

```
ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 1.2f)
    {
        nt = nt / nc; ddn = ddn * ddn;
        cos2t = 1.0f - nnt * nnt;
        D, N );
    }
    at a = nt - nc, b = nt + nc;
    at Tr = 1 - (R0 + (1 - R0) * ddn);
    (Tr) R = (D * nnt - N * (ddn > 0 ? 1 : -1));
    E * diffuse;
    = true;
    -
    refl + refr)) && (depth < MAXDEPTH)
    {
        D, N );
        refl * E * diffuse;
        = true;
    }
    MAXDEPTH)
    survive = SurvivalProbability( diffuse );
    estimation - doing it properly, closely following
    df;
    radiance = SampleLight( &rand, I, &L, &light,
    e.x + radiance.y + radiance.z) > 0) && (ace)
    w = true;
    at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
    at3 factor = diffuse * INVPI;
    at weight = Mis2( directPdf, brdfPdf );
    at cosThetaOut = dot( N, L );
    E * ((weight * cosThetaOut) / directPdf) * (radiance
    random walk - done properly, closely following
    vive)
    ;
    at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf,
    survive;
    pdf;
    n = E * brdf * (dot( N, R ) / pdf);
    sion = true;
}
```

/INFOMOV/

Optimization & Vectorization

J. Bikker - Sep-Nov 2019 - Lecture 7: “Data-Oriented Design”

Welcome!



Today's Agenda:

- OOP Performance Pitfalls
- DOD Concepts
- DOD or OO?



Fact Checking

“Floating point code is (typically) undeterministic”

```

ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 0)
    {
        nt = nt / nc; ddn = sqrt(1 - nt * nt);
        pos2t = 1.0f + nnt * nt;
        D, N );
    }
    {
        at a = nt - nc, b = nt + nc;
        at Tr = 1 - (R0 + (1 - R0) * pos2t);
        Tr) R = (D * nnt - N * (ddn * pos2t));
    }
    E * diffuse;
    = true;
    -
    efl + refr)) && (depth < MAXDEPTH)
    {
        D, N );
        -refl * E * diffuse;
        = true;
    }
    MAXDEPTH)
    survive = SurvivalProbability( diffuse, i );
    estimation - doing it properly, closely following
    if;
    radiance = SampleLight( &rand, I, &L, &light, &pos, &N );
    e.x + radiance.y + radiance.z) > 0) && (depth < MAXDEPTH)
    {
        w = true;
        at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
        at3 factor = diffuse * INVPI;
        at weight = Mis2( directPdf, brdfPdf );
        at cosThetaOut = dot( N, L );
        E * ((weight * cosThetaOut) / directPdf) * (radiance.x + radiance.y + radiance.z);
    }
    random walk - done properly, closely following SurvivalProbability
    vive)
    {
        at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf );
        survive;
        pdf;
        n = E * brdf * (dot( N, R ) / pdf);
        sion = true;
    }
}

```

```

float v0 = 1;
float v1 = 1;
float v2 = 1;
float v3 = 1;
float v4 = 1;
float v5 = 1;
float v6 = 1;
float v7 = 1;
for (int i = 0; i < 2000000; i++)
{
    v0 *= 1.00001f;
    v1 *= 1.00001f;
    v2 *= 1.00001f;
    v3 *= 1.00001f;
    v4 *= 1.00001f;
    v5 *= 1.00001f;
    v6 *= 1.00001f;
    v7 *= 1.00001f;
}

```

```

fld1
fld st(0)
fld st(1)
fld st(2)
fld st(3)
fld st(4)
fld st(5)
fld st(6)

fmul st(7),st ; fxch st(7) ; fstp [v0]
fxch st(5) ; fmul st,st(6)
fxch st(4) ; fmul st,st(6)
fxch st(3) ; fmul st,st(6)
fxch st(2) ; fmul st,st(6)
fxch st(1) ; fmul st,st(6)
fxch st(5) ; fmul st,st(6)
fld [v7] fmul st,st(7) fstp [v7]

```



Fact Checking

“Doubles are slower than floats (4x)”

This statement is **mostly true**. The real story, CPU (win32, x64):

- A **float** takes 32-bit in memory, but gets promoted to 80 bits in an FPU register.
- A **double** takes 64-bit in memory, but gets promoted to 80 bits in an FPU register.
- A **long double** takes 64-bit in memory, but gets promoted to 80 bits in an FPU register.

Calculation time on 80-bit FPU registers does not depend on the source of the data.

HOWEVER: the fp registers are rarely used anymore...

The real story, GPU (Nvidia, AMD): <https://www.geeks3d.com/20140305/amd-radeon-and-nvidia-geforce-fp32-fp64-gflops-table-computing>

- Titan V: FP64 = 1/2 * FP32 (6900 vs 13800 GFLOPS)
- Titan X Pascal: FP64 = 1/32 * FP32 (350 vs 11300 GFLOPS) (same for all 10xx)
- Radeon RX Vega 64: FP64 = 1/16 * FP32 (790 vs 12700 GFLOPS)
- Radeon HD 7990: FP64 = 1/4 * FP32 (1946 vs 7782)

FP16 (GPU only): <https://www.anandtech.com/show/10325/the-nvidia-geforce-gtx-1080-and-1070-founders-edition-review/5>

- GTX 1080Ti: FP16 = 1/64 * FP32 (ouch)
- Radeon RX Vega 64: FP16 = 2 * FP32 (!)



Today's Agenda:

- OOP Performance Pitfalls
- DOD Concepts
- DOD or OO?

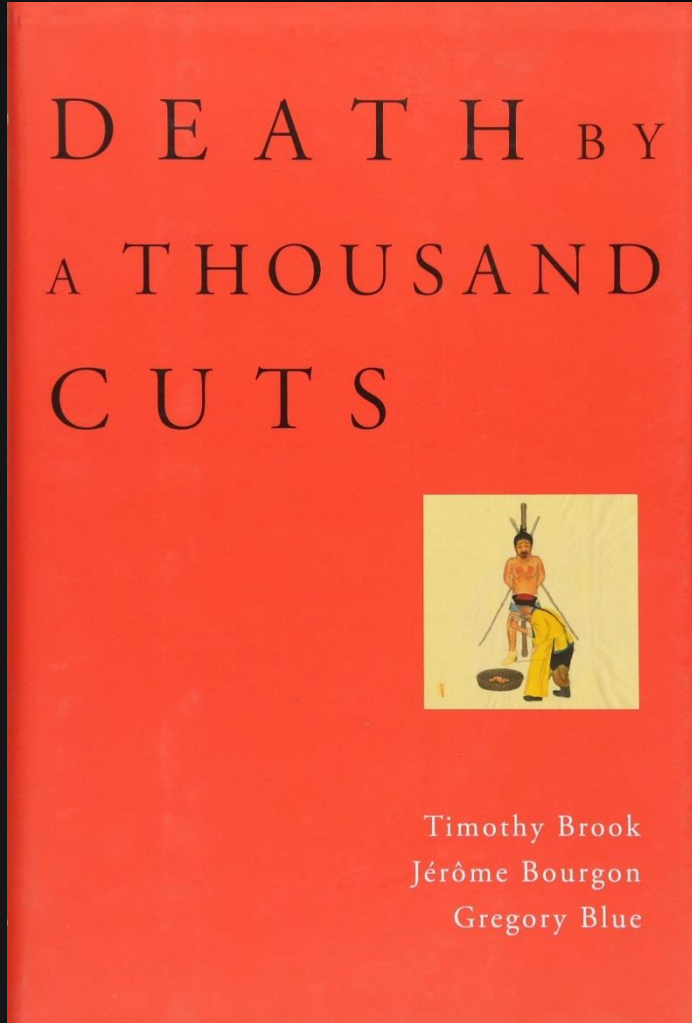


OOP

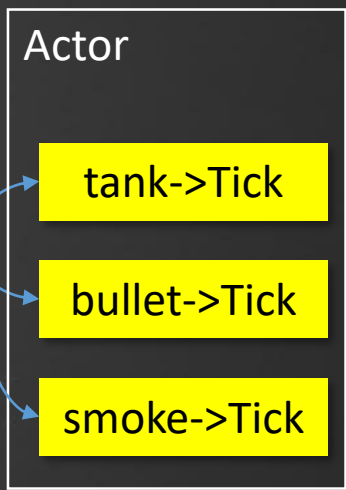
“Death by a Thousand Cuts”

