

Mind the cache



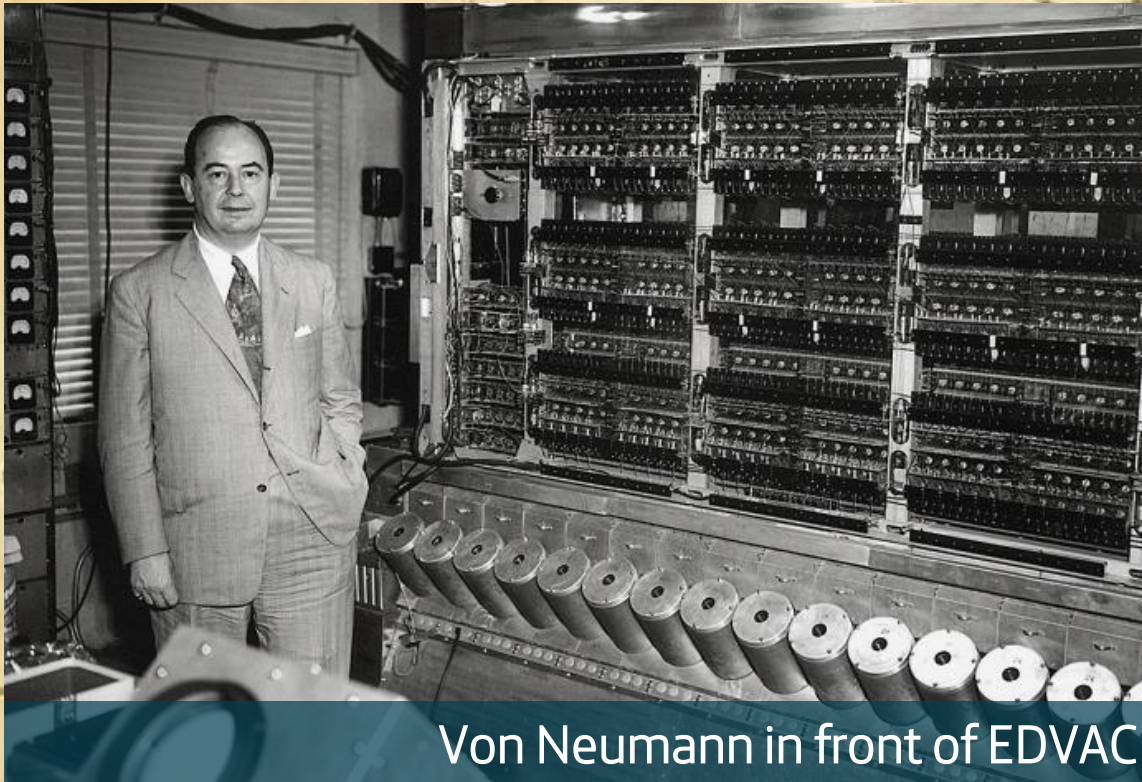
using `std::cpp` 2015

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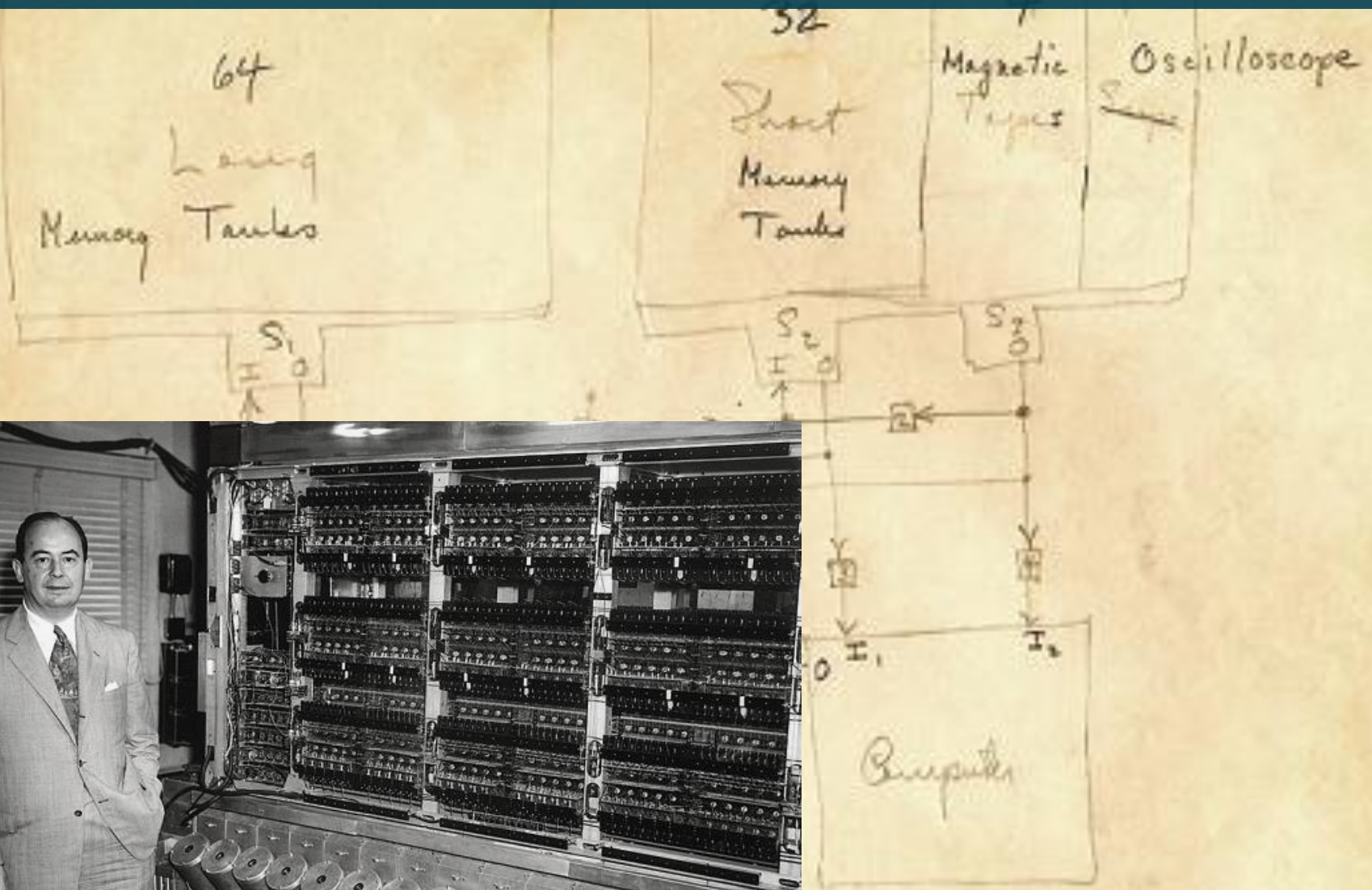
Madrid, November 2015

Telefonica

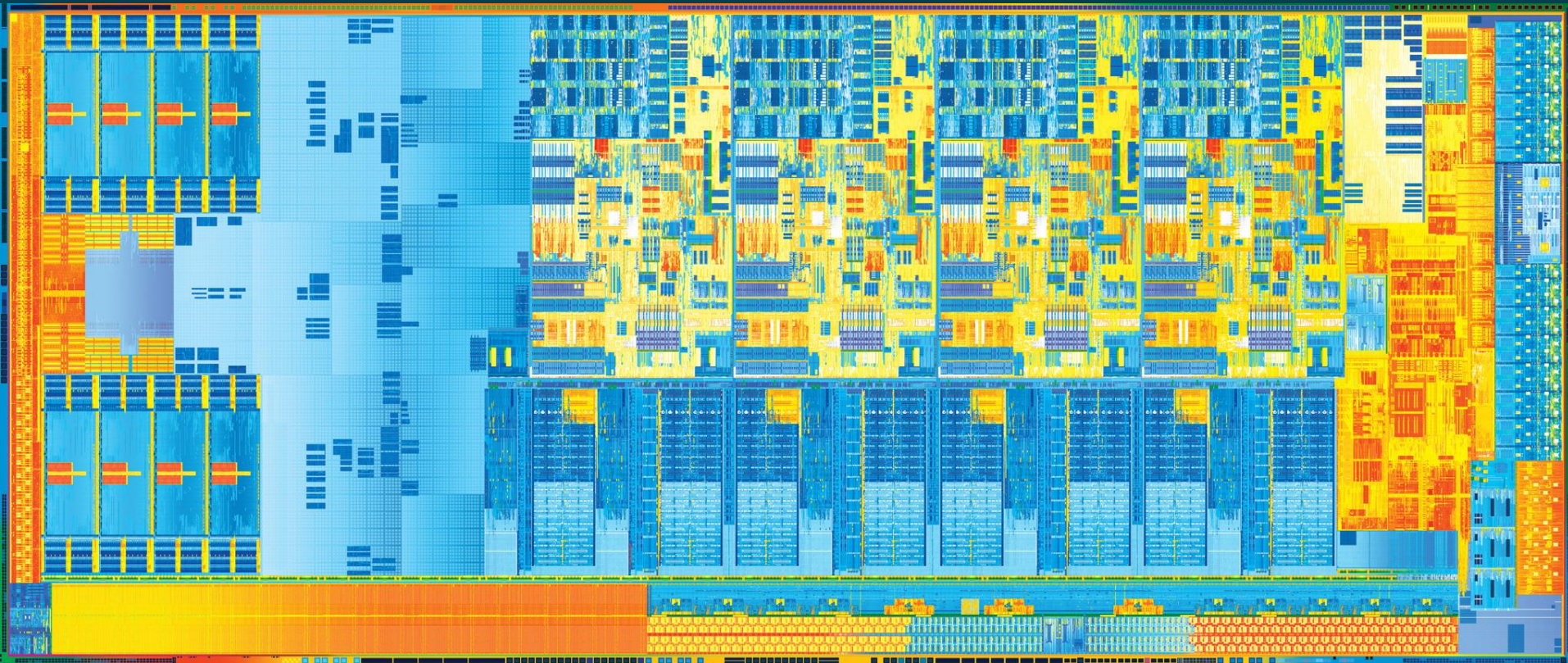
Our mental model of the CPU is 70 years old



