



30/11/23

Real-Time Visual and Machine Learning Systems



Types

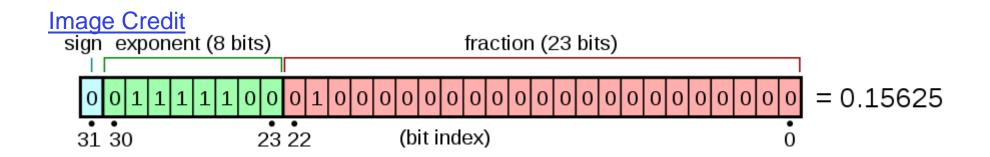
- GPU Hand-in
 - Any Questions?
 - Common problems?
 - Bind groups and unused elements
- Analysis Hand-in
 - Any Questions?
- Projects
 - Any Questions?



Types

Integers and floats Bitwise operators

Any questions?





- It is almost impossible for two floats to be exactly equal, which also makes it hard to compress
- Every calculation results in an error
- X / N is only the same as X * (1/N) when N is in the set of numbers 2^M

Image credits



- Floating point numbers small to large and sorting
- Pairwise additions

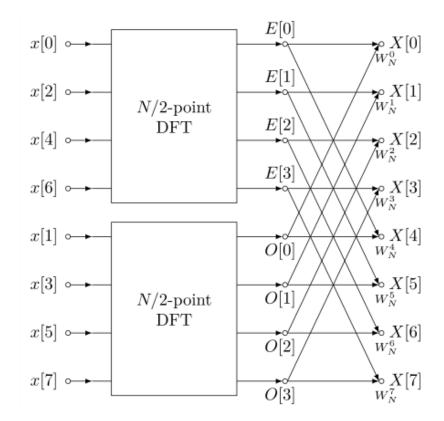


Image credits

Date

The upper bound on the relative error for the Cooley–Tukey algorithm is $O(\varepsilon \log n)$, compared to $O(\varepsilon n^{3/2})$ for the naïve DFT formula, where ε is the machine floating-point relative precision. In fact, the root mean square (rms) errors are much better than these upper bounds, being only $O(\varepsilon \sqrt{\log n})$ for Cooley–Tukey and $O(\varepsilon \sqrt{n})$ for the naïve DFT (Schatzman, 1996). These results, however, are very



Fused multiplication-addition results in 1 error instead of 2

```
fn fma(e1: T, e2: T, e3: T) -> S is AbstractFloat, f32, or f16T is S
T or vecN<S>
```



$$d = a\frac{1}{z} + b$$

- Precision in Depth Buffers (reversing the logarithm)
 - Depth values stored as 1/z
 - Near plane (a) = 0.0, Far plane (b) = 1.0

Image Credit

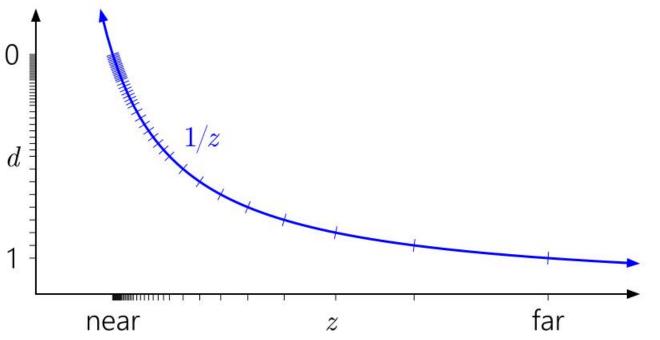
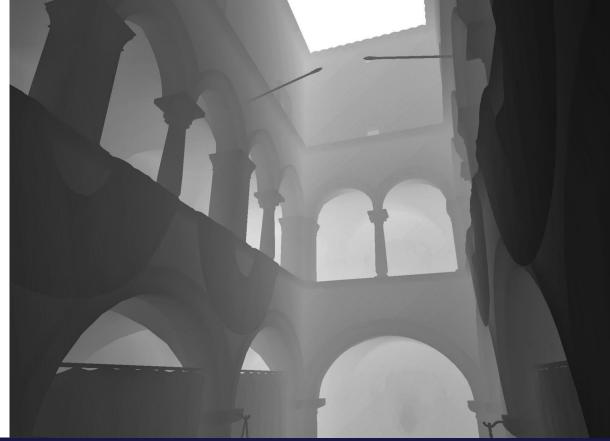