Object Oriented Programming:

- Objects
 - Data
 - Methods
 - Instances



```
...ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 0)
    {
        nt = nt / nc; ddn = ddn * nc;
        pos2t = 1.0f + nnt * nnt;
        D, N );
    }
    at a = nt - nc; b = nt + nc;
    at Tr = 1 - (R0 + (1 - R0) * ddn);
    Tr) R = (D * nnt - N * (ddn * nnt));
    E * diffuse;
    = true;
    defl + refr)) && (depth < MAXDEPTH)
    {
        D, N );
        refl * E * diffuse;
        = true;
    }
    MAXDEPTH)
    survive = SurvivalProbability( diffuse );
    estimation - doing it properly, closely following
    df;
    radiance = SampleLight( &L, N );
    e.x + radiance.y + radiance.z;
    w = true;
    at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
    at3 factor = diffuse * INVPI;
    at weight = Mis2( directPdf, brdfPdf );
    at cosThetaOut = dot( N, L );
    E * ((weight * cosThetaOut) / directPdf) * (radiance);
    random walk - done properly, closely following S&S
    vive)
    {
        at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf );
        survive;
        pdf;
        n = E * brdf * (dot( N, R ) / pdf);
        sion = true;
    }
}
```

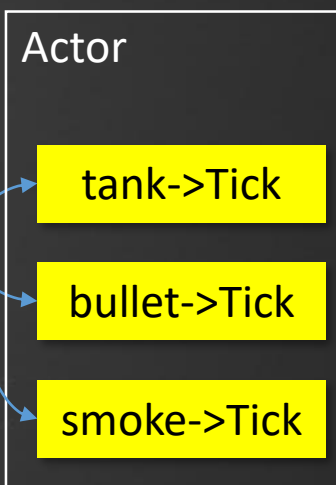


OOP

“Death by a Thousand Cuts”

Object Oriented Programming:

- Objects
 - Data
 - Methods
 - Instances



Cost of a virtual function call:

1. Virtual Function Table
2. No inlining

...

Calling such a function:

cache miss

cache miss

branch

1. Read pointer to VFT of base class
2. Add function offset
3. Read function address from VFT
4. Load address in PC (jump)

But, that isn't realistic, right?

It is, if we use OO for what it was designed for: operating on heterogeneous objects.



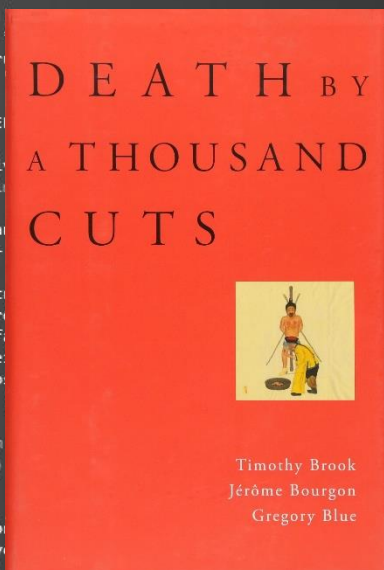
OOP

“Death by a Thousand Cuts”

Characteristics of OO:

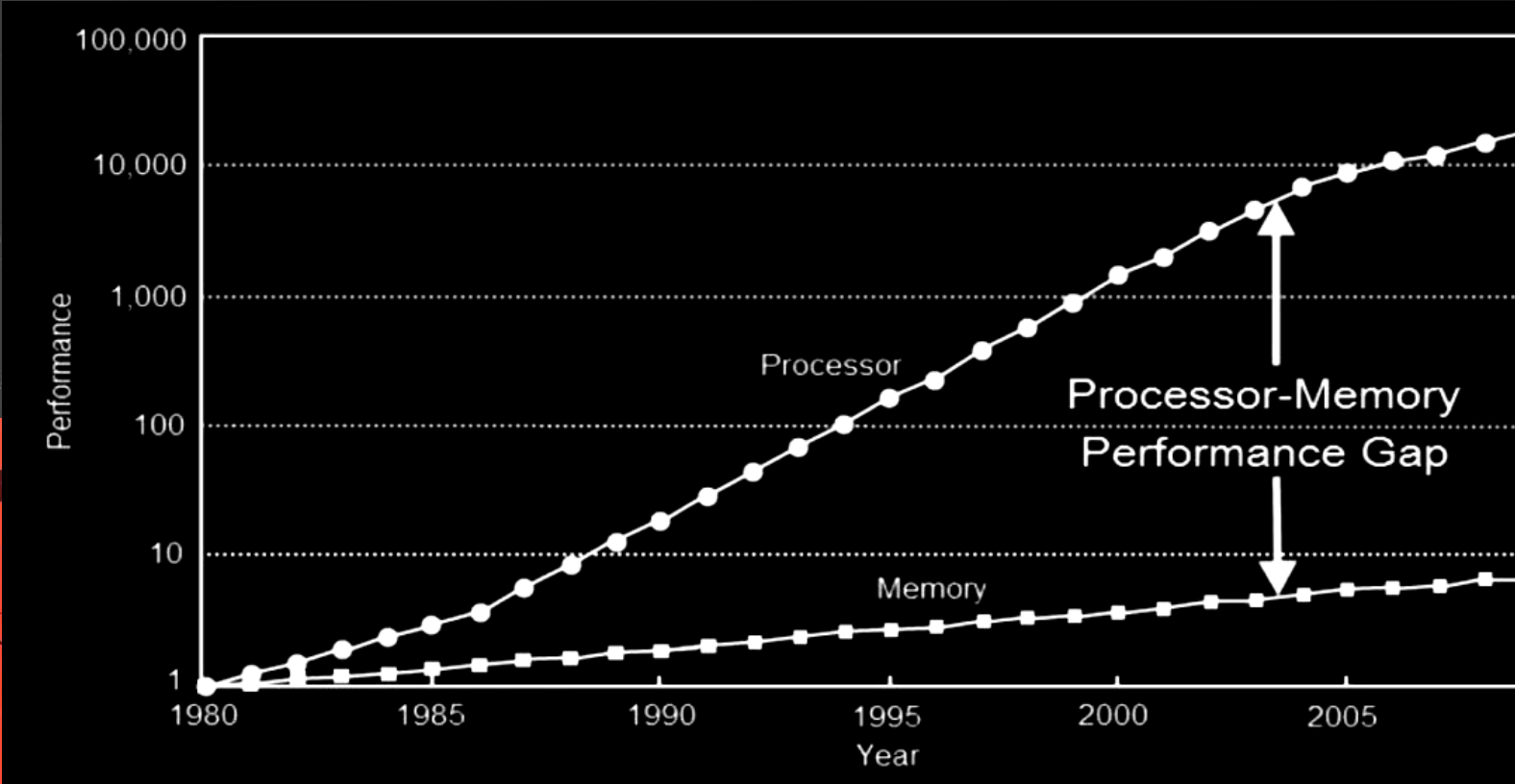
- Virtual calls
- Scattered individual objects

```
ics
& (depth < MAXDEPTH)
c = inside ? 1 : 0;
nt = nt / nc; ddn = ddn / nc;
os2t = 1.0f / nnt * ddn;
D, N );
0);
at a = nt - nc, b = nt + nc;
at Tr = 1 - (R0 + (1 - R0) *
Tr) R = (D * nnt - N * (ddn
E * diffuse;
= true;
efl + refr)) && (depth < MAXDEPTH)
```



OOP

“Death by a Thousand Cuts”



Reading memory: 40 cycles @ 300Mhz



Reading memory: 600 cycles @ 3.2Ghz

Timothy Brook
Jérôme Bourgon
Gregory Blue

The problem is growing with time.