Von Neumann in front of EDVAC

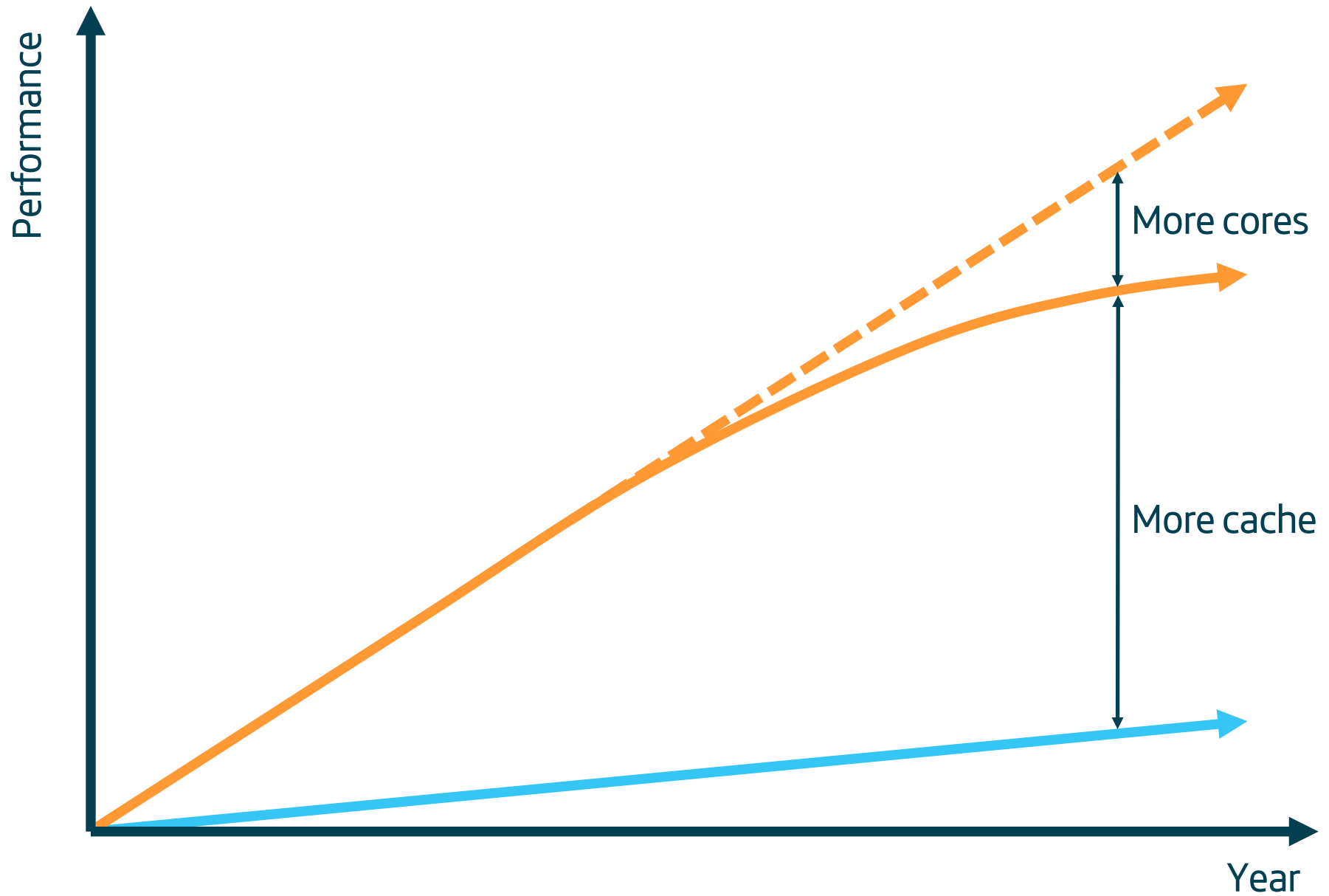


Current reality looks quite different

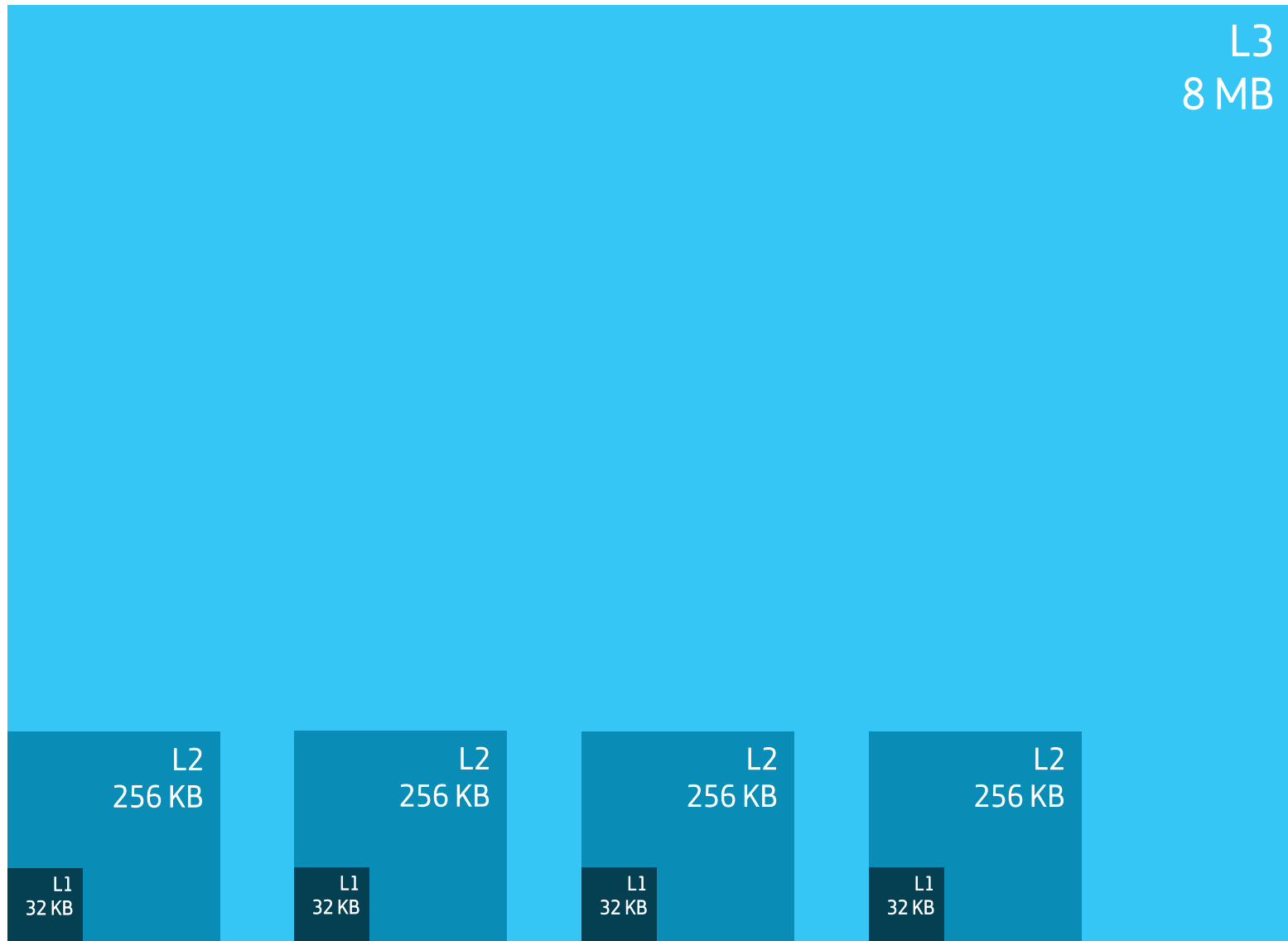


Intel Core i7-3770K (Ivy Bridge)