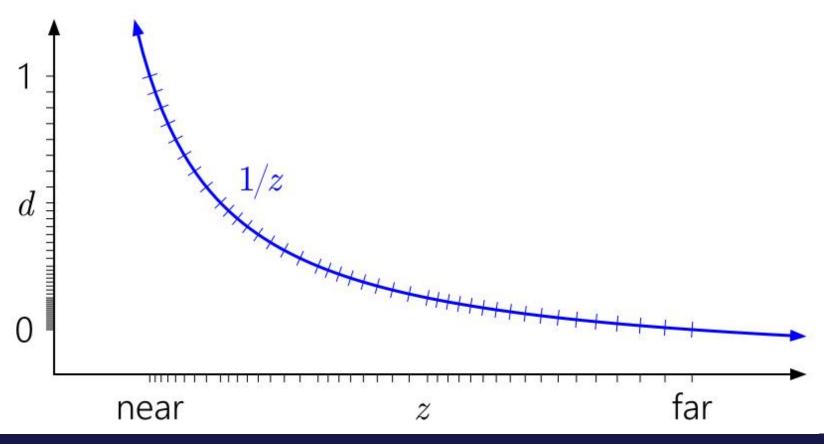
Image credit





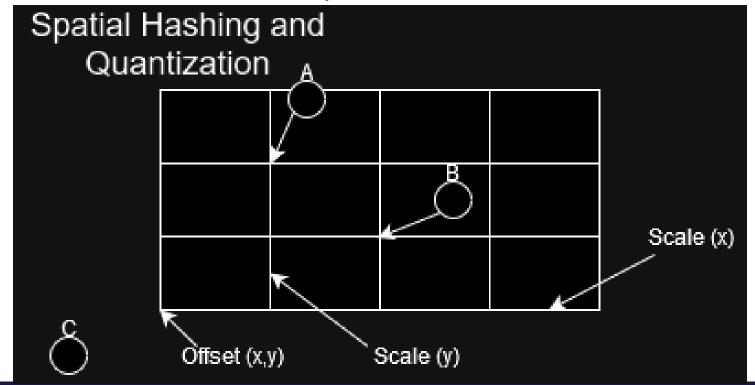
- Precision in Depth Buffers (reversing the logarithm)
 - Reverse Depth Buffer: Near plane = 1.0, Far plane = 0.0

Image Credit





- We can quantize floats to integers
- Integers compress much better than floats and can be sorted in an exciting number of ways
- Integers allow us to do bit tricks! Yay!





 Spatial hashing is basically quantization, some shifts and some additions

```
// This would be given for this specific grid
// If we aren't TOO bothered by the loss of precision we can
// multiply by the inverse of the scaling term.
let x_scale: f32 = 5.0f;
let x_offset: f32 = 0.32f:
let y_scale: f32 = 4.0f;
let y_offset: f32 = 3.50f;
let z_scale: f32 = 2.0f;
let z_{offset}: f32 = 0.42f;
let x: f32 = 7.2f;
let y: f32 = 5.5f;
let z: f32 = 2.2f:
let x_quantized: u64 = ((x - x_offset) / x_scale) as u64;
let y_quantized: u64 = ((y - y_offset ) / y_scale) as u64;
let z_quantized: u64 = ((z - z_offset) / z_scale) as u64;
let x_hash: u64 = (x_quantized & 0xFFFFF) << 40;</pre>
let y_hash: u64 = (y_quantized & 0xFFFFF) << 20;</pre>
let z_hash: u64 = z_quantized & 0xFFFFF;
let hash: u64 = x_hash \mid y_hash \mid z_hash;
```

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Types - Sorting

- .sort() and floats/ordering in Rust
- Quantizing floats to integers
- Sorting usually also has performance implications and is usually a good idea in preprocessing

Date Technical University of Denmark Title

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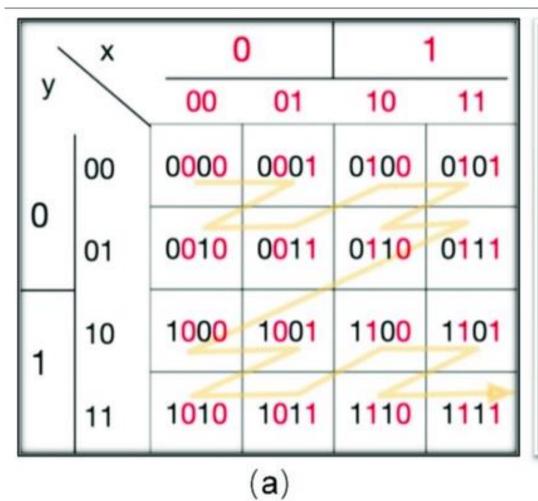


Types – Morton Codes

 An encoding, not a sorting method, but we can change the way numbers are sorted by encoding them differently and shuffling bits around



Types – Sorting – Morton



```
bits | index
node
0,0,0
          000
                    0
0,0,1
          001
0,1,0
          010
0,1,1
          011
1,0,0
          100
                    4
                    5
          101
1,0,1
          110
                    6
1,1,0
1,1,1
          111
```

(b)