OOP

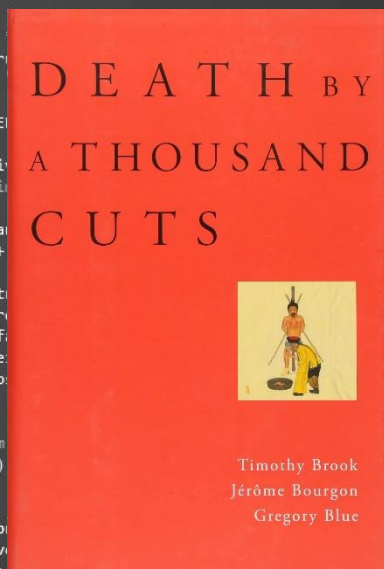
“Death by a Thousand Cuts”

Dealing with “bandwidth starvation”:

Caching

Continuous memory access (full cache lines)

Large array continuous memory access



(caches ‘read ahead’)



OOP

“Death by a Thousand Cuts”

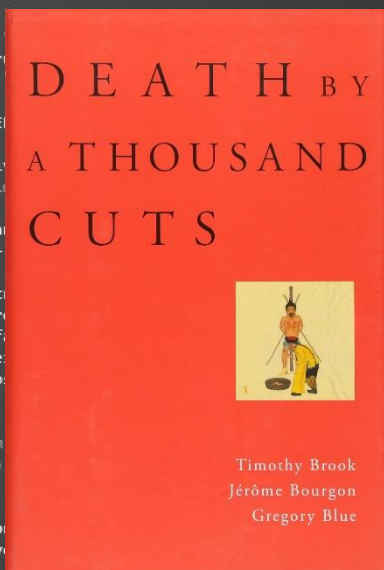
Code performance is typically bound by memory access.

“The ideal data is in a format that we can use with the least amount of effort.”

→ Effort = CPU-effort.

“Most programs are made faster if we improve their memory access patterns.”
(this will be more true every year)

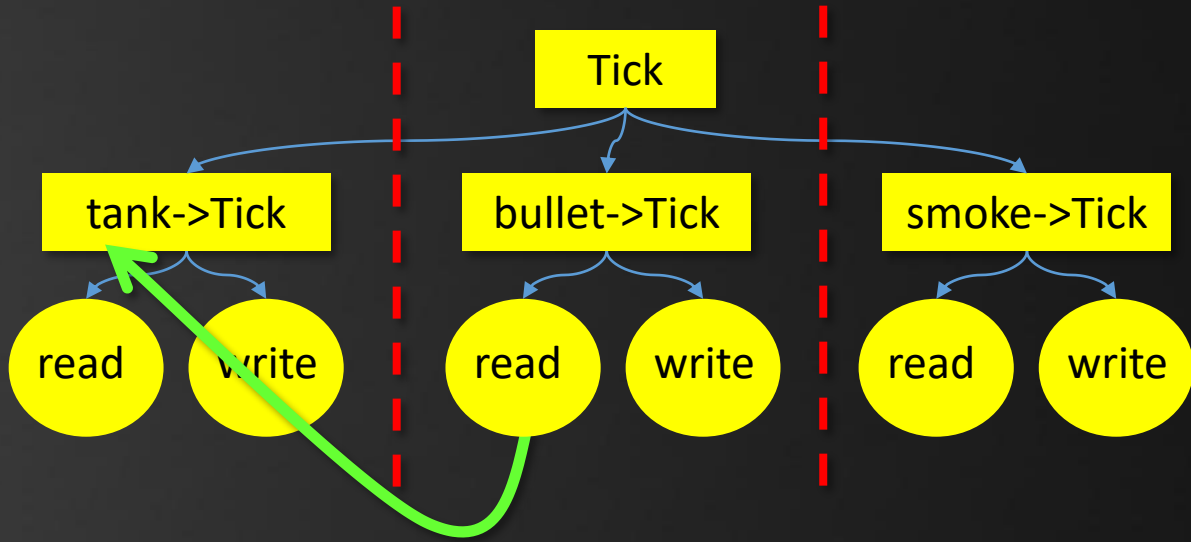
“You cannot be fast without knowing how data is touched.”



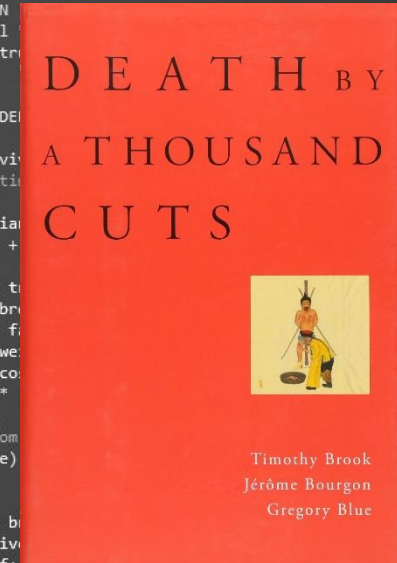
OOP

“Death by a Thousand Cuts”

Parallel processing typically requires synchronization.



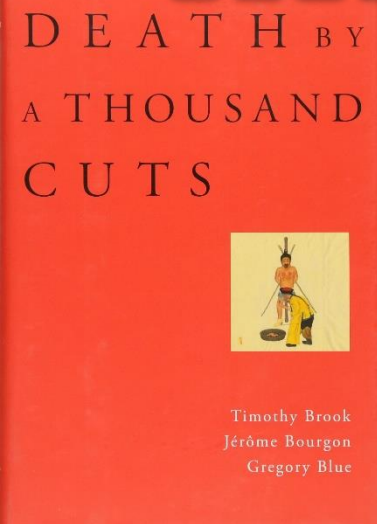
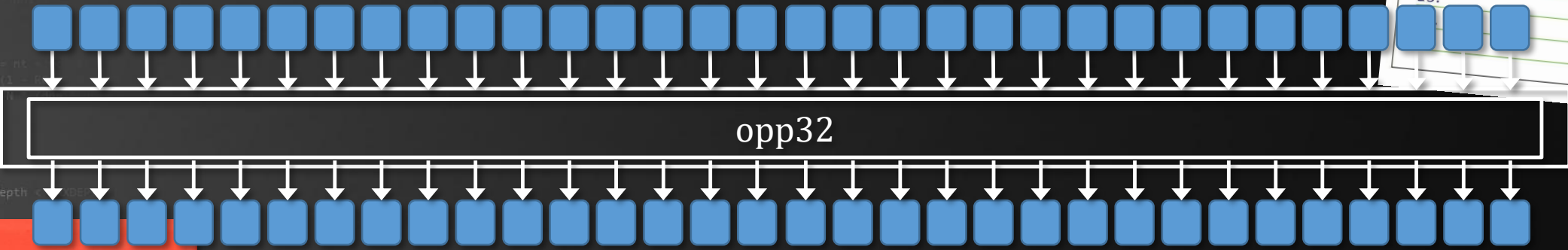
“You cannot *multi-thread* without knowing how data is touched.”



OOP

“Death by a Thousand Cuts”

Parallel processing requires coherent program flow.