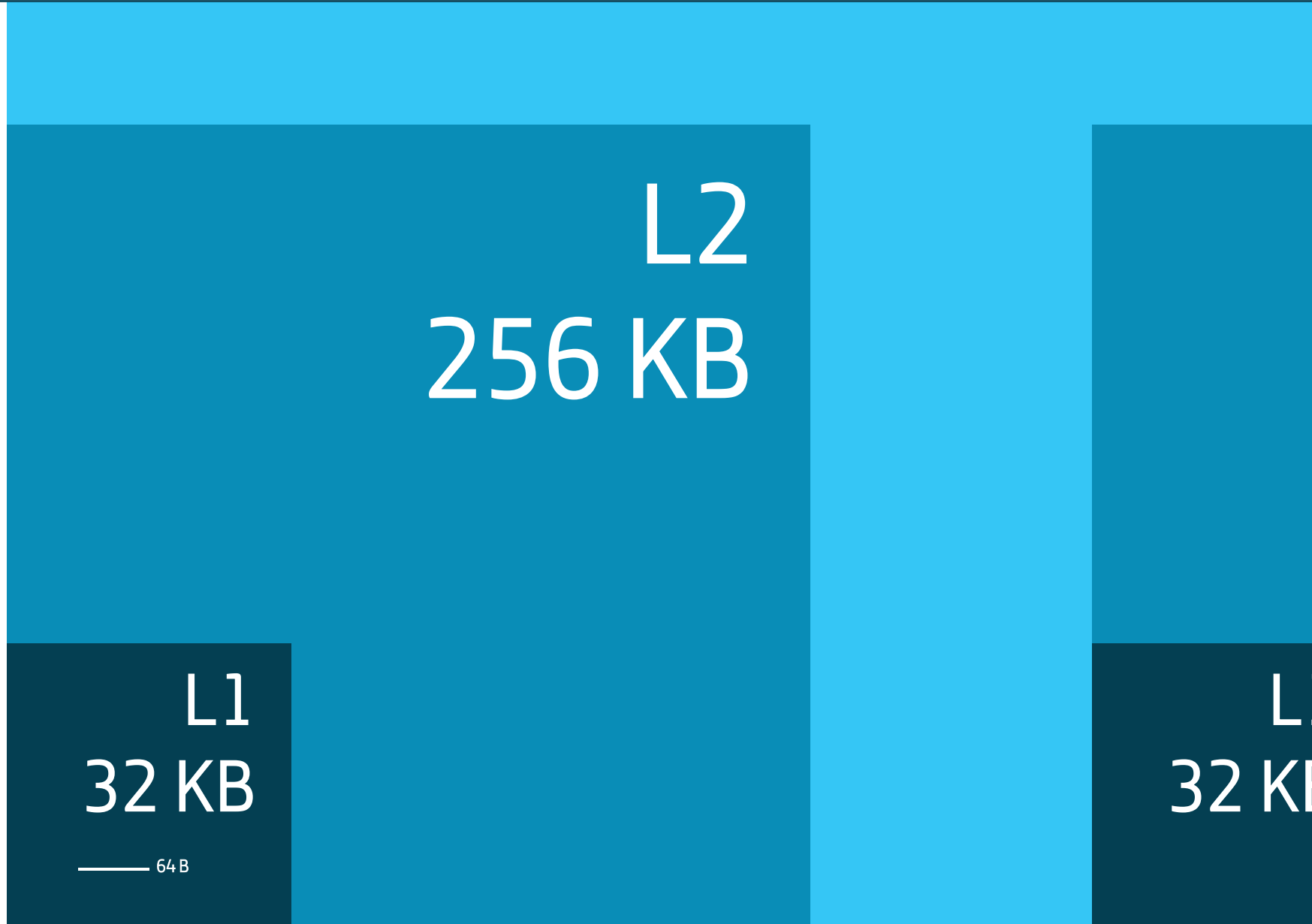
Processor-memory gap



Cache hierarchy



Cache line



Access times



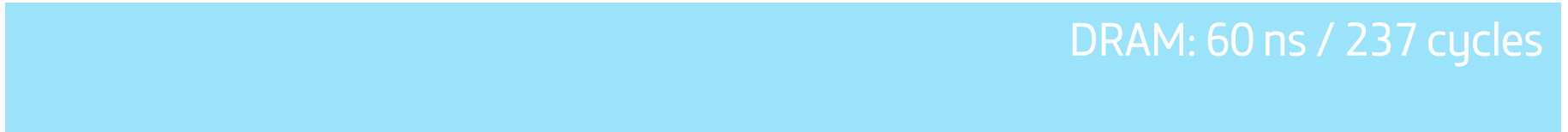
L1: 1 ns / 4 cycles



L2: 3.1 ns / 12 cycles



L3: 7.7 ns / 30 cycles



DRAM: 60 ns / 237 cycles

Measure measure measure



Our little measuring tool

```
template<typename F>
double measure(F f)
{
    using namespace std::chrono;

    static const int          num_trials=10;
    static const milliseconds min_time_per_trial(200);
    std::array<double,num_trials> trials;
    volatile decltype(f())    res; // to avoid optimizing f() away

    for(int i=0;i<num_trials;++i){
        int                runs=0;
        high_resolution_clock::time_point t2;

        measure_start=high_resolution_clock::now();
        do{
            res=f();
            ++runs;
            t2=high_resolution_clock::now();
        }while(t2-measure_start<min_time_per_trial);
        trials[i]=duration_cast<duration<double>>(t2-measure_start).count()/runs;
    }
    (void)(res); // var not used warn

    std::sort(trials.begin(),trials.end());
    return std::accumulate(trials.begin()+2,trials.end()-2,0.0)/(trials.size()-4)*1E6;
}
```

Profiling environment

- Intel Core i5-2520M @ 2.5 GHz (Sandy Bridge)
 - L1: 2×32 KB 8-way 3 cycles
 - L2: 2×256 KB 8-way 9 cycles
 - L3: shared 3 MB 12-way 22 cycles
- 4.0 GB DRAM
- Windows 7 Enterprise
- Visual Studio Express 2015
 - x86 mode (32 bit), default release settings

Container traversal



Linear traversal

// vector

```
std::vector<int> v=...;  
return std::accumulate(v.begin(),v.end(),0);
```

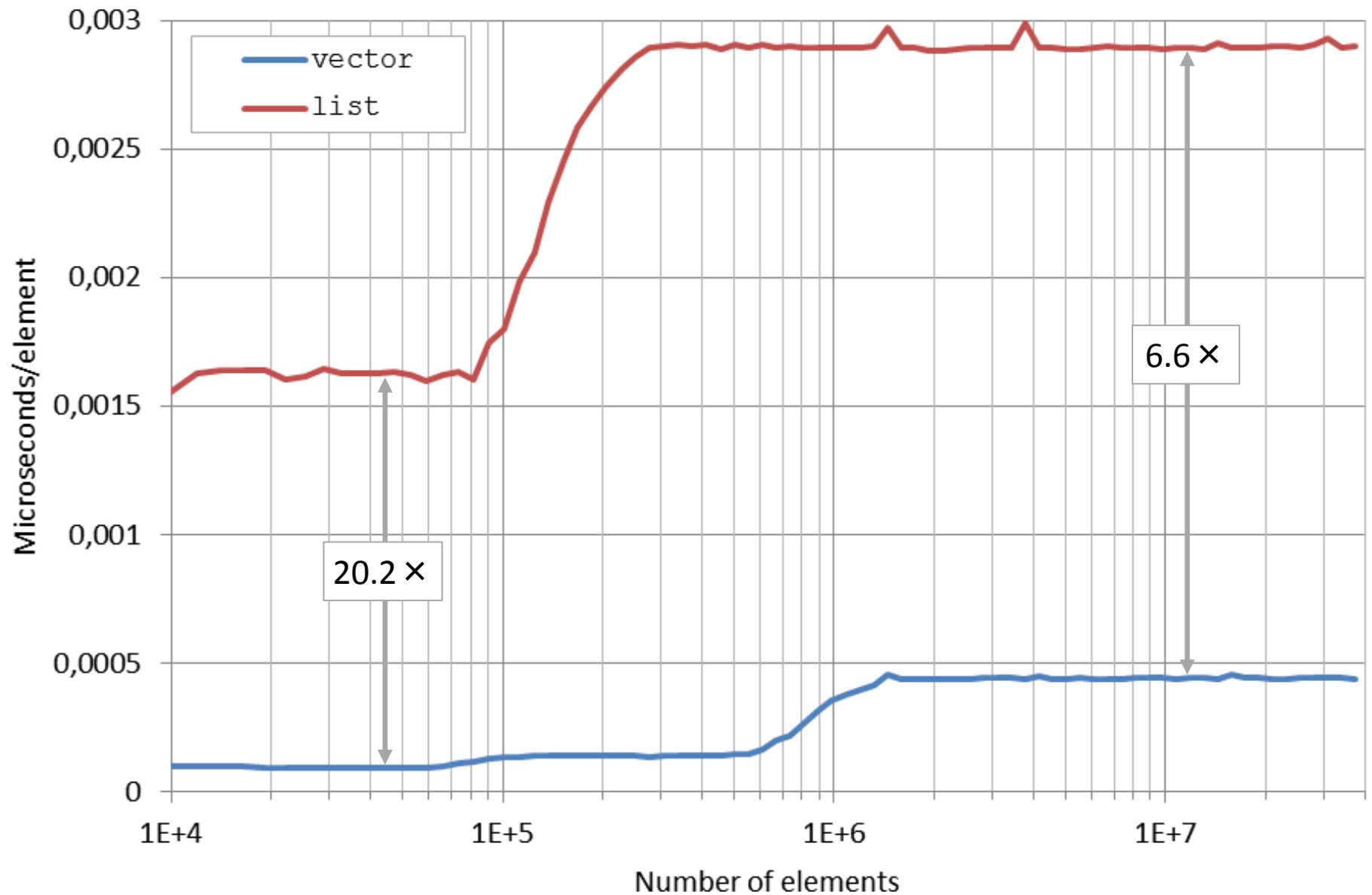
// list

```
std::list<int> l=...;  
return std::accumulate(l.begin(),l.end(),0);
```

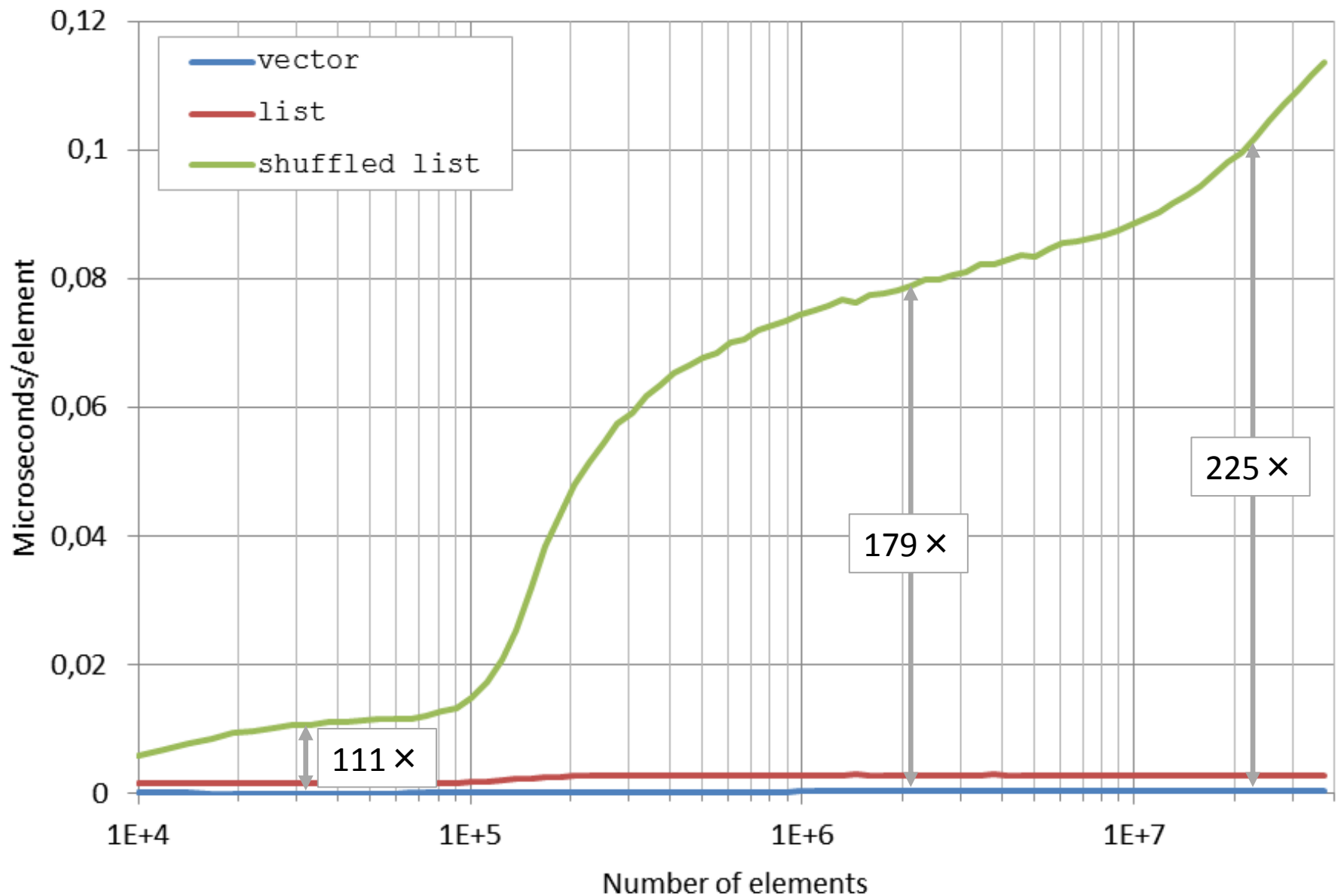
// shuffled list

```
std::list<int> l=...; // nodes shuffled in memory  
return std::accumulate(l.begin(),l.end(),0);
```

