latent Space of Autoencoder Networks

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Overview

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- Project overview
- Convolutional Neural Network
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- Parallelism
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Related Work

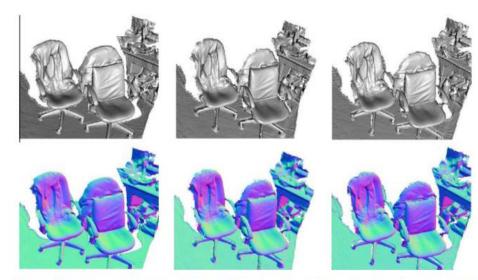


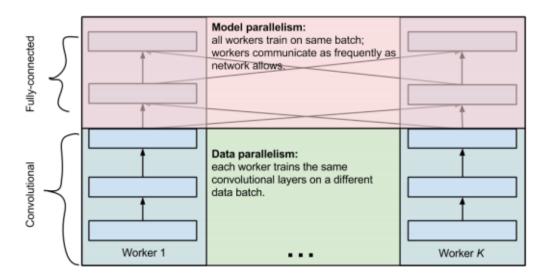
Fig. 6. Reconstruction of the "chairs" scene. Top row are the Phong shading results and bottom are normal maps. The first column are KF₅₁₂, the middle column are OF₉ and the third column are OF₁₀.

Octree-based fusion for realtime 3D reconstruction (2012) [Ming Zeng, Fukai Zhao, Jiaxiang Zheng, Xinguo Liu]



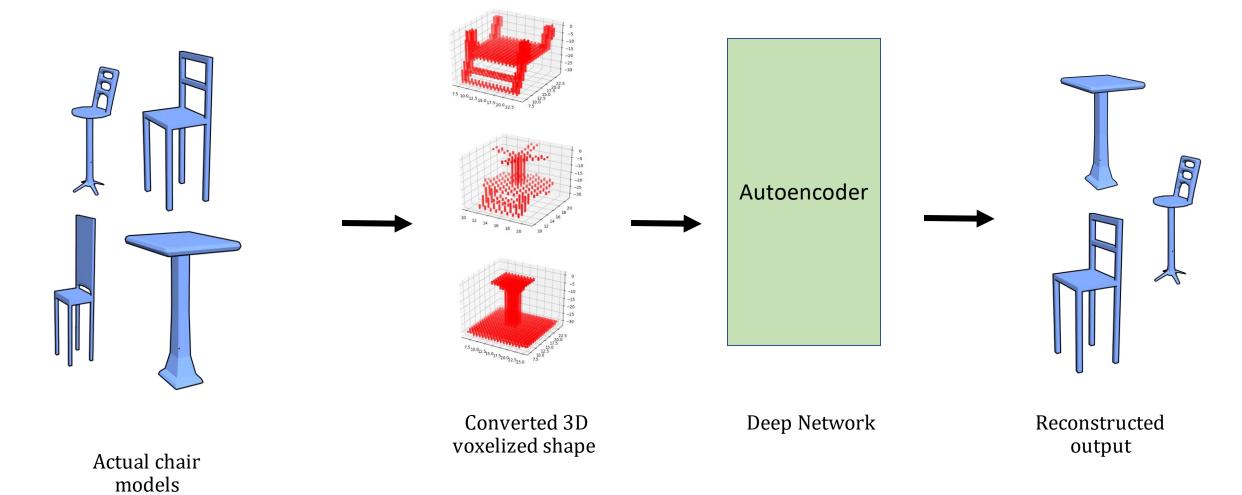
(a) Images of objects we wish to reconstruct (b) Overview of the network

3D-R2N2: A Unified Approach for Single and Multi-view 3D Object Reconstruction (2016) [Christopher B. Choy, Danfei Xu, JunYoung Gwak, Kevin Chen, Silvio Savarese]