*“You cannot **multi-thread** without knowing how data is touched.”*



OOP

“Death by a Thousand Cuts”

```
class Bot : public Enemy
{
    ...
    vec3 m_position;
    ...
    float m_mod;
    ...
    float m_aimDirection;
    ...
    virtual void updateAim( vec3 target )
    {
        m_aimDirection = dot3( m_position, target ) * m_mod;
    }
}
```

branch
cache miss
cache miss

cached but not used

cached but not used

cache miss

cache miss

cache miss



OOP

“Death by a Thousand Cuts”

```
void updateAims(
    float* aimDir,
    const AimingData* aim,
    vec3 target,
    uint count
```

only reads data that
is actually needed to
cache

```
for (uint i = 0; i < count; ++i)
{
```

reads from
linear array

```
    aimDir[i] = dot3(aim->positions[i], target) * aim->mod[i];
```

writes to
linear array

actual functionality is unchanged



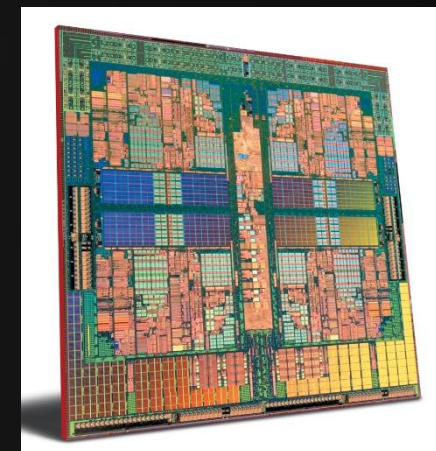
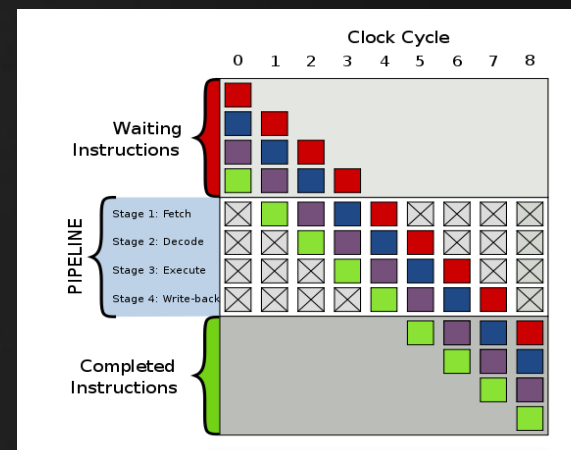
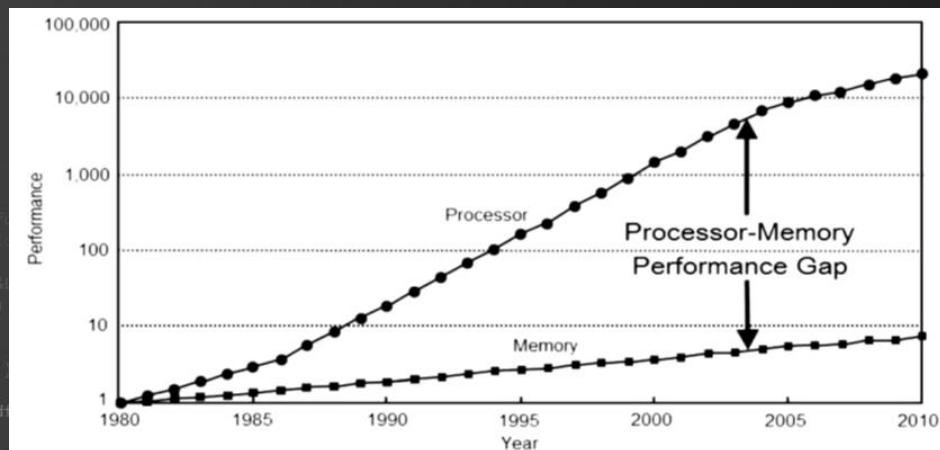
OOP

Algorithm Performance Factors

Estimating algorithm cost:

1. Algorithmic Complexity : $O(N)$, $O(N^2)$, $O(N \log N)$, ...
2. Cyclomatic Complexity* (or: Conditional Complexity)
3. Amdahl's Law / Work-Span Model
4. Cache Effectiveness

t



*. McCabe, A Complexity Measure, 1976.



Today's Agenda:

- OOP Performance Pitfalls
- DOD Concepts
- DOD or OO?



DOD

Data Oriented Design*

Origin: low-level game development.

Core idea: *focus software design on CPU- and cache-aware data layout.*

Take into account:

- Cache line size
- Data alignment
- Data size
- Access patterns
- Data transformations

Strive for a simple, linear access pattern as much as possible.

*: Nikos Drakos, “Data Oriented Design”, 2008. <http://www.dataorienteddesign.com/dodmain>



DOD



```
ics
& (depth < MAXDEPTH)
t = inside ? 1 : 0;
nt = nt / nc; ddn = ddn * ddn;
```



```
MAXDEPTH)
survive = SurvivalProbability( diffuse );
estimation - doing it properly, closely following
if;
radiance = SampleLight( &rand, I, &L, &light;
e.x + radiance.y + radiance.z) > 0) && (acc < 0.0001)
v = true;
at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
at3 factor = diffuse * INVPI;
at weight = Mis2( directPdf, brdfPdf );
at cosThetaOut = dot( N, L );
E * ((weight * cosThetaOut) / directPdf) * (radiance
random walk - done properly, closely following Monte Carlo
vive)
```



```
at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf;
survive;
pdf;
n = E * brdf * (dot( N, R ) / pdf);
sion = true;
```

Bad Access Patterns: Linked List

The Perfect LinkedList™:

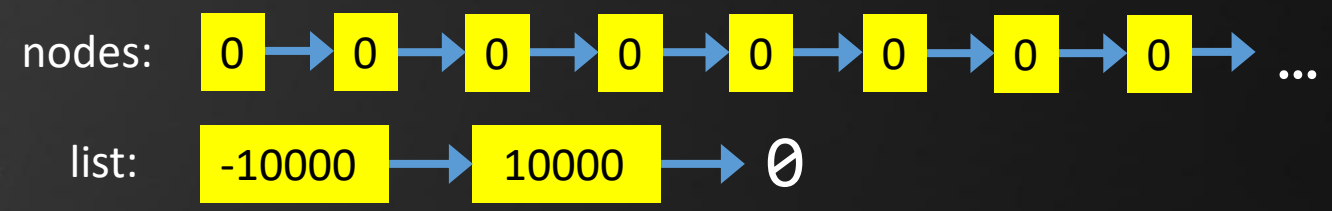
```
struct LLNode
{
    LLNode* next;
    int value;
};

LLNode* nodes = new LLNode[...];
LLNode* pool = nodes;

for( int i = 0; i < ...; i++ )
    nodes[i].next = &nodes[i + 1];
```

```
LLNode* NewNode( int value )
{
    LLNode* retval = pool;
    pool = pool->next;
    retval->value = value;
    return retval;
}

list = NewNode( -MAXINT );
list->next = NewNode( MAXINT );
list->next->next = 0;
```



DOD

Bad Access Patterns: Linked List

The Perfect LinkedList™, experiment:

Insert 25000 random values in the list so that we obtain a sorted sequence.

```
for( int i = 0; i < COUNT; i++ )
{
    LLNode* node = NewNode( rand() & 8191);
    LLNode* iter = list;
    while (iter->next->value < node->value)
        iter = iter->next;
    node->next = iter->next;
    iter->next = node;
}
```



DOD

Bad Access Patterns: Linked List

KISS Array™:

```
data = new int[...];
memset( data, 0, ... * sizeof( int ) );
data[0] = -10000;
data[1] = 10000;
N = 2;
```

```
for( int i = 0; i < COUNT; i++ )
{
    int pos = 1, value = rand() & 8191;
    while (data[pos] < value) pos++;
    memcpy( data + pos + 1,
            data + pos,
            (N - pos + 1) * sizeof( int ) );
    data[pos] = value, N++;
}
```



