

# Modern programming styles vs. performance

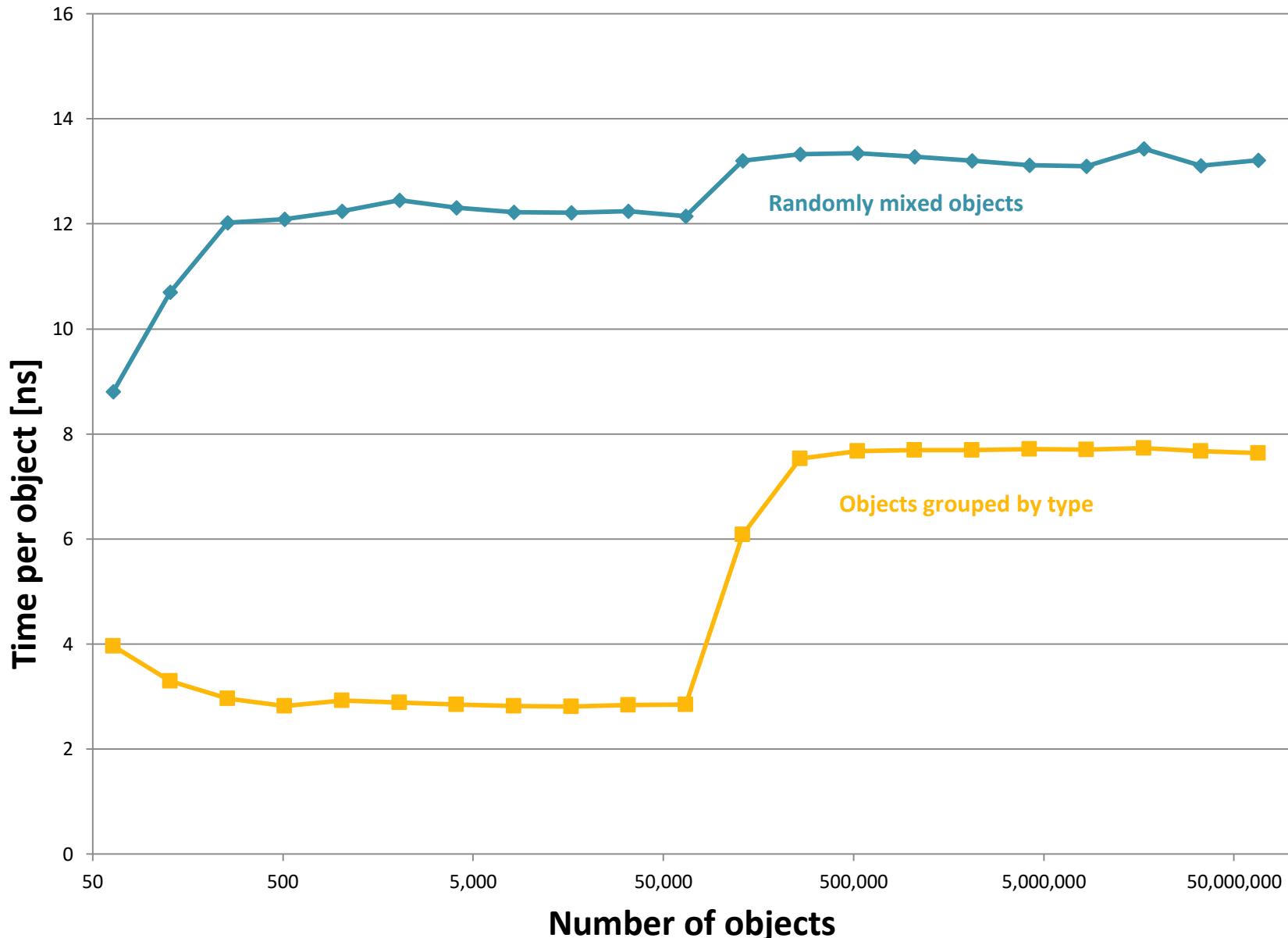
# Virtual functions

# Virtual functions - example

```
class A { public:  
    virtual int f() = 0;  
};  
  
class C3 : public A {  
    virtual int f() { return d2; }  
    int d1, d2, d3;  
};  
  
class C5 : public class Fruit {  
    virtual int f() { return d3; }  
    int d1, d2, d3, d4, d5;  
};  
  
vector< A * > data;  
  
int s = 0;  
  
for_each( data.begin(), data.end(),  
    [&]( A * p ) { s += p->f(); } );
```

- ▶ A container containing pointers to objects of several types
  - ▶ Different implementations of f()
  - ▶ Sum the values returned from f()
  
- ▶ Experiment setup
  - ▶ 5 concrete classes, varied size
  - ▶ f() returns a data field at different positions
  