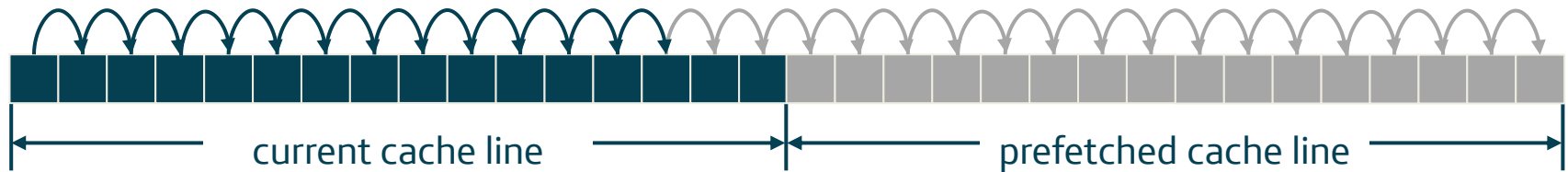
Linear traversal: results



Linear traversal: results



Linear traversal: analysis



- Two cache aspects to watch out for
 - Locality
 - Prefetching
- `std::vector` excels at both
- `std::list`: sequentially allocated nodes provide some sort of non-guaranteed locality
- Shuffled nodes is the worst scenario

Matrix sum

```
boost::multi_array<int,2> a(boost::extents[m][m]);
```

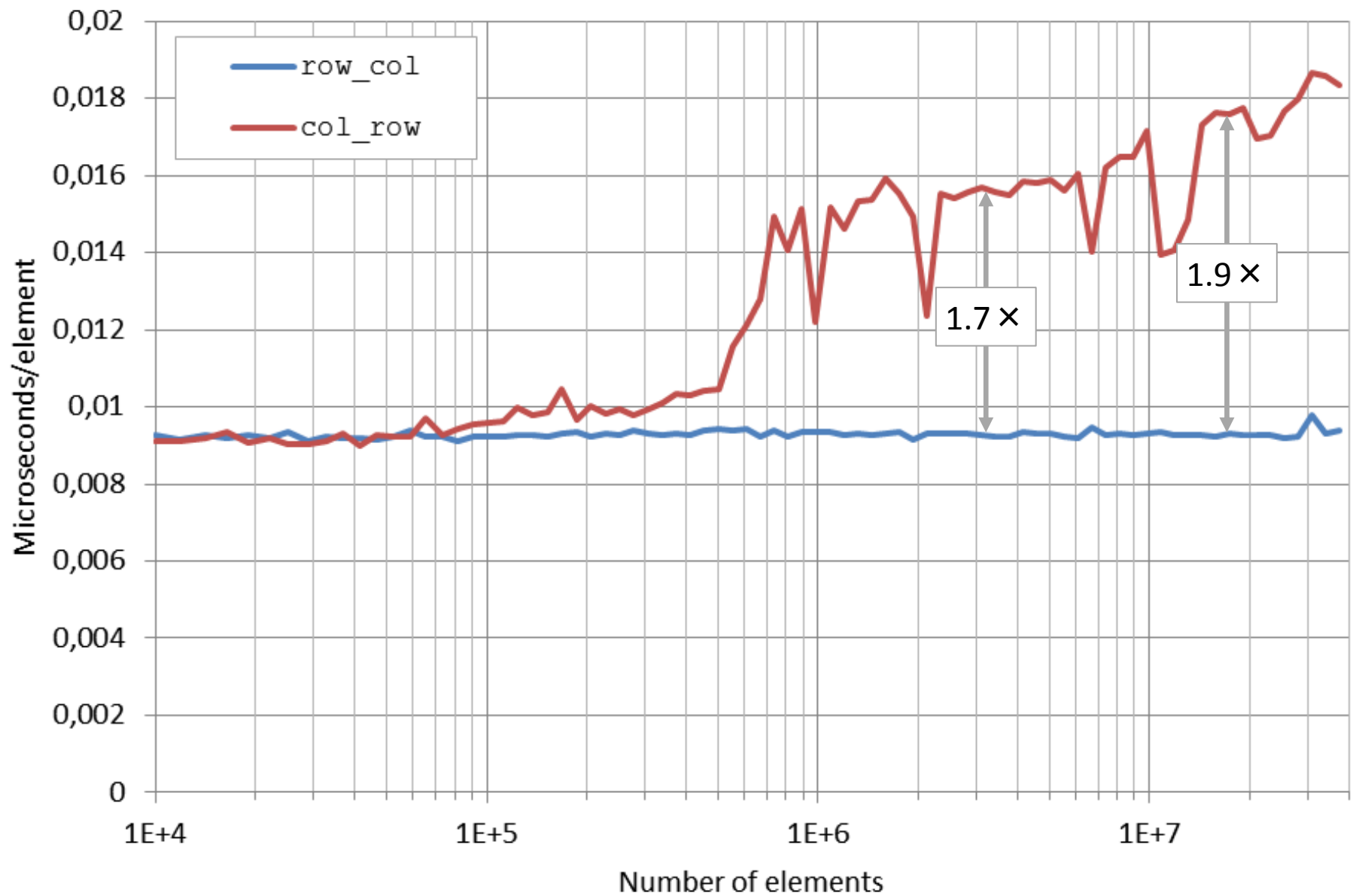
```
// row_col
```

```
long int res=0;
for(std::size_t i=0;i<m;++i){
    for(std::size_t j=0;j<m;++j){
        res+=a[i][j];
    }
}
return res;
```

```
// col_row
```

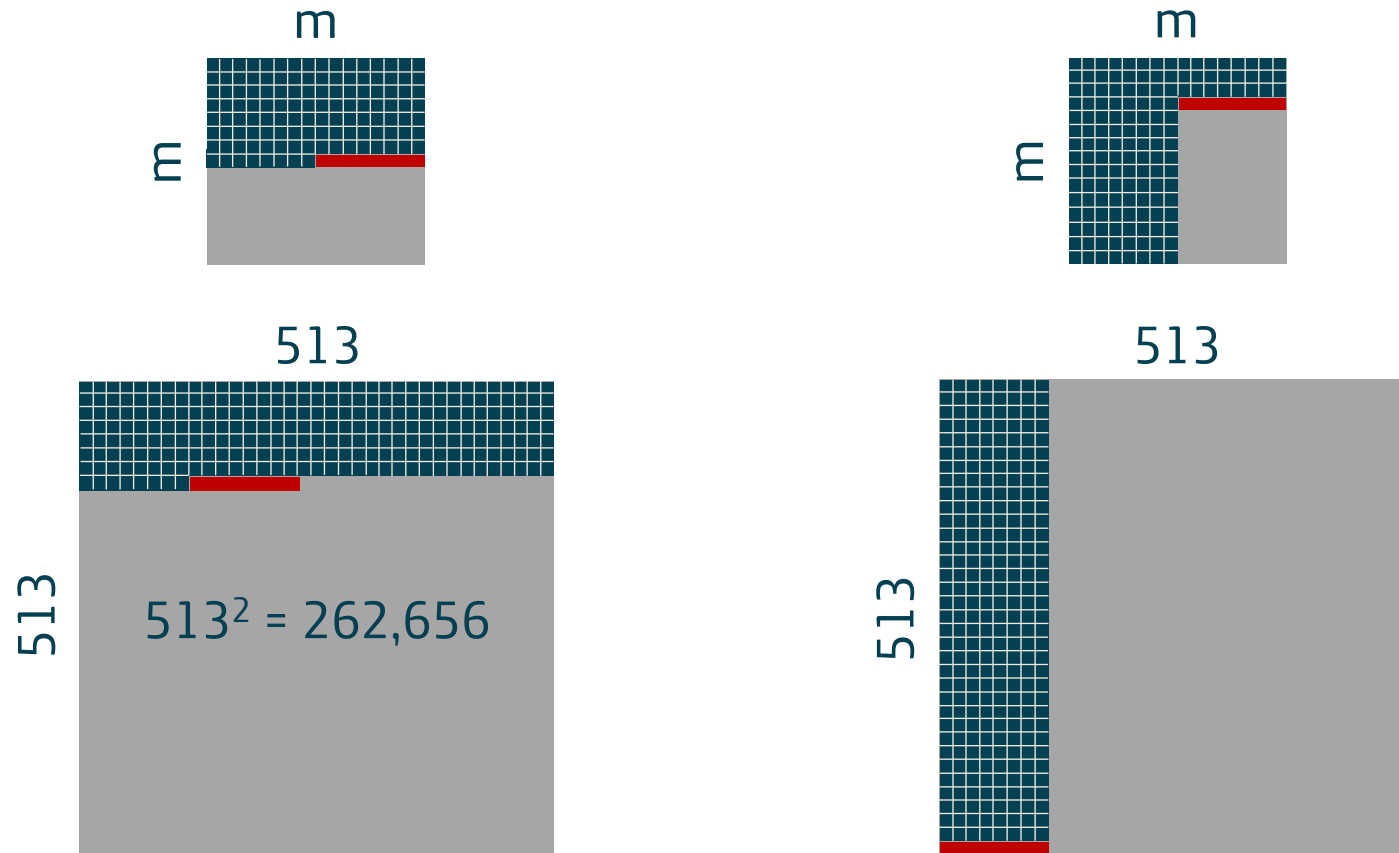
```
long int res=0;
for(std::size_t j=0;j<m;++j){
    for(std::size_t i=0;i<m;++i){
        res+=a[i][j];
    }
}
return res;
```