DOD



```

ics
& (depth < MAXDEPTH)
{
    if (inside & 1)
    {
        nt = nt / nc;
        pos2t = 1.0f - nnt;
        D, N );
    }
    at a = nt - nc, b = nt;
    at Tr = 1 - (R0 + (1 - R0)
    (Tr) R = (D * nnt - N * (D
    E * diffuse;
    = true;
    refl + refr)) && (depth < MAXDEPTH)
    D, N );
    refl * E * diffuse;
    = true;
    MAXDEPTH)
}
survive = SurvivalProbability( diffuse );
estimation - doing it properly, closely following
df;
radiance = SampleLight( &rand, I, &L, &align,
e.x + radiance.y + radiance.z) > 0) && (ace)
w = true;
at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
at3 factor = diffuse * INVPI;
at weight = Mis2( directPdf, brdfPdf );
at cosThetaOut = dot( N, L );
E * ((weight * cosThetaOut) / directPdf) * (radiance
random walk - done properly, closely following
ive)
;
at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf);
survive;
pdf;
n = E * brdf * (dot( N, R ) / pdf);
sion = true;

```

```

for( int i = 0; i < COUNT; i++ )
{
    int pos = 1, value = rand() & 8191;
    while (data[pos] < value) pos++;
    memcpy( data + pos + 1, data + pos,
            (N - pos + 1) * sizeof( int ) );
    data[pos] = value, N++;
}

```



DOD

Bad Access Patterns: Linked List*

Inserting elements in an array by shifting the remainder of the array is *significantly faster* than using an optimized linked list.

Why?

- Finding the location in the array: pure linear access
- Shifting the remainder: pure linear access.

➔ Even though the amount of transferred memory is huge, this approach wins.

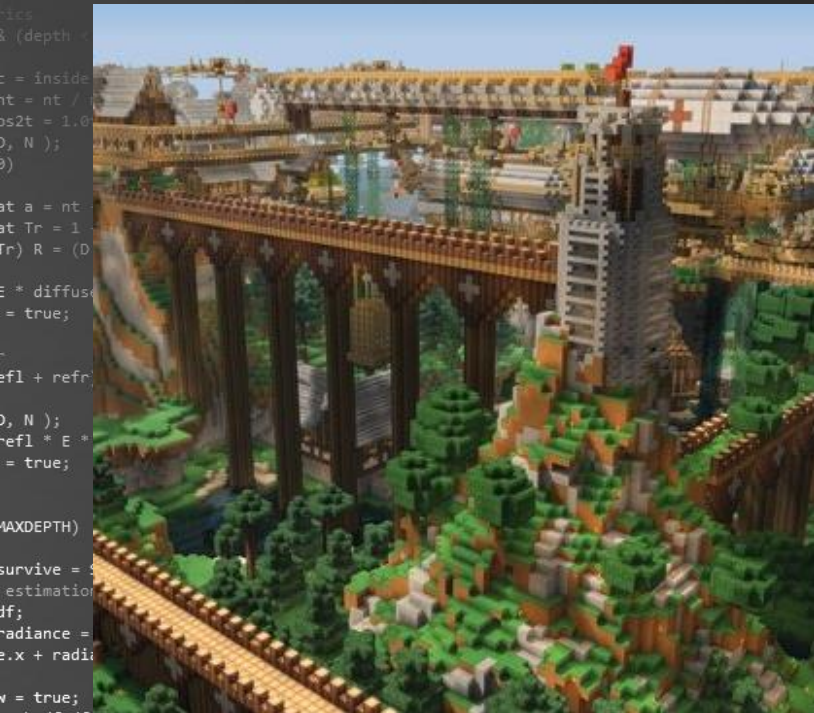
*: Also see: Nathan Reed, Data Oriented Hash Table, 2015.

<http://www.reedbeta.com/blog/data-oriented-hash-table>



DOD

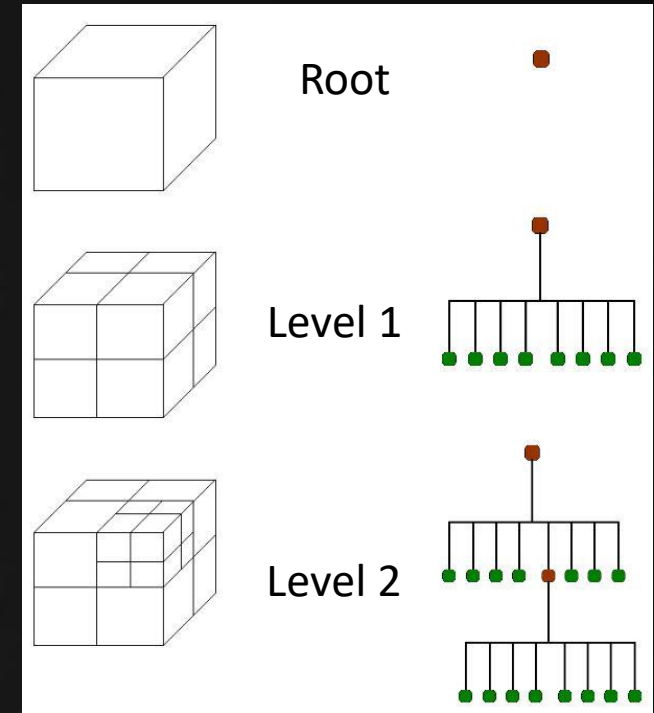
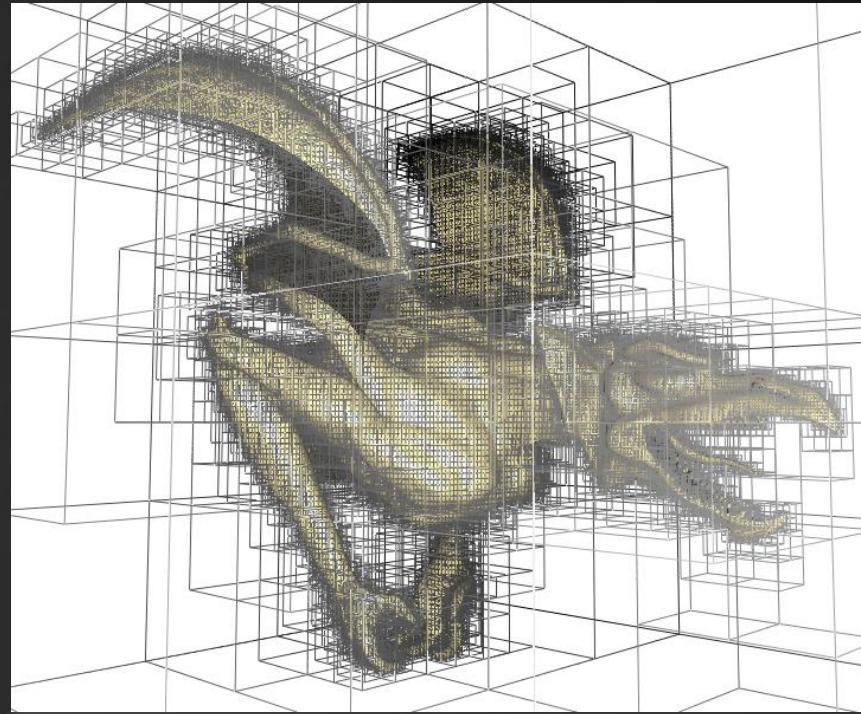
Bad Access Patterns: Octree



```

ics
& (depth
t = inside
nt = nt /
os2t = 1.0
D, N );
0);
at a = nt
at Tr = 1
Tr) R = (D
E * diffuse
= true;
efl + refr
D, N );
refl * E *
= true;
MAXDEPTH)
survive =
estimation
df;
radiance =
e.x + radi
w = true;
at brdfPdf = EvaluateDiffuse(
at3 factor = diffuse * INVPI;
at weight = Mis2( directPdf, brdfPdf );
at cosThetaOut = dot( N, L );
E * ((weight * cosThetaOut) / directPdf) * (rad
random walk - done properly, closely following
ive)
;
at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf
survive;
pdf;
n = E * brdf * (dot( N, R ) / pdf);
sion = true;