- ▶ A container filled with random mixture

# Results (Intel Core Microarchitecture, 2.66 GHz)

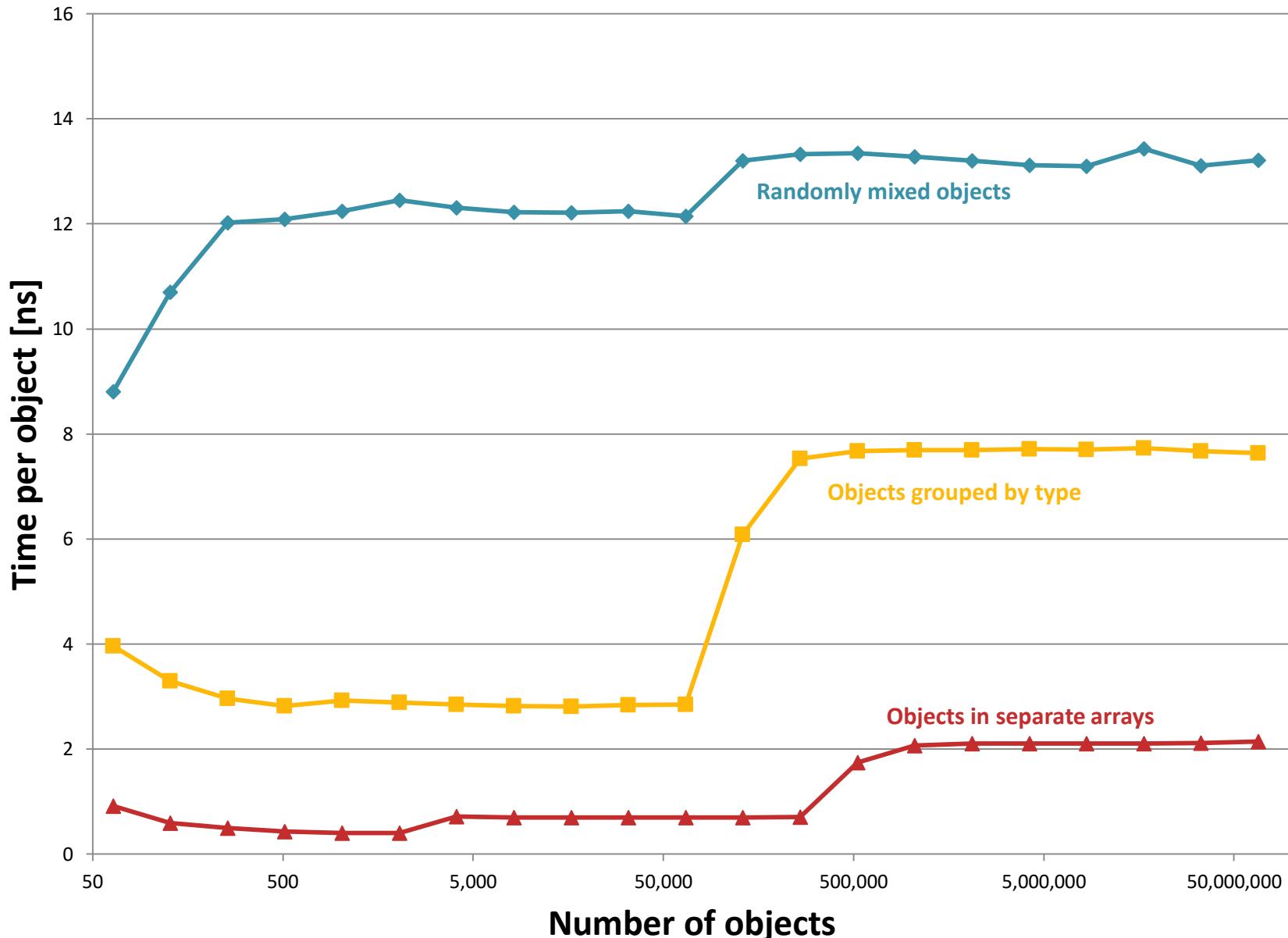


# Non-object implementation - example

```
class C3 { public:  
    int f() { return d2; }  
    int d1, d2, d3;  
};  
  
class C5 { public:  
    int f() { return d3; }  
    int d1, d2, d3, d4, d5;  
};  
  
vector< C3> data3;  
vector< C5> data5;  
  
int s = 0;  
  
for_each( data3.begin(), data3.end(),  
    [&]( C3 & p) { s += p.f(); });  
  
for_each( data5.begin(), data5.end(),  
    [&]( C5 & p) { s += p.f(); });
```

- ▶ No base class, no virtual functions
- ▶ No pointers required
- ▶ Several containers required
  - ▶ Difficult to extend with new object types
- ▶ std::tuple<std::vector<TLIST>...>
- ▶ Does not preserve the order of objects
  - ▶ If required, an additional array of indexes is needed
  - ▶ For non-commutative aggregation, any advantage is lost