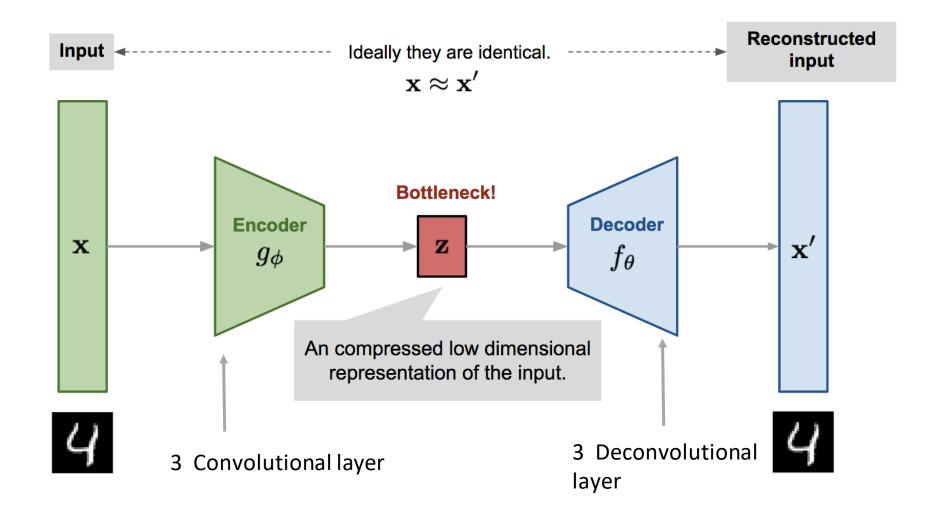


One weird trick for parallelizing convolutional neural networks (2014) [Alex Krizhevsky]

Project Overview



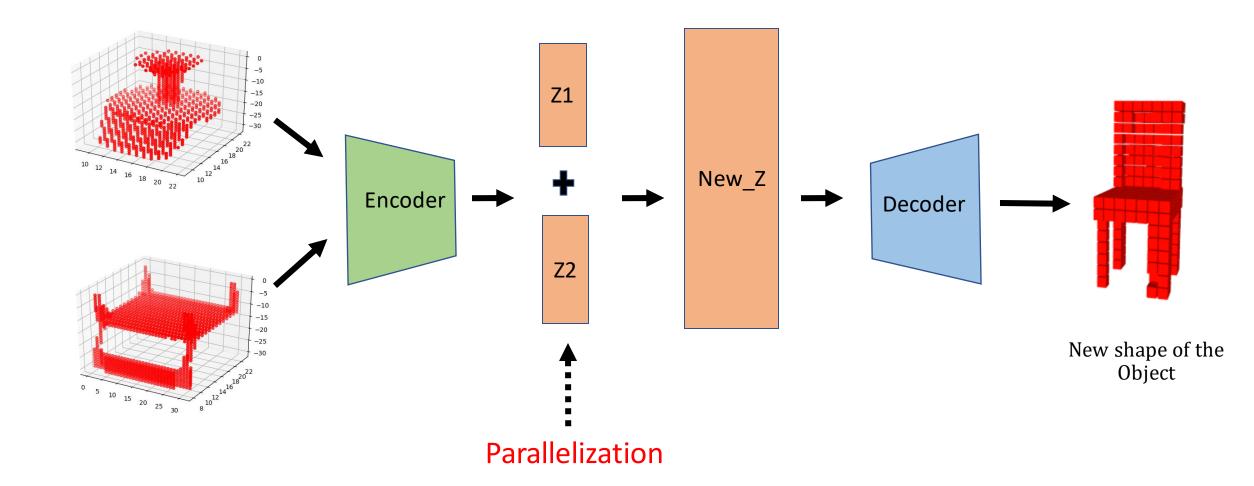
Autoencoder



Autoencoder

```
def autoencoder(inputs):
       # encoder
         32 x 32 x 32 x 1 -> 16 x 16 x 16 x 32
16 x 16 x 16 x 32 -> 8 x 8 x 16
8 x 8 x 8 x 16 -> 2 x 2 x 2 x 8
      net = lays.conv3d(inputs, 32, [5, 5, 5], stride=2, padding='SAME')
net = lays.conv3d(net, 16, [5, 5, 5], stride=2, padding='SAME')
net = lays.conv3d(net, 8, [5, 5, 5], stride=4, padding='SAME')
       latent space = net
       # decoder
       # 2 x 2 x 2 x 8 -> 8 x 8 x 8 x 16
# 8 x 8 x 8 x 16 -> 16 x 16 x 32
       # 16 x 16 x 16 x 32 -> 32 x 32 x 32 x 3
      net = lays.conv3d_transpose(net, 16, [5, 5, 5], stride=4, padding='SAME')
net = lays.conv3d_transpose(net, 32, [5, 5, 5], stride=2, padding='SAME')
net = lays.conv3d_transpose(net, 1, [5, 5, 5], stride=2,
padding='SAME', activation fn=tf.nn.tanh)
       return latent space, net
```