Matrix sum: results



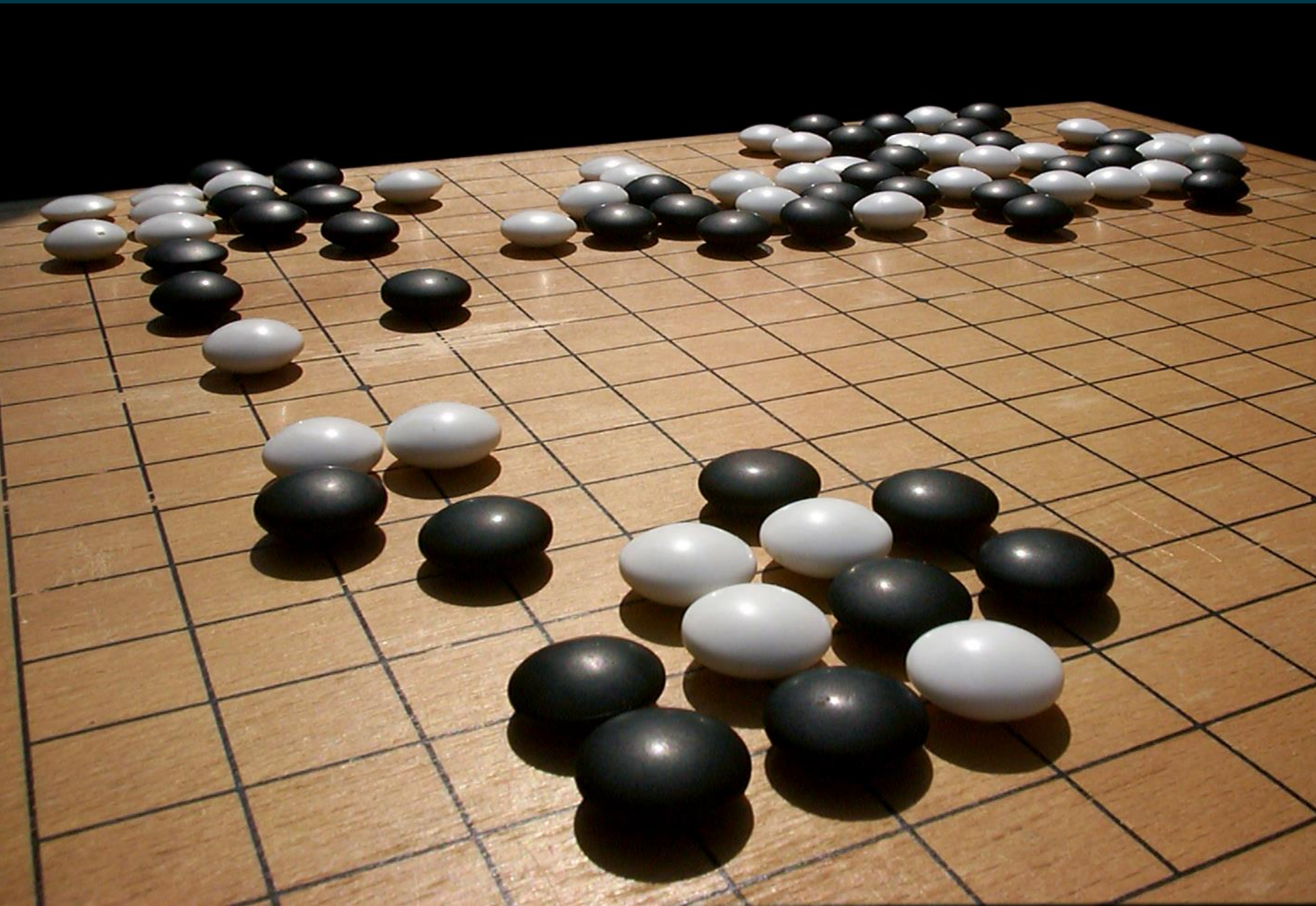
Matrix sum: analysis

- Max # of segments in L1: $32 \text{ KB} / 64 \text{ B} = 512$



- L2 saturation @ $4097^2 = 16,785,409$
- Partial cache associativity makes things actually worse

Layout tricks



AOS vs SOA

// aos

```
struct particle
{
    int x,y,z;
    int dx,dy,dz;
};
```

```
using particle_aos=std::vector<particle>;
```

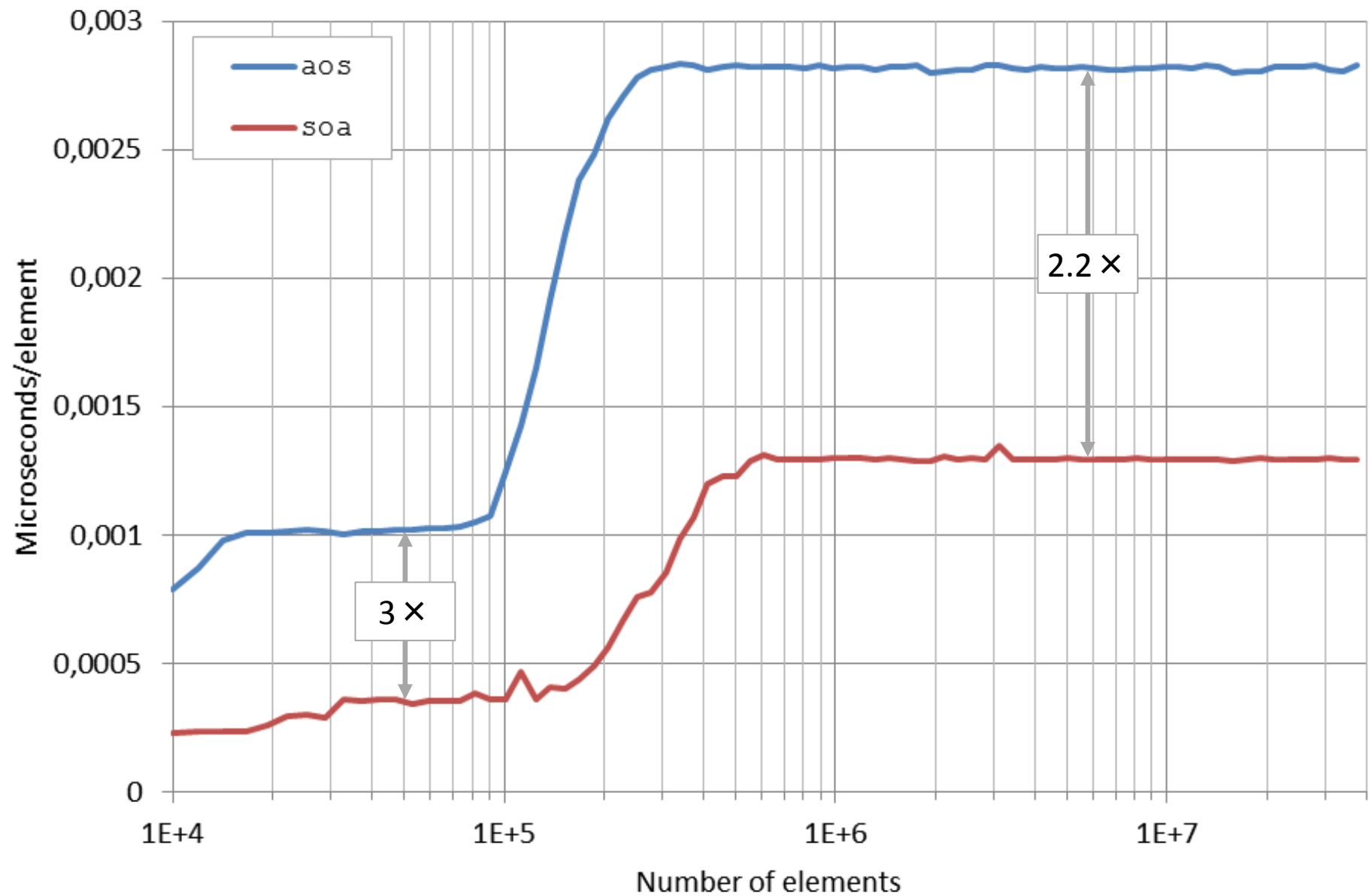
```
auto      ps=create_particle_aos(n);
long int  res=0;
for(std::size_t i=0;i<n;++i)res+=ps[i].x+ps[i].y+ps[i].z;
return res;
```

// soa

```
struct particle_soa
{
    std::vector<int> x,y,z;
    std::vector<int> dx,dy,dz;
};
```

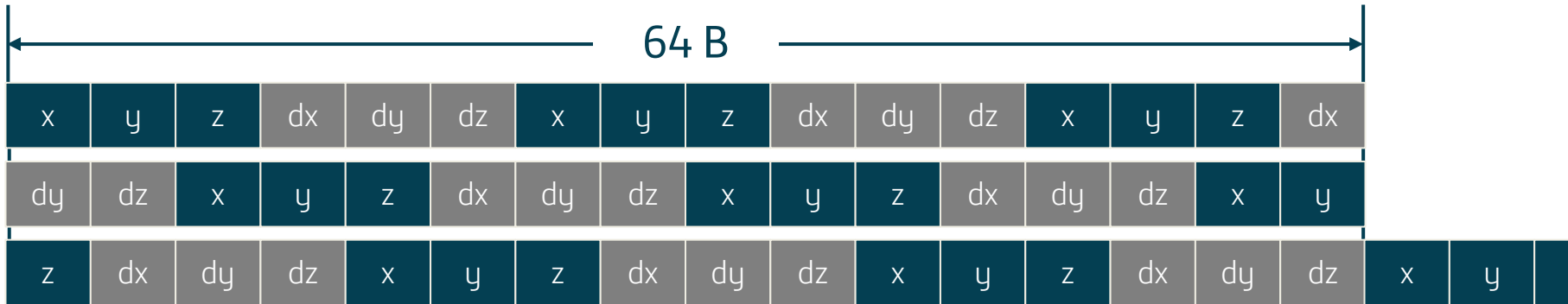
```
auto      ps=create_particle_soa(n);
long int  res=0;
for(std::size_t i=0;i<n;++i)res+=ps.x[i]+ps.y[i]+ps.z[i];
return res;
```


AOS vs SOA: results



AOS vs SOA: analysis

- AOS cache load throughput: 8 *useful* values per fetch (average)



- SOA throughput: 16 useful values per fetch



- $16/8 = 2$ times faster
- Yet results are even better than this!