```



DOD

Bad Access Patterns: Octree

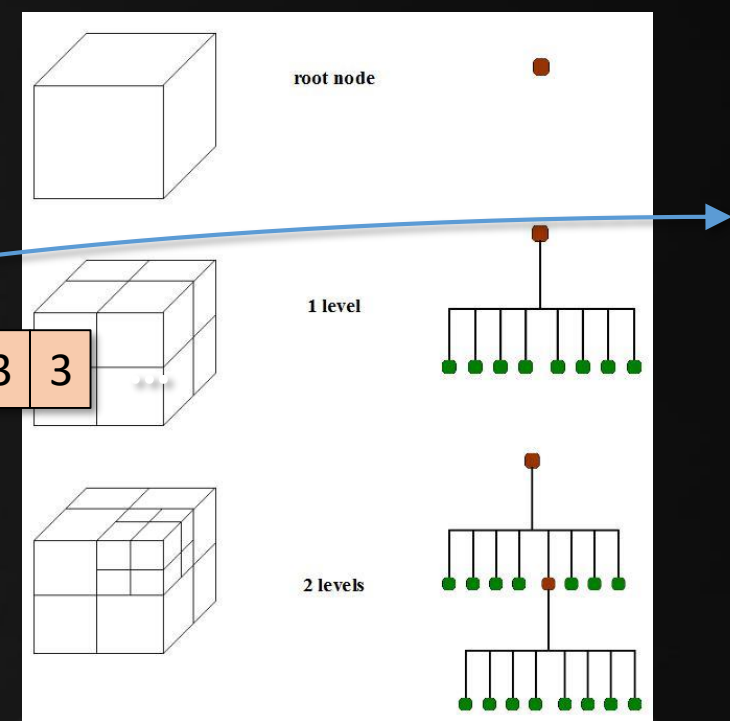
Query: find the color of a voxel visible through pixel (x,y).

Operation: ‘3DDDA’ (basically: Bresenham).

Data layout:



Color data: 32-bit (ARGB).



DOD

Bad Access Patterns: Octree

Alternative layout:

1. Tree 1: occlusion (1 bit per voxel);
2. Tree 2: color information (32 bits per voxel).

Use tree 1 to find the voxel you are looking for.
Lookup the correct voxel (incurring a single cache miss) in tree 2.

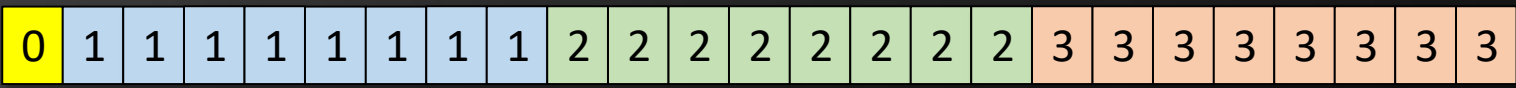
- Caching in tree 1:
- A cache line holds 64*8=512 voxels
 - Accessing the root gets several levels in L1 cache



DOD

Bad Access Patterns: Octree

Alternative layout (part 2):



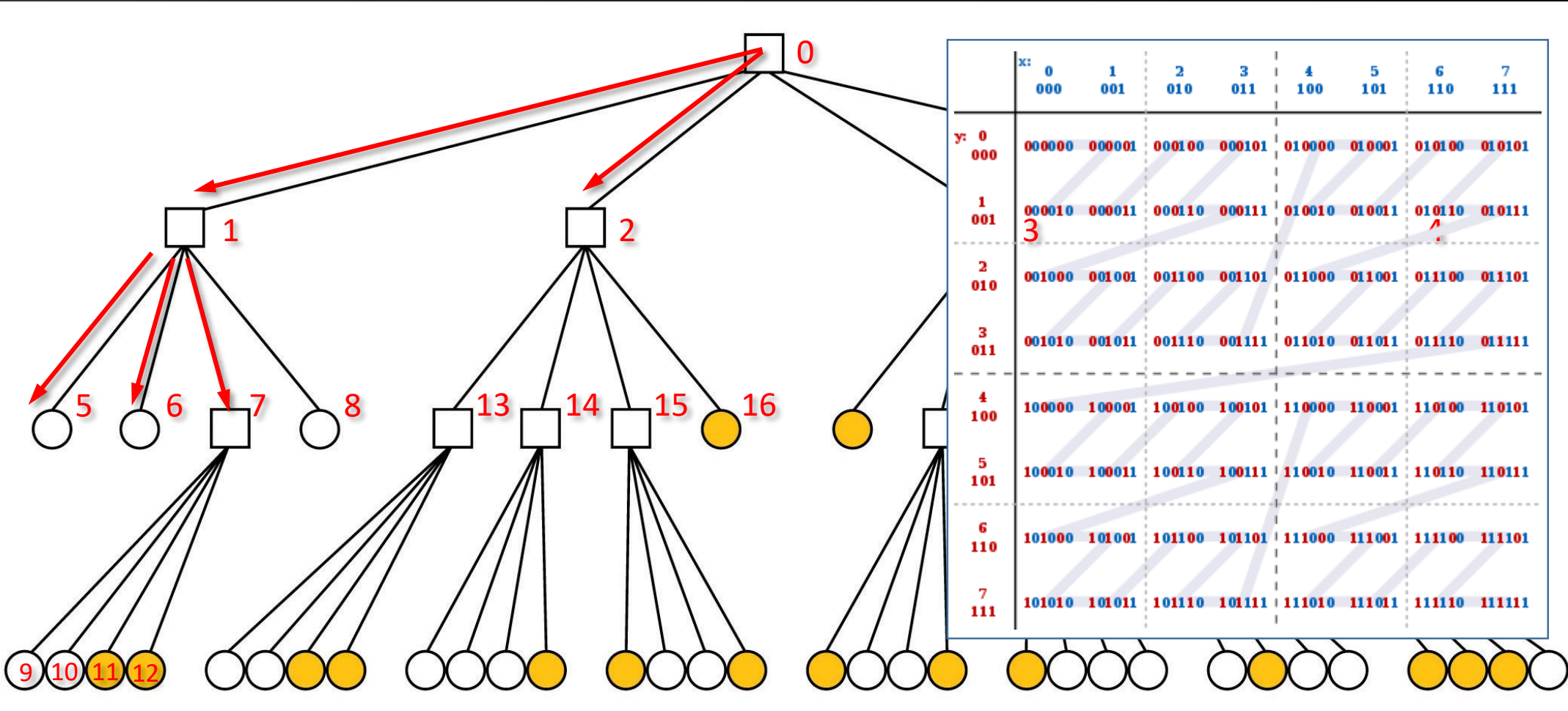
Trees are typically generated by a divide-and-conquer algorithm, in a depth-first fashion.

Compact storage:

```
struct OTNode
{
    int firstChild;
    // bit 31 set: empty
};
```



DOD



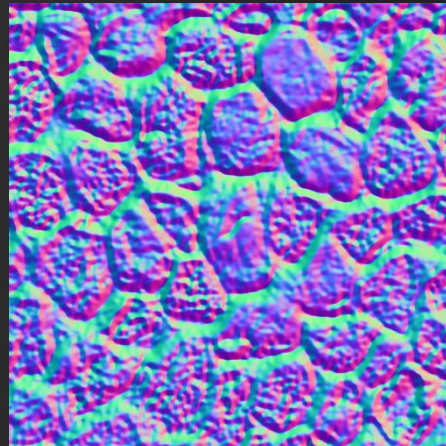
DOD

Bad Access Patterns: Textures in a Ray Tracer

Typical process for tracing a ray:

- Traverse a tree (multiple kilobytes)
- Intersect triangles in the leaf nodes (quite a few bytes)
- If a hit is found, fetch texture.

This is almost always a cache miss.