Interpolation in latent space



Convolutional Neural Network

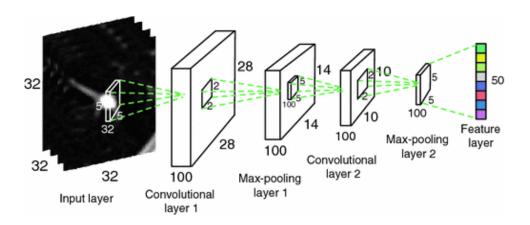
A simple Convolutional Network is a sequence of layers: Convolutional Layer, Pooling Layer, Fully-Connected Layer

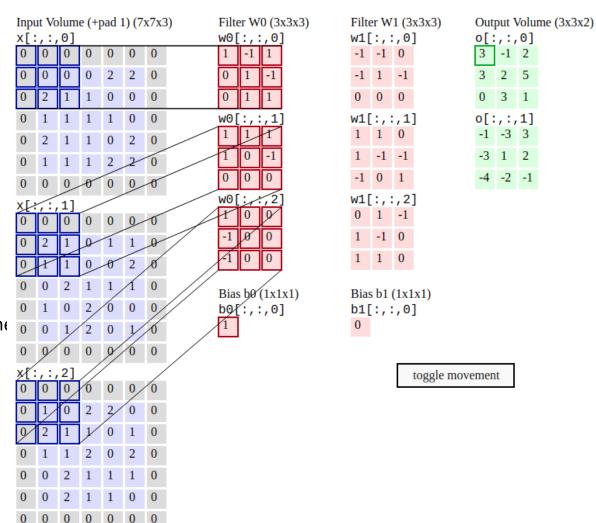
for example -

[INPUT - CONV - RELU - POOL - FC]

Convolution Layer -

- Takes input with dimension [wxhxd]
- Requires number of filters, F (e.g. [3x3x3])
- Define stride, S and other hyperparameters
- Produce output volume [wxhxd] using F, S and other hyperparame





Algorithm

```
    for i = first input volume
    for = second input volume
    Z1 = z vector for i
    Z2 = z vector for i
    for t = 0 to 1
    new_z = (1-t) x Z1 + t x Z2
    run decoder network
    t = t + 0.1
```

i= all 50 rows from the input file

j= all 50 rows from the input file except i

$$Z1 = [2 \times 2 \times 2 \times 8] = 64$$

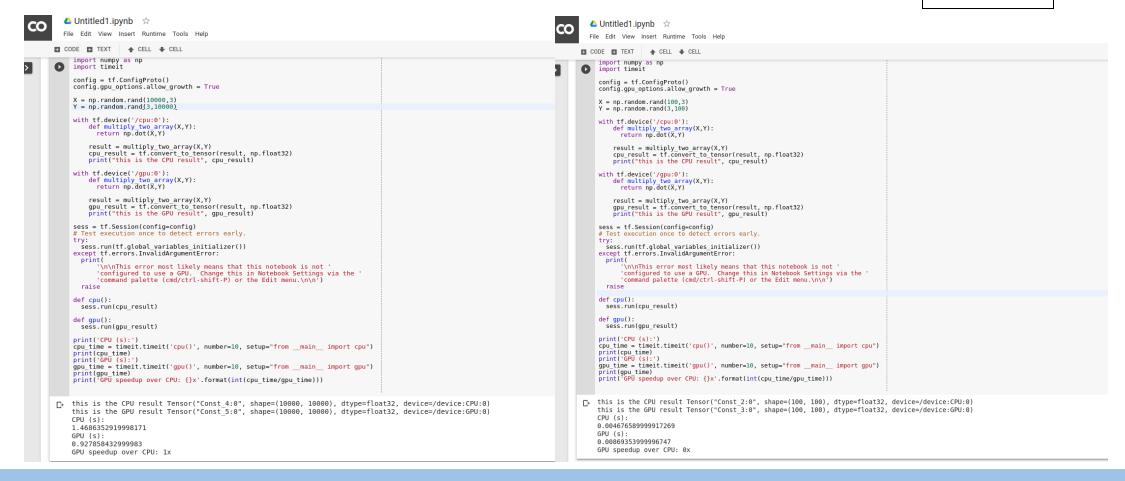
 $Z2 = [2 \times 2 \times 2 \times 8] = 64$

Why parallelism

- GPU utilization for heavy computation of data
- Increase performance (linear speedup)

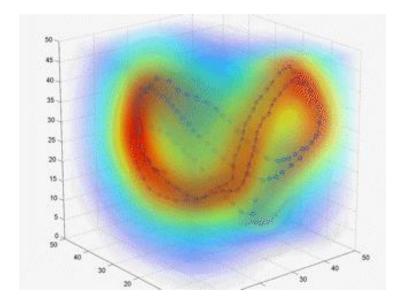
Parallel Languages -----

Cilk OpenCl MPI CUDA



Why Parallelization inside **Z** vector

- Lower dimension but maximum features
- Can be reshape easily
- Linearly grows with input data
- Involve significant amount of time while comparing with different combination of dataset