Compact AOS vs SOA

```
// aos
```

```
struct particle
{
    int x,y,z;
};
```

```
using particle_aos=std::vector<particle>;
```

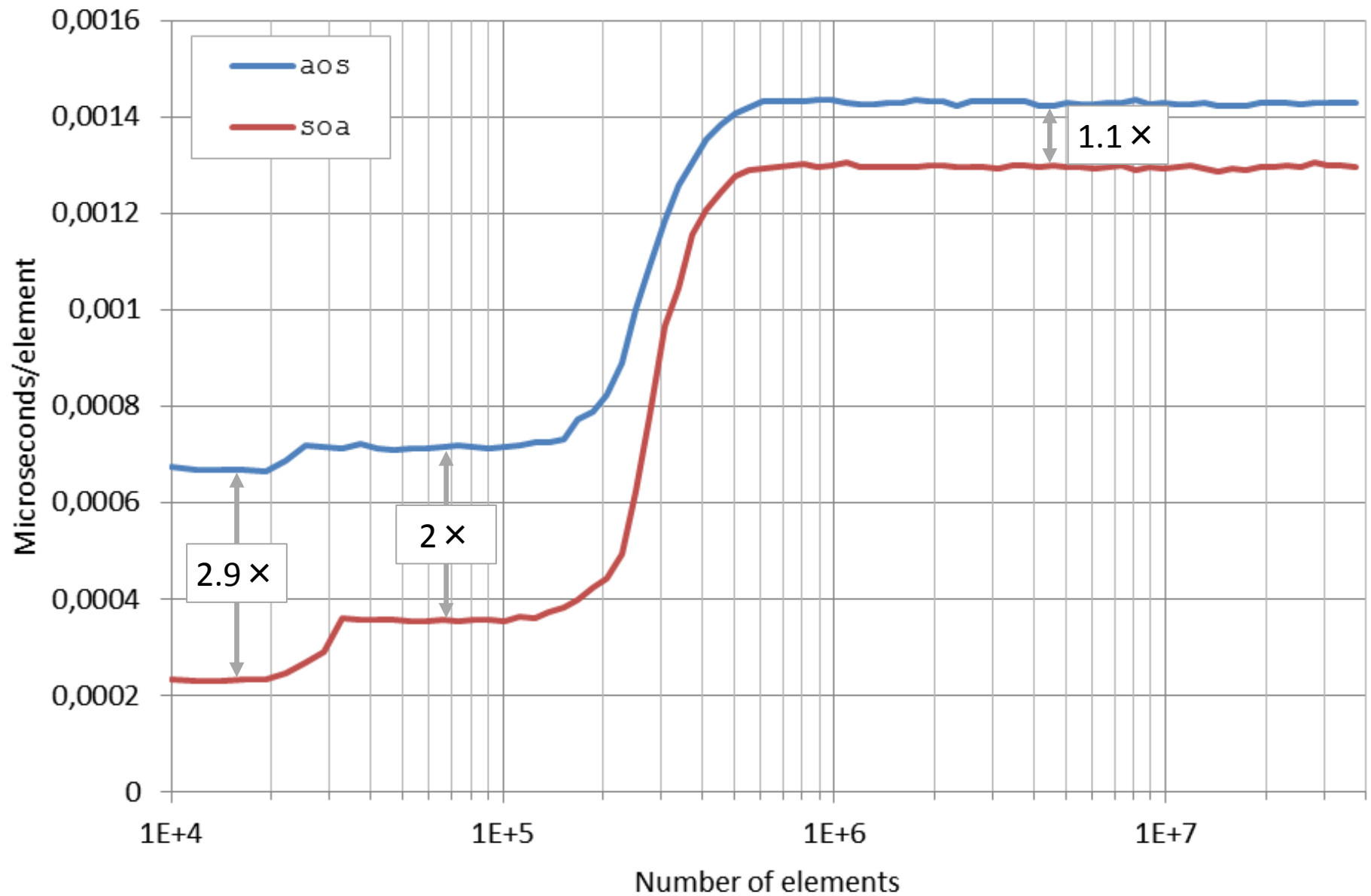
```
// ...same as before
```

```
// soa
```

```
struct particle_soa
{
    std::vector<int> x,y,z;
};
```

```
// ...same as before
```

Compact AOS vs SOA: results



Compact AOS vs SOA: analysis

- Both AOS and SOA fetch only useful values
- Yet SOA is (diminishingly as n grows) faster
- The CPU is prefetching x, y, z *in parallel*
- $L1 \leftarrow L2$ better than $L1 \leftarrow L2 \leftarrow L3$ better than $L1 \leftarrow \dots \leftarrow \text{DRAM}$

Random access in AOS vs SOA

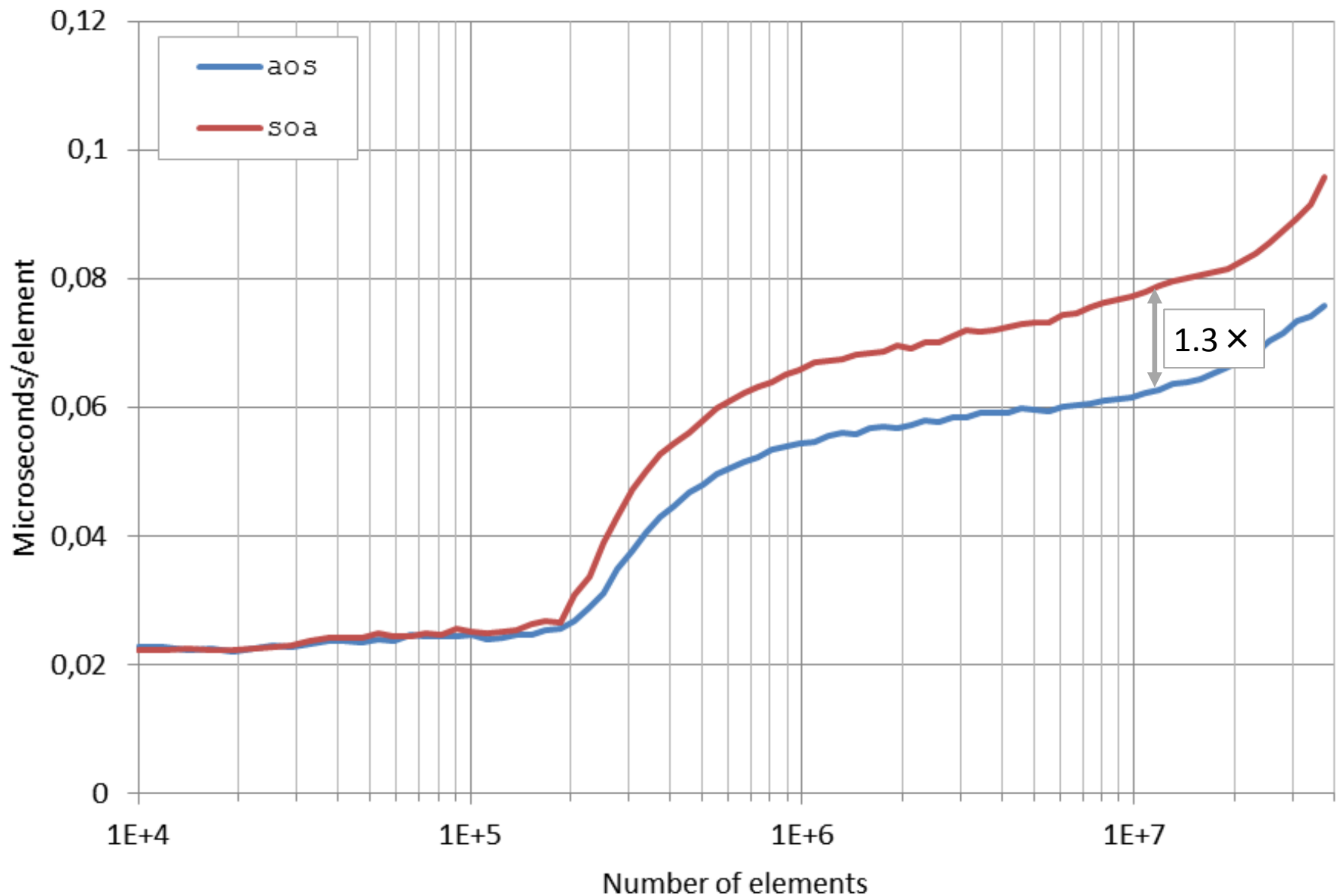
// aos

```
auto    ps=create_particle_aos(n);
long int res=0;
for(std::size_t i=0;i<n;++i){
    auto idx=rnd(gen);
    res+=ps[idx].x+ps[idx].y+ps[idx].z;
}
return res;
```

// soa

```
auto    ps=create_particle_soa(n);
long int res=0;
for(std::size_t i=0;i<n;++i){
    auto idx=rnd(gen);
    res+=ps.x[idx]+ps.y[idx]+ps.z[idx];
}
return res;
```

Random access in AOS vs SOA: results



Random access in AOS vs SOA: analysis

- You're now in a position to make an educated guess 😊
- Moral: don't trust your intuition