# Results (Intel Core Microarchitecture, 2.66 GHz)

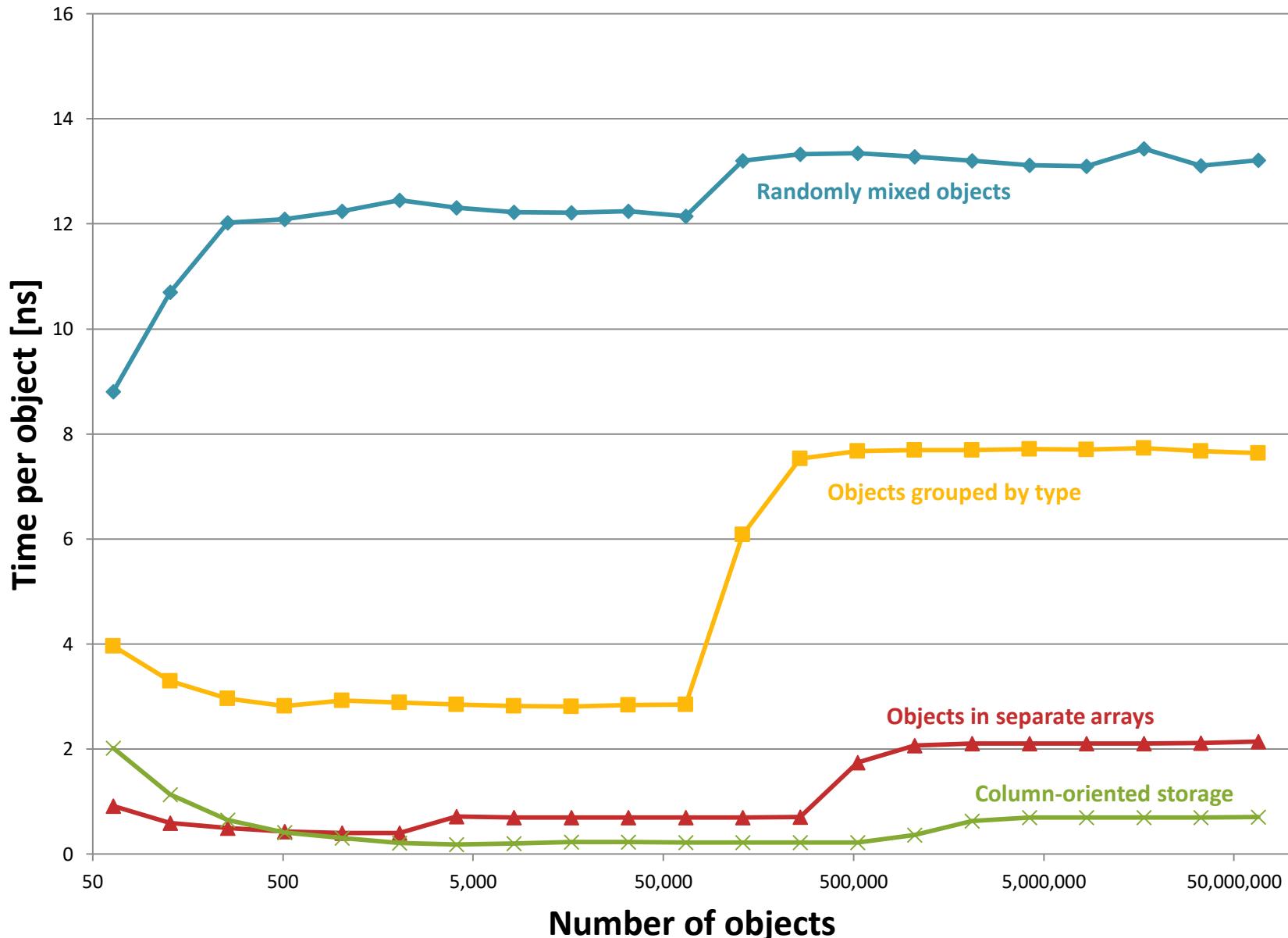


# Column-oriented implementation - example

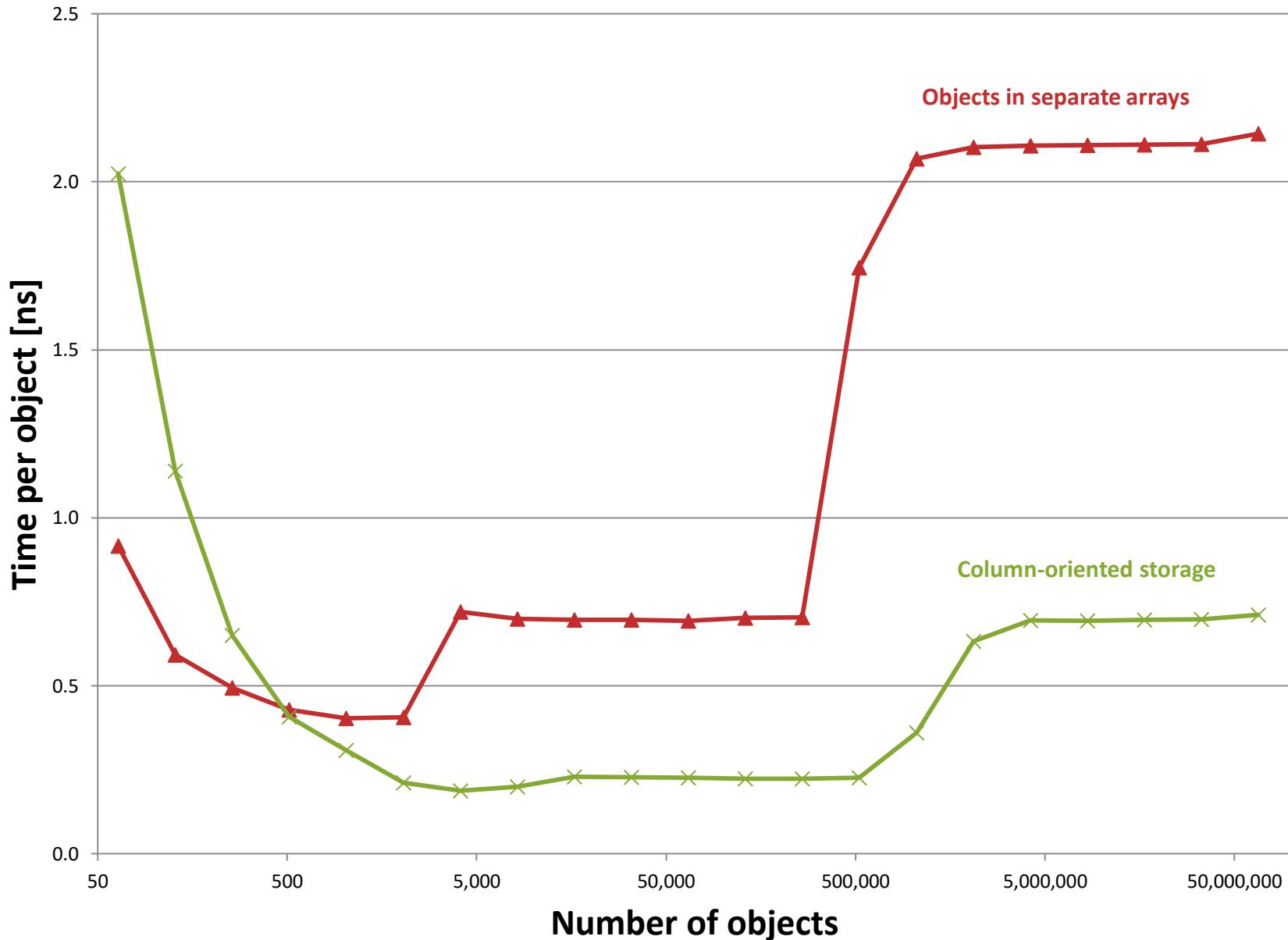
```
vector< int> data3d1;  
  
vector< int> data3d2;  
  
vector< int> data3d3;  
  
vector< int> data5d1;  
  
vector< int> data5d2;  
  
  
vector< int> data5d3;  
  
vector< int> data5d4;  
  
vector< int> data5d5;  
  
  
int s = 0;  
  
s += std::reduce( data3d2.begin(),  
                 data3d2.end(), 0, std::plus<int>());  
  
s += std::reduce( data5d3.begin(),  
                 data5d3.end(), 0, std::plus<int>());
```

- ▶ No classes at all
- ▶ Just arrays of elementary types
  - ▶ Extremely difficult and error-prone
  - ▶ No support for generic programming in this style
  
- ▶ For each column, data are dense
  - ▶ Improves cache behavior for column-oriented access
  - ▶ Degrades cache behavior for row-oriented access
  - ▶ Allows vectorization
    - std::reduce [C++17] allows compilers to vectorize automatically
    - The experiments were done with manual vectorization

# Results (Intel Core Microarchitecture, 2.66 GHz)



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- ▶ Grouping the data by type
  - ▶ No change in code
  - ▶ More than **4**-fold speed-up (with 10000 objects)
- ▶ C-style programming
  - ▶ Separate arrays of structures
  - ▶ Additional **4**-fold speed-up
- ▶ FORTRAN-style programming
  - ▶ Arrays of elementary types
  - ▶ Vectorization enabled
  - ▶ Additional **3**-fold speed-up
- ▶ FORTRAN vs. Object-oriented programming: **50 : 1**
  - ▶ Fast but breaking all software-engineering rules
- ▶ Warning: This is a well-designed example
  - ▶ Very simple operations, order not preserved, ...