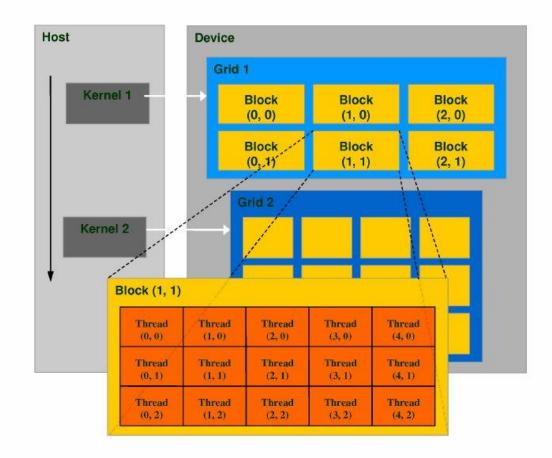


CUDA Parallel Programming Framework

- Host
- Device
- Kernel
- Threads
- Block
- Grid

CUDA supported languages

- C
- C++
- fortran
- Python



CUDA programming (Introduction to Numba)

- Numba supports CUDA GPU programming
- Compiles Python code into CUDA kernels and device functions following the CUDA execution model.
- Numba makes it appear that the kernel has direct access to NumPy arrays

```
import numpy as np
from timeit import default timer as timer
                                                _ Python decorators
from numba import vectorize
@vectorize(["float32(float32,float32)"], target='cuda')
def vectorAdd(a,b):
     return a + b
def main():
 N = 3200000
 A = np.ones(N, dtype = np.float32)
 B = np.ones(N, dtype = np.float32)
 C = np.zeros(N, dtype = np.float32)
 start = timer()
 C = vectorAdd(A, B)
 vectoradd time = timer() - start
 print("C[:5] = " + str(C[:5]))
 print("C[-5:] = " + str(C[-5:]))
 print("Vector Add took %f seconds:" , vectoradd time)
if name == " main ":
   main()
```

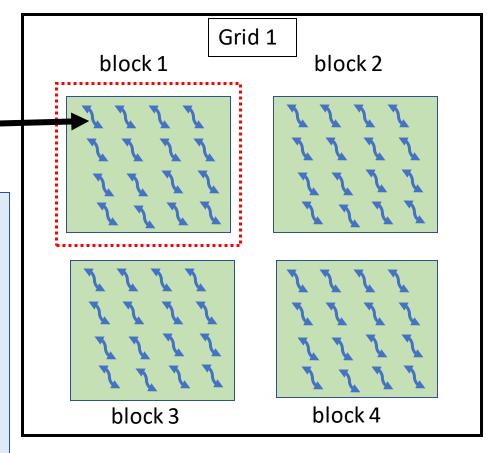
Numba Kernels

| Kernel Declaration | Kernel Invocation |
|-------------------------------------|---|
| from numba import cuda | import numpy |
| | |
| @cuda.jit | # Create the data array - usually initialized some other way |
| <pre>def my_kernel(io_array):</pre> | data = numpy.ones(256) |
| | |
| Code for kernel. | # Set the number of threads in a block |
| или | threadsperblock = 32 |
| # code here | |
| | # Calculate the number of thread blocks in the grid |
| | <pre>blockspergrid = (data.size + (threadsperblock - 1)) // threadsperblock</pre> |
| | |
| | # Now start the kernel |
| | <pre>my_kernel[blockspergrid, threadsperblock](data)</pre> |
| | |
| | # Print the result |
| | print(data) |
| | |

Project Implementation

- Dataset -- > 50 shapes of chair (each with dimension [32 x 32 x 32])
- Epoch = 5
- Z vector = [2x2x2x8] -> reshape into [8, 8]
- Thread = [4,4]
- Blocks = [2, 2]

```
TPB = 4
   l space1 = np.reshape(l space1,(8,8))
   1 space2 = np.reshape(1 space2, (8,8))
   print("lspace 1", l space1)
   print("lspace 2", l space2)
   dA = cuda.to device(l space1)
   dB = cuda.to device(l space2)
   c = np.empty like(l space2)
   dC = cuda.to device(c)
   start = timer()
   threadsperblock = (TPB, TPB)
   blockspergrid x = int(math.ceil(1 space1.shape[0] / threadsperblock[0]))
   blockspergrid y = int(math.ceil(1 space2.shape[1] / threadsperblock[1]))
   blockspergrid = (blockspergrid x, blockspergrid y)
   for t in range (0, 100):
       interpolationBetnLatentSpace[blockspergrid, threadsperblock] (dA, dB,
dC, t)
```