Parallel count

```
std::vector<int> v=...;

auto f=[&](int* px,int* py,int* pz,int* pw){
    auto th=[](int* p,int* first,int* last){
        *p=0;
        while(first!=last){
            int x=*first++;
            *p+=x%2;
        }
    };
    std::thread t1(th,px,v.data(),v.data()+n/4);
    std::thread t2(th,py,v.data()+n/4,v.data()+n/2);
    std::thread t3(th,pz,v.data()+n/2,v.data()+n*3/4);
    std::thread t4(th,pw,v.data()+n*3/4,v.data()+n);
    t1.join();t2.join();t3.join();t4.join();
    return *px+*py+*pz+*pw;
};

int res[49];

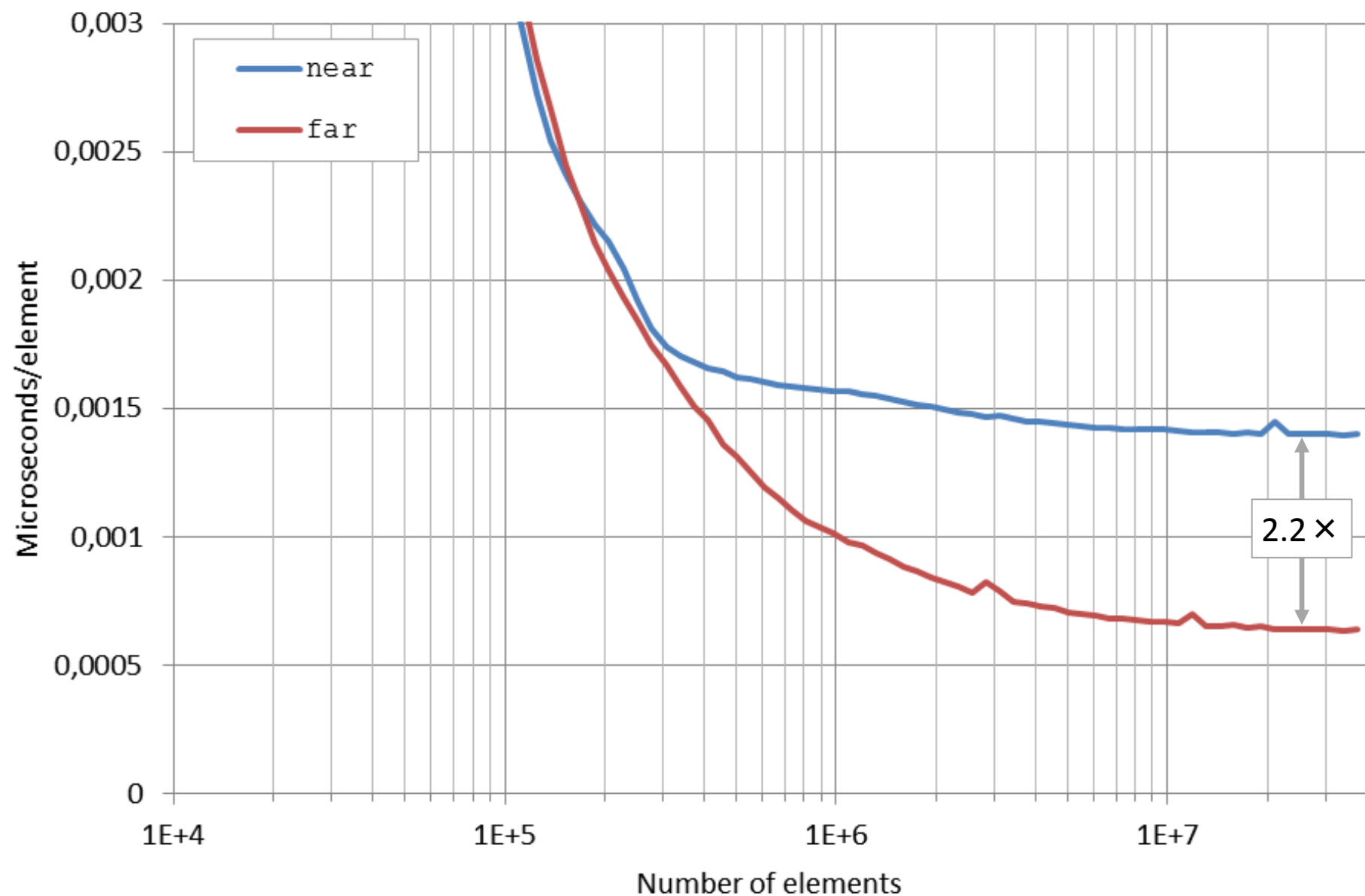
// near result addresses

return f(&res[0],&res[1],&res[2],&res[3]);

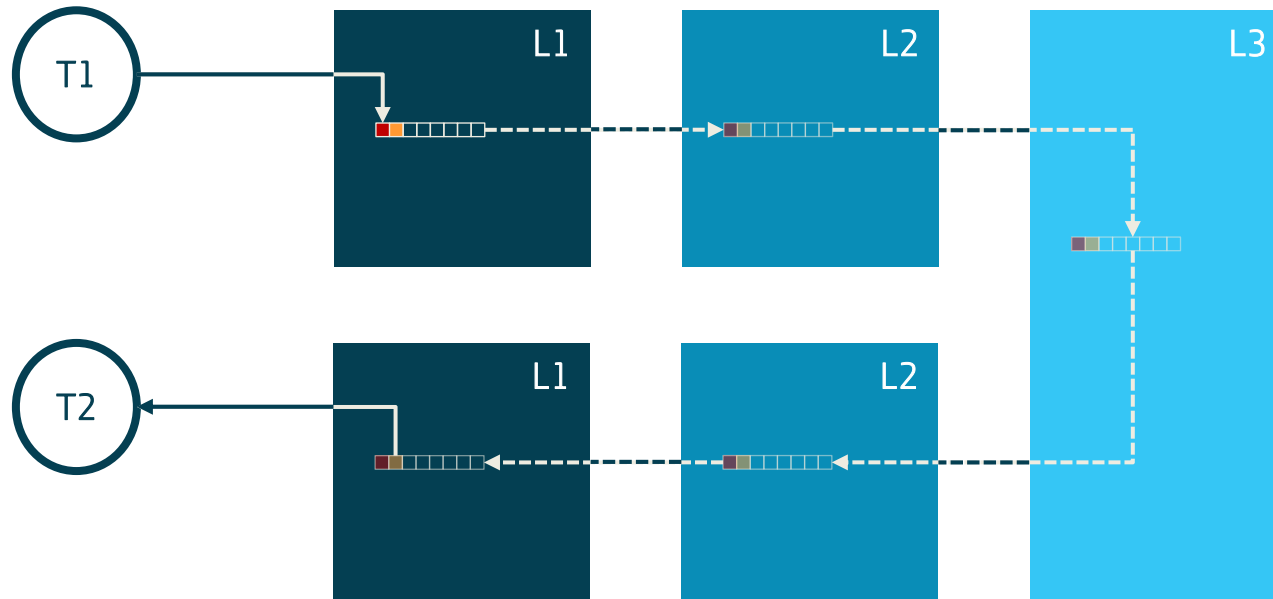
// far result addresses

return f(&res[0],&res[16],&res[32],&res[48]);
```

Parallel count: results



False sharing



- Writing to `res[0]` invalidates cached `res[1]`
- Coherence management requires full write to DRAM
- Rule of thumb: use shared write memory *only* to communicate thread results

Branch prediction



Filtered sum

```
std::vector<int> v=...;
```

```
// unsorted
```

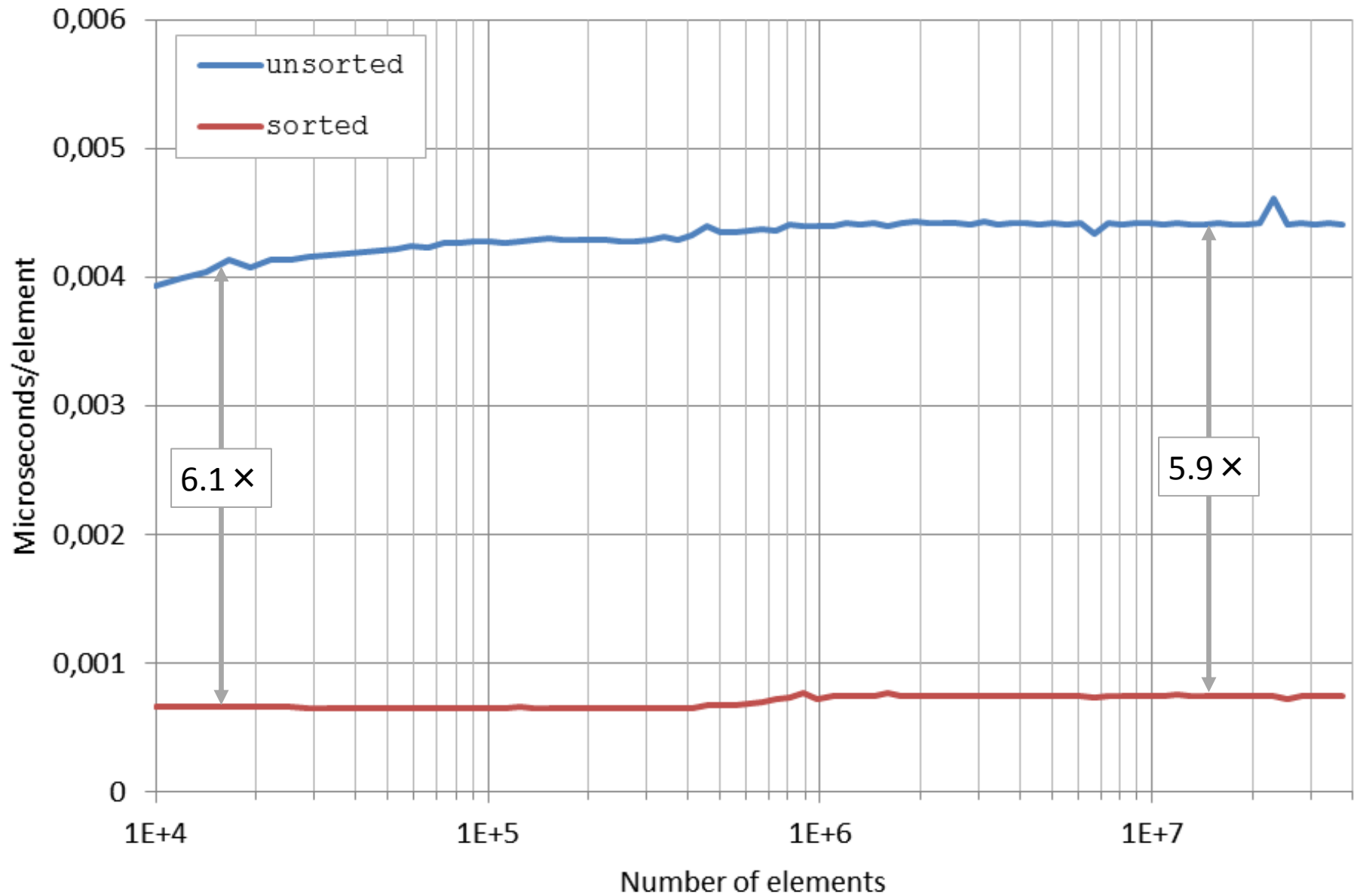
```
long int res=0;  
for(int x:v)if(x>128)res+=x;  
return res;
```

```
// sorted
```