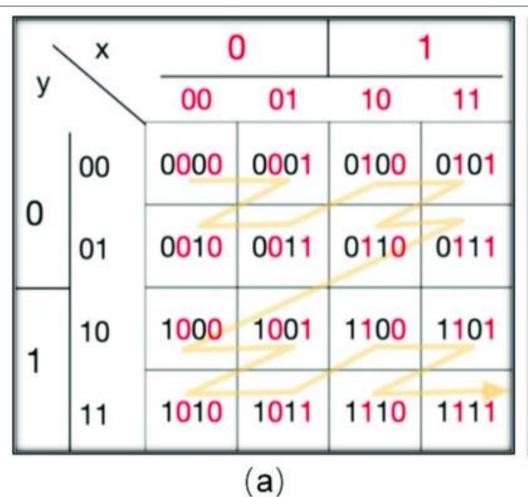
Types – Sorting – Morton

```
let x: u32 = 8;
let y: u32 = 1;
let z: u32 = 2;
let component_count: u32 = 3;
let bits_of_precision: u32 = 10;
let mut encoded: u32 = 0;
let mut mask: u32 = 1;
for index in 0..bits_of_precision {
    encoded |= ((x & (mask << index)) << component_count * index);
    encoded |= ((y & (mask << index)) << (component_count * index + 1));
    encoded |= ((z & (mask << index)) << (component_count * index + 2));
```



Date

Types – Sorting – Morton



```
bits | index
node
0,0,0
          000
                    0
0,0,1
          001
0,1,0
          010
0,1,1
          011
1,0,0
          100
                    4
                    5
          101
1,0,1
          110
                    6
1,1,0
1,1,1
          111
```

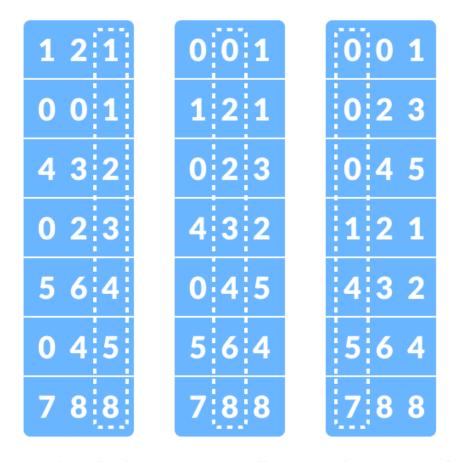
(b)

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Types – Sorting – Radix

- Radix sort!
- There are many ways to sort, but Radix has known GPU implementations to do so in a single shader dispatch



sorting the integers according to units, tens and hundreds place digits

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Types - Compression

- Optimize your data for a more advanced algorithm to do the heavy lifting
- Avoid floats and strings if you can, especially floats represented as strings
- Point cloud compression case:
 - Hierarchical data structure and spatial hashing
 - Quantification (16-bits)
 - Bit nulling
 - Sorting
 - Delta encoding
 - Gzip
 - Transfer
 - Un-gzip
 - Transfer to gpu
 - Dequantify in shader



Date

Types – Energy Usage

Table VI ENERGY CONSUMPTION OF ARITHMETIC OPERATIONS ON THE INTEL XEON X5670 TEST SYSTEM

Workload	operations per 16 Byte	biop	E _{calc}
add pi	2 (64 Bit)	8 Byte/op	428 pJ/op
mul_pi	2 (64 Bit)	8 Byte/op	476 pJ/op
add_pd	2 (64 Bit)	8 Byte/op	319 pJ/op
mul_pd	2 (64 Bit)	8 Byte/op	387 pJ/op
mul+add _pd	4 (64 Bit)	4 Byte/op	464 pJ/op
add ps	4 (32 Bit)	4 Byte/op	111 рЈ/ор
mul_ps	4 (32 Bit)	4 Byte/op	164 pJ/op



Types – Energy Usage

Table V COMPARISON OF THE ENERGY CONSUMPTION OF DATA TRANSFERS (movAP s), AMD OPTERON 2435 AND INTEL XEON X5670

Location	Et,. AMD	E _{trans} Intel	Factor
Li	105 pJ/Byte	64 pJ/Byte	1.6
L2	357 pJ/Byte	121 pJ/Byte	2.9
L3	654 pJ/Byte	254 pJ/Byte	2.6
RAM	3590 pJ/Byte	1250 pJ/Byte	2.8



Profilers

- Timing
- System monitoring (Task manager, htop, nvidia-smi)
- General profilers (Visual Studio, perf, tracy)
- System specific (AMD, Intel VTune)
- GPU specific(nsight-compute, Renderdoc, AMD)
- App specific (PyTorch profiler)
- Disk is hard to profile, system monitors are likely the way to go

Too many profilers to go into too much detail, do the exercise!