Implemented CUDA architecture for the project

threads

Threads Index Calculation

@numba.cuda.jit

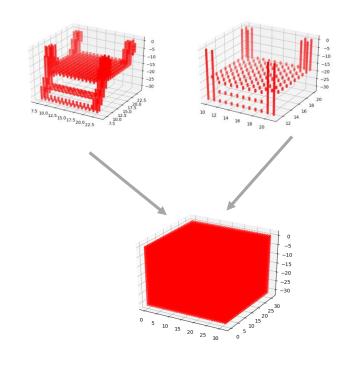
```
def interpolationBetnLatentSpace(A,B,C,t):
    tx = cuda.threadIdx.x
    ty = cuda.threadIdx.y
    bx = cuda.blockIdx.x
    by = cuda.blockIdx.y
    bw = cuda.blockDim.x
    bh = cuda.blockDim.y
    row = tx + bx * bw
    col = ty + by * bh
    if row < C.shape[0] and col < C.shape[1]:</pre>
        for k in range(A.shape[1]):
            tmp = (1-t) * A[row, k] + t *
B[row, k]
            C[row, k] = tmp
```

Global memory

@numba.cuda.jit

```
def interpolationBetnLatentSpace(A,B,C,t):
    sA = cuda.shared.array(shape=(8, 8),
dtype=float32)
    sB = cuda.shared.array(shape=(8, 8),
dtype=float32)
    x, y = cuda.grid(2)
    tx = cuda.threadIdx.x
    ty = cuda.threadIdx.y
    if x < C.shape[0] and y < C.shape[1]:
        for i in range(int(A.shape[1] / TPB)):
            # Preload data into shared memory
            sA[tx, ty] = A[x, ty + i * TPB]
            sB[tx, ty] = B[tx + i * TPB, y]
            cuda.syncthreads()
            for k in range(8):
                tmp = (1-t) * A[tx, k] + t * B[tx, k]
                cuda.syncthreads()
                                 Shared memory
                C[tx, k] = tmp
```

Results



Output volume after interpolation

CPU vs GPU analysis

For 5 epochs:

Total time CPU = 37.15s

Vector Add time CPU = 0.0017s

Time taken by CNN layers = 11s

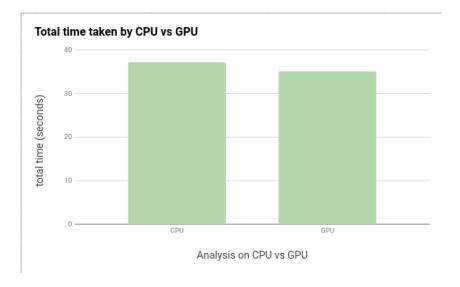
Total time GPU = 35.04s

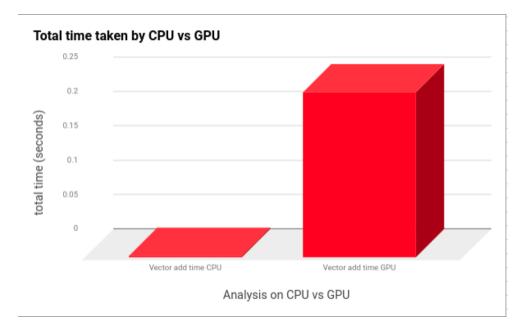
Vector Add time GPU (global memory) = 0.24s

Vector Add time GPU (shared memory) = 0.17s

Time taken by CNN layers = 11s

GPU speedup over CPU = 37.15 / 35.04 seconds = 1.06 seconds





Limitation

- 1. For one pair comparison of volumes, CPU performs better than GPU
- 2. Very low dimension of Z could effect GPU performance
- 3. Running threads in shared memory of GPU is taking 1.5x more than global memory

Challenges

- 1. As a dynamic type system, explicit memory allocation in Python is not possible
- 2. Less example and documentation of Python-CUDA compared to CUDA-C or CUDA-C++
- 3. Inside Kernel function, debugging is not available



Questions

- 1. What is Linear Interpolation?
- 2. How does Numba work?
- 3. Considering the results, does increasing the dimension of Z vectors or volumes of input shape would increase the GPU performance?