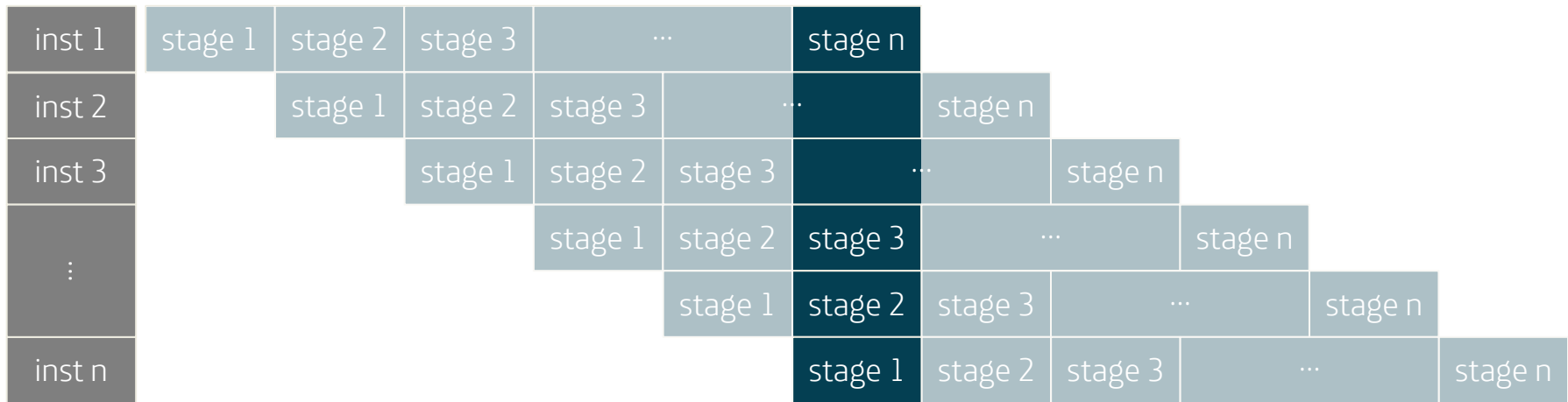
```
std::sort(v.begin(),v.end());
```

```
long int res=0;  
for(int x:v)if(x>128)res+=x;  
return res;
```

Filtered sum : results



Instruction pipeline



- Increases execution throughput by n (latency remains the same)
 - Sandy Bridge: 14-17 stages
 - Ivy Bridge: 14-19 stages
- Branch prediction failure invalidates partially executed queue

A particularly obnoxious case: bool-based processing

```
struct particle{
    bool active;
    int x,y,z;
    ...
};
std::vector<particle> particles;

void render_particles(){
    for(const particle& p:particles)if(p.active)render(p);
}
```

■ Remove active flag by smartly laying out objects

```
struct particle{
    int x,y,z;
    ...
};
std::vector<particle> particles;
std::size_t num_active;

void render_particles(){
    for(std::size_t i=0;i<num_active;++i)render(particles[i]);
}
```

■ Better cache density, no branching, fewer elements processed

Polymorphic containers

```
struct base{virtual int f()const=0; virtual ~base(){};};
struct derived1:base{virtual int f()const{return 1;}};
struct derived2:base{virtual int f()const{return 2;}};
struct derived3:base{virtual int f()const{return 3;}};
using pointer=std::shared_ptr<base>;
std::vector<pointer> v=...;
```

// unsorted

```
long int res=0;
for(const auto& p:v)res+=p->f();
return res;
```

// sorted

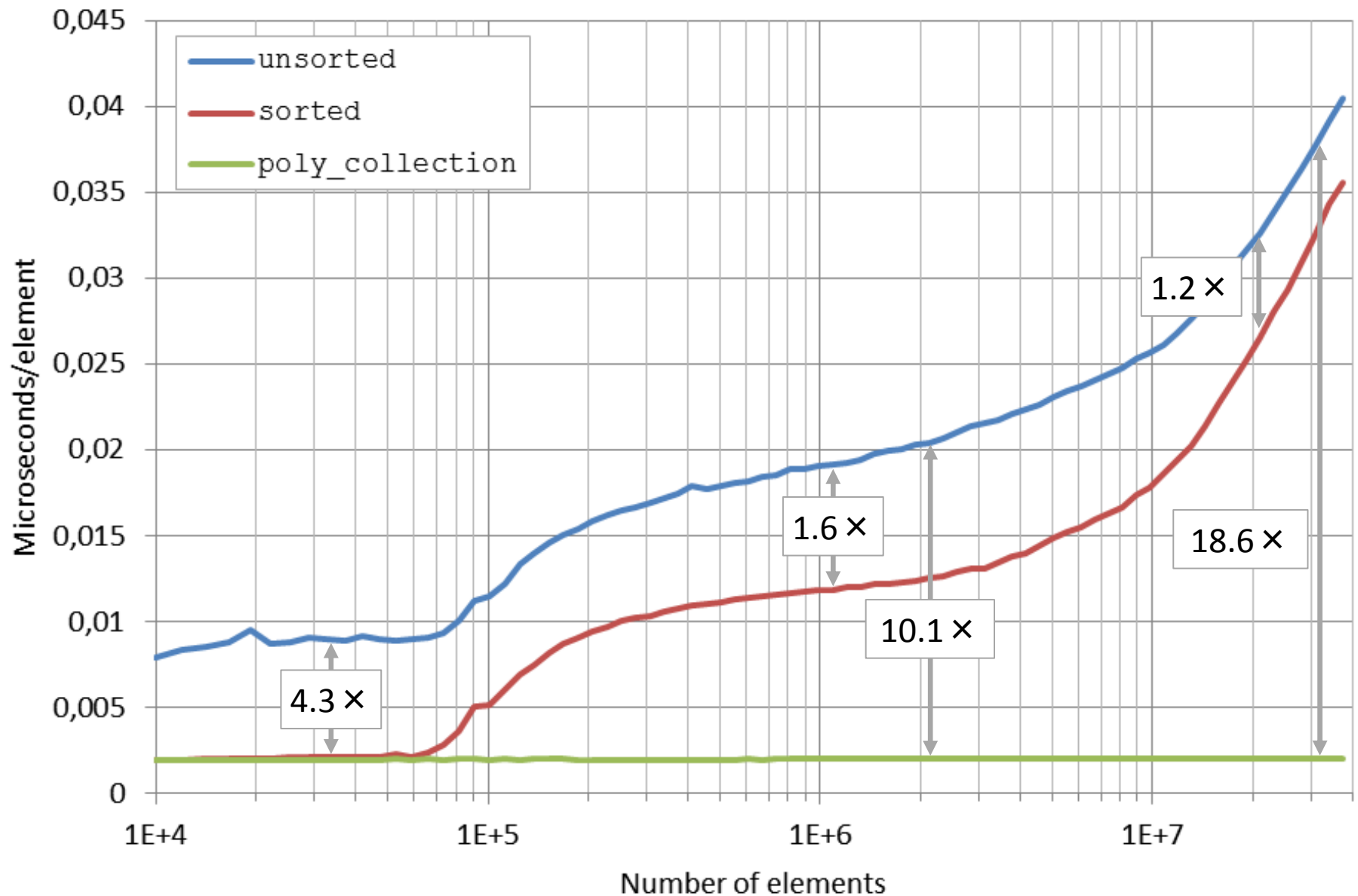
```
std::sort(v.begin(),v.end(),[](const pointer& p,const pointer& q){
    return std::type_index(typeid(*p))<std::type_index(typeid(*q));
});
```

```
long int res=0;
for(const auto& p:v)res+=p->f();
return res;
```

// poly_collection: more on this later

```
poly_collection<base> v=...;
v.for_each([&](const base& x){res+=x.f();});
```

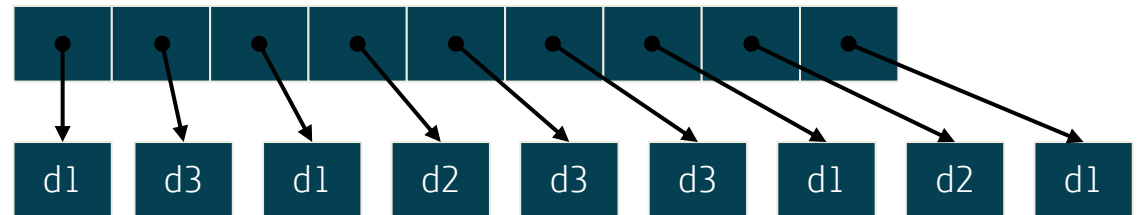
Polymorphic containers: results



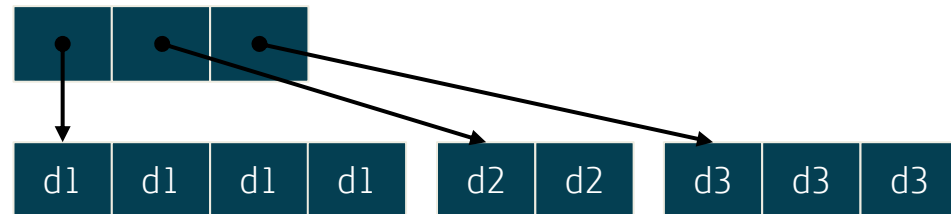
Polymorphic containers: analysis

- Sorting improves branch prediction even without data locality
- What is this `poly_collection` thing?

`vector<shared_ptr<base>>`



`poly_collection<base>`



- Branch prediction-friendly + optimum data locality
- More info on <http://tinyurl.com/poly-collection>
- Cutting-edge compilers: look for *devirtualization*

Concluding remarks



Concluding remarks

- You can't just ignore reality if you're serious about performance
- Think about the planet: don't waste energy
- Measure measure measure
- (Almost) nothing beats traversing a vector in its natural order
 - One indirection \approx one cache miss
- If multithreading, beware of false sharing
- SOA can get you huge speedups (and can not)
- Can you remove flags?
 - Look for *data-oriented design* if you want to know more
- Be regular: sort your objects

Mind the cache

Thank you

github.com/joaquintides/usingstdcpp2015

`using std::cpp 2015`

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Madrid, November 2015

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