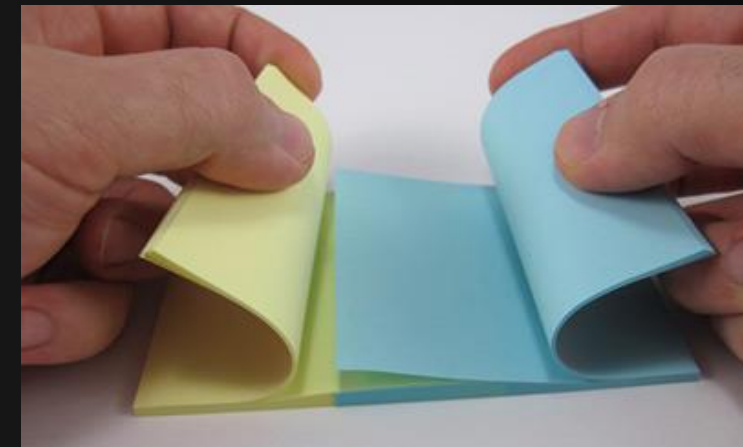
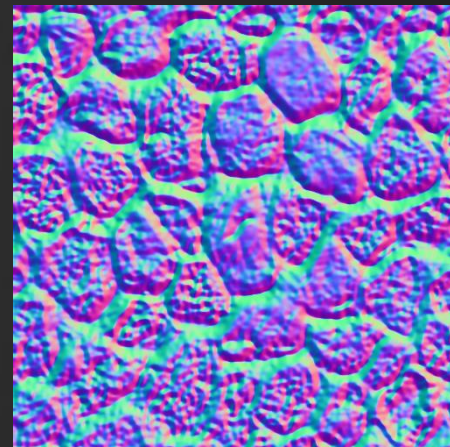
DOD

Bad Access Patterns: Textures in a Ray Tracer

We suffer the cache miss *twice*:

- Once for the texture;
- Once for the normal map.

Note: both values are 32-bit.



DOD

Interleaved texture / normal:

- One value now becomes 64-bit and contains the normal and the color.
- We still suffer a cache miss –
- But only once.



DOD

Previously in INFOMOV

```
struct Particle
{
    float x, y, z;
    float vx, vy, vz;
    float mass;
};
// size: 28 bytes
```

Better:

```
struct Particle
{
    float x, y, z;
    float vx, vy, vz;
    float mass, dummy;
};
// size: 32 bytes
```



DOD

Previously in INFOMOV

```
struct Particle
{
    float x, y, z;
    int mass;
};
Particle particle[512];
```

AOS

```
union { float x[512]; __m128 x4[128]; };
union { float y[512]; __m128 y4[128]; };
union { float z[512]; __m128 z4[128]; };
union { int mass[512]; __m128i mass4[128]; };
```

SOA

structure of arrays



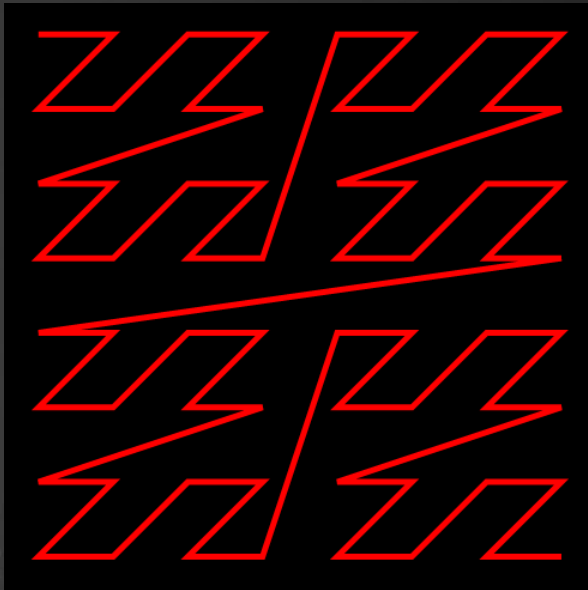
DOD

Previously in INFOMOV

```

ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 1.2)
    {
        nt = nt / nc; ddn = ddn * nc;
        ps2t = 1.0f - nnt * nnt;
        D, N );
    }
    at a = nt - nc, b = nt + nc;
    at Tr = 1 - (R0 + (1 - R0) *
    (Tr) R = (D * nnt - N * (ddn *
    E * diffuse;
    = true;
    -
    refl + refr)) && (depth < MAXDEPTH)
    D, N );
    refl * E * diffuse;
    = true;
    MAXDEPTH)
    survive = SurvivalProbability( diffuse );
    estimation - doing it properly, closely
    df;
    radiance = SampleLight( &rand, I, &L, &align
    e.x + radiance.y + radiance.z) > 0) && (doc
    w = true;
    at brdfPdf = EvaluateDiffuse( L, N ) * Psum;
    at3 factor = diffuse * INVPI;
    at weight = Mis2( directPdf, brdfPdf );
    at cosThetaOut = dot( N, L );
    E * ((weight * cosThetaOut) / directPdf) * (radiance
    random walk - done properly, closely following 3rd
    vive)
    ;
    at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf;
    survive;
    pdf;
    n = E * brdf * (dot( N, R ) / pdf);
    sion = true;

```



Method:

X = 1 1 0 0 0 1 0 1 1 0 1 1 0 1

Y = 1 0 1 1 0 1 1 0 1 0 1 1 1 0

M = 1101101000111001110011111001



Today's Agenda:

- OOP Performance Pitfalls
- DOD Concepts
- DOD or OO?



2B|~2B

OO = Evil, DO = Good?

10% of your code runs 90% of the time. DO is good for this 10%.

For all other code, please:

- Use STL
- Apply OO
- Program in C#
- Use event handling
- Check return values
- Focus on productivity



2B|~2B

```

ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 0.2)
    {
        nt = nt / nc; ddn = ddn * ddn;
        r = 1.0f - nnt * nnt;
        D, N );
    }
    at a = nt - nc, b = nt - nc;
    at Tr = 1 - (R0 + (1 - R0) * r);
    Tr) R = (D * nnt - N * (ddn * nnt));
    E * diffuse;
    = true;
    defl + refr)) && (depth < MAXDEPTH)
    {
        D, N );
        refl * E * diffuse;
        = true;
    }
    MAXDEPTH)
    survive = SurvivalProbability( diffuse
    estimation - doing it properly, close
    if;
    radiance = SampleLight( &rand, I, &L,
    e.x + radiance.y + radiance.z) > 0) &&
    w = true;
    at brdfPdf = EvaluateDiffuse( L, N ) *
    at3 factor = diffuse * INVPI;
    at weight = Mis2( directPdf, brdfPdf );
    at cosThetaOut = dot( N, L );
    E * ((weight * cosThetaOut) / directPdf);
    random walk - done properly, closely following 3rd order
    vive)
    ;
    at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf );
    survive;
    pdf;
    n = E * brdf * (dot( N, R ) / pdf);
    sion = true;
    
```



DATA-ORIENTED DESIGN AND C++

Presenter: Mike Acton

<https://www.youtube.com/watch?v=rX0ltVEVjHc>



2B|~2B

[Next](#) [Up](#) [Previous](#) [Contents](#)

Next: [Contents](#) [Contents](#)

Online release of Data-Oriented Design :

This is the free, online, reduced version. Some inessential chapters are excluded from this version, but in the spirit of this being an education resource, the essentials are present for anyone wanting to learn about data-oriented design.

Expect some odd formatting and some broken images and listings as this is auto generated and the Latex to html converters available are not perfect. If the source code listing is broken, you should be able to find the referenced source on [github](#). If you like what you read here, consider purchasing the real paper book from [here](#), as not only will it look a lot better, but it will help keep this version online for those who cannot afford to buy it. Please send any feedback to support@dataorienteddesign.com

Data-Oriented Design

Richard Fabian

- [Contents](#)
- [Data-Oriented Design](#)
 - [It's all about the data](#)
 - [Data is not the problem domain](#)
 - [Data and statistics](#)
 - [Data can change](#)
 - [How is data formed?](#)
 - [The framework](#)
 - [Conclusions and takeaways](#)

<http://www.dataorienteddesign.com/dodbook/>



2B|~2B

```

ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 0)
    {
        nt = nt / nc; ddn = sqrt(1 - nt);
        r = 1.0f - nnt * 0.5f;
        D, N );
    }
    else
    {
        at a = nt - nc, b = nt + nc;
        at Tr = 1 - (R0 + (1 - R0) * r);
        Tr) R = (D * nnt - N * (ddn *
    }
    E * diffuse;
    = true;
}
efl + refr)) && (depth < MAXDEPTH)
{
    D, N );
    refl * E * diffuse;
    = true;
}
MAXDEPTH)
survive = SurvivalProbability( diffuse
estimation - doing it properly, close
df;
radiance = SampleLight( &rand, I, &L,
e.x + radiance.y + radiance.z) > 0) &&
w = true;
at brdfPdf = EvaluateDiffuse( L, N ) *
at3 factor = diffuse * INVPI;
at weight = Mis2( directPdf, brdfPdf );
at cosThetaOut = dot( N, L );
E * ((weight * cosThetaOut) / directP
andom walk - done properly, closely fo
vive)
;
at3 brdf = SampleDiffuse( diffuse, N,
survive;
pdf;
n = E * brdf * (dot( N, R ) / pdf);
sion = true;

```

Data Oriented Design Resources

A curated list of awesome data oriented design resources.

Feel free to contribute by sending PR!

- [Data Oriented Design](#)
 - [Presentations](#)
 - [Blog Posts](#)
 - [Videos](#)
 - [Other](#)
 - [Code Examples](#)

Presentations

- [A Step Towards Data Orientation \(2010\)](#) - Johan Torp
- [Introduction To Data Oriented Design \(2010\)](#) - DICE
- [Memory Optimization \(2003\)](#) - Christer Ericson
- [Practical Examples In Data Oriented Design \(2013\)](#) - Niklas Frykholm
- [Three Big Lies \(2008\)](#) - Mike Acton
- [Typical C++ Bullshit \(2008\)](#) - Mike Acton
- [Data-Oriented Design and C++ \(2014\)](#) - Mike Acton
- [Entity Component Systems & Data Oriented Design \(2018\)](#) - Aras Pranckevičius

Blog Posts

<https://github.com/dbartolini/data-oriented-design>

- [Adventures in data-oriented design – Part 1: Mesh data \(2011\)](#) - Stefan Reinalter

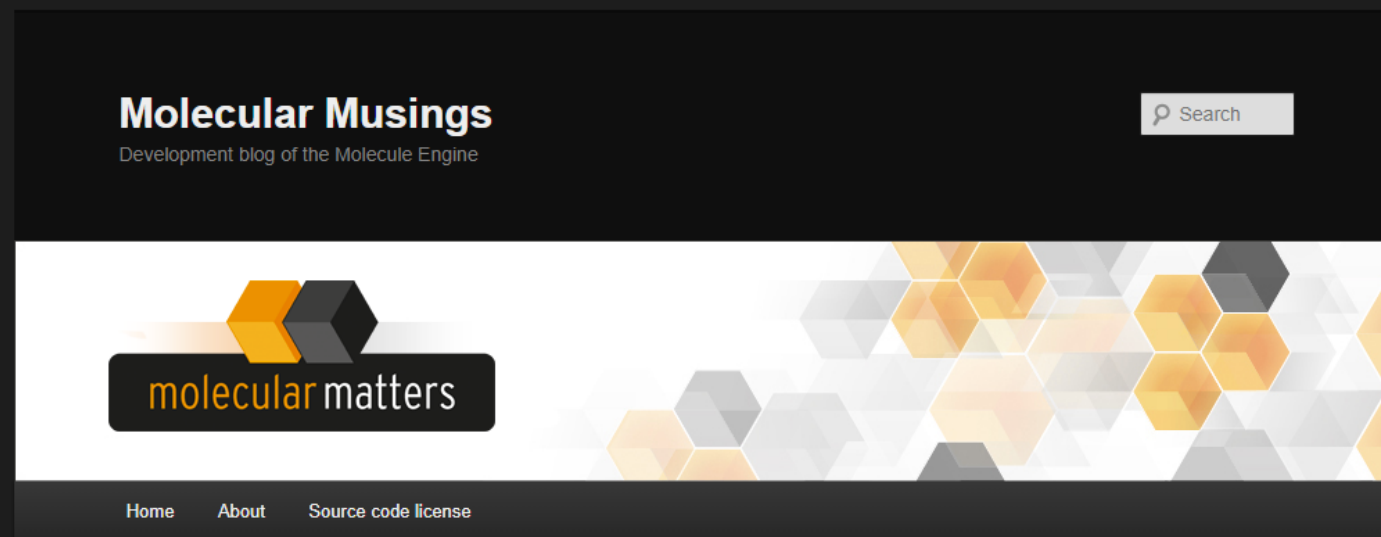


2B|~2B

```

ics
& (depth < MAXDEPTH)
{
    if (inside ? 1 : 0)
    {
        nt = nt / nc; ddn = ddn * nc;
        pos2t = 1.0f + nnt * nnt;
        D, N );
    }
    at a = nt - nc, b = nt + nc;
    at Tr = 1 - (R0 + (1 - R0) * pos2t);
    R = (D * nnt - N * (ddn * nnt));
    E * diffuse;
    = true;
    refl + refr)) && (depth < MAXDEPTH)
    D, N );
    refl * E * diffuse;
    = true;
    MAXDEPTH)
    survive = SurvivalProbability( diffuse );
    estimation - doing it properly, closely following
    if;
    radiance = SampleLight( &rand, I, &L, &light);
    e.x + radiance.y + radiance.z) > 0) && (depth < MAXDEPTH)
    w = true;
    at brdfPdf = EvaluateDiffuse( L, N ) * Psurvive;
    at3 factor = diffuse * INVPI;
    at weight = Mis2( directPdf, brdfPdf );
    at cosThetaOut = dot( N, L );
    E * ((weight * cosThetaOut) / directPdf) * (radiance
    random walk - done properly, closely following 3rd
    ive)
    ;
    at3 brdf = SampleDiffuse( diffuse, N, r1, r2, &R, &pdf);
    survive;
    pdf;
    n = E * brdf * (dot( N, R ) / pdf);
    sion = true;

```



<https://blog.molecular-matters.com/2011/11/03/adventures-in-data-oriented-design-part-1-mesh-data-3/>
<https://blog.molecular-matters.com/2013/02/22/adventures-in-data-oriented-design-part-2-hierarchical-data/>
<https://blog.molecular-matters.com/2013/05/02/adventures-in-data-oriented-design-part-3a-ownership/>
<https://blog.molecular-matters.com/2013/05/17/adventures-in-data-oriented-design-part-3b-internal-references/>

Adventures in data-oriented design – Part 1: Mesh data

Let's face it, performance on modern processors (be it PCs, consoles or mobiles) is mostly governed by memory access patterns. Still, data-oriented design is considered something new and novel, and only slowly creeps into programmers' brains, and this really needs to change. Having co-workers fix your code and improving its performance really is no excuse for writing crappy code (from a performance point-of-view) in the first place.

This post is the first in an ongoing series about how certain things in the Molecule Engine are done in a data-oriented fashion, while still making use of OOP concepts. A common misconception about data-oriented design is that it is “C-like”, and “not OOP”, and therefore less maintainable – but that need not be the case. The concrete example we are going to look at today is how to organize your mesh data, but let's start with the pre-requisites first.



Today's Agenda:

- OOP Performance Pitfalls
- DOD Concepts
- DOD or OO?



/INFOMOV/

END of “Data-Oriented Design”

next lecture: “GPGPU (1)”

