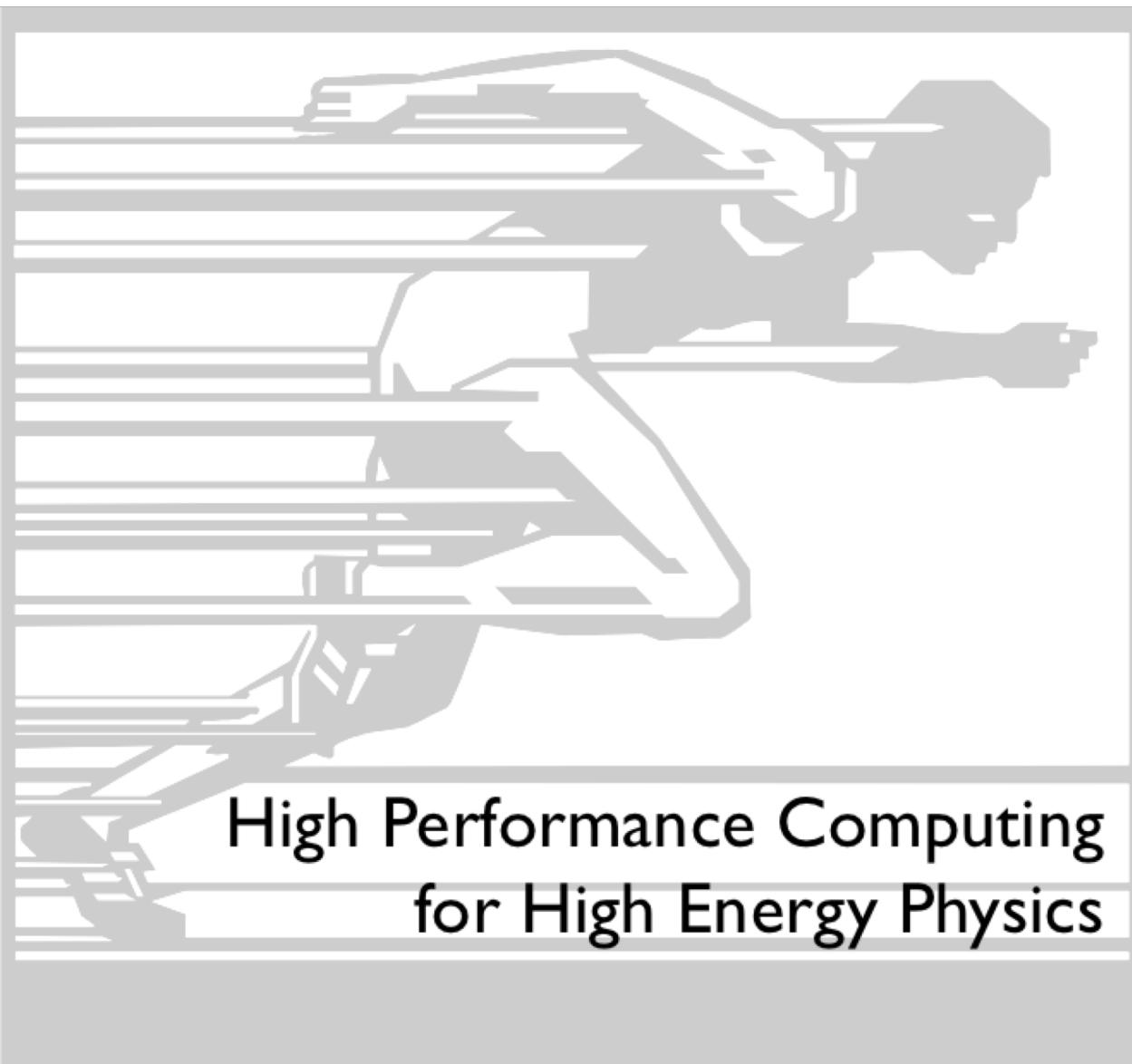


# A SOA Event Model

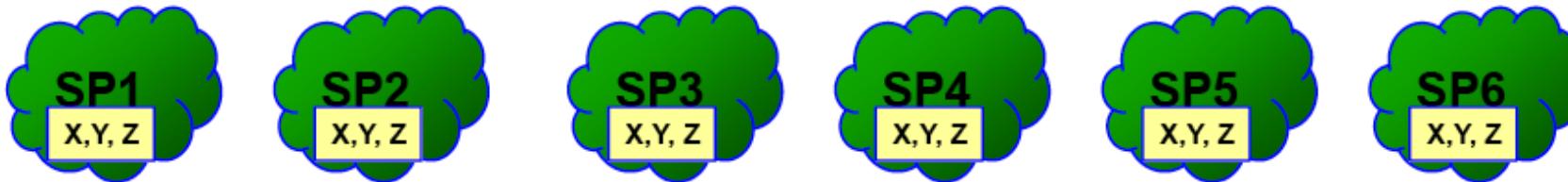


High Performance Computing  
for High Energy Physics

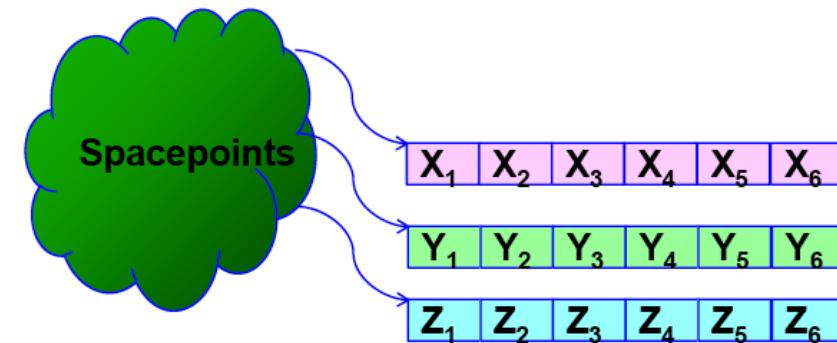
Vincenzo Innocente  
CERN/EP/SFT  
CMS Experiment

# Data Organization: AoS vs SoA

- Traditional Object organization is an Array of Structures
  - Abstraction often used to hide implementation details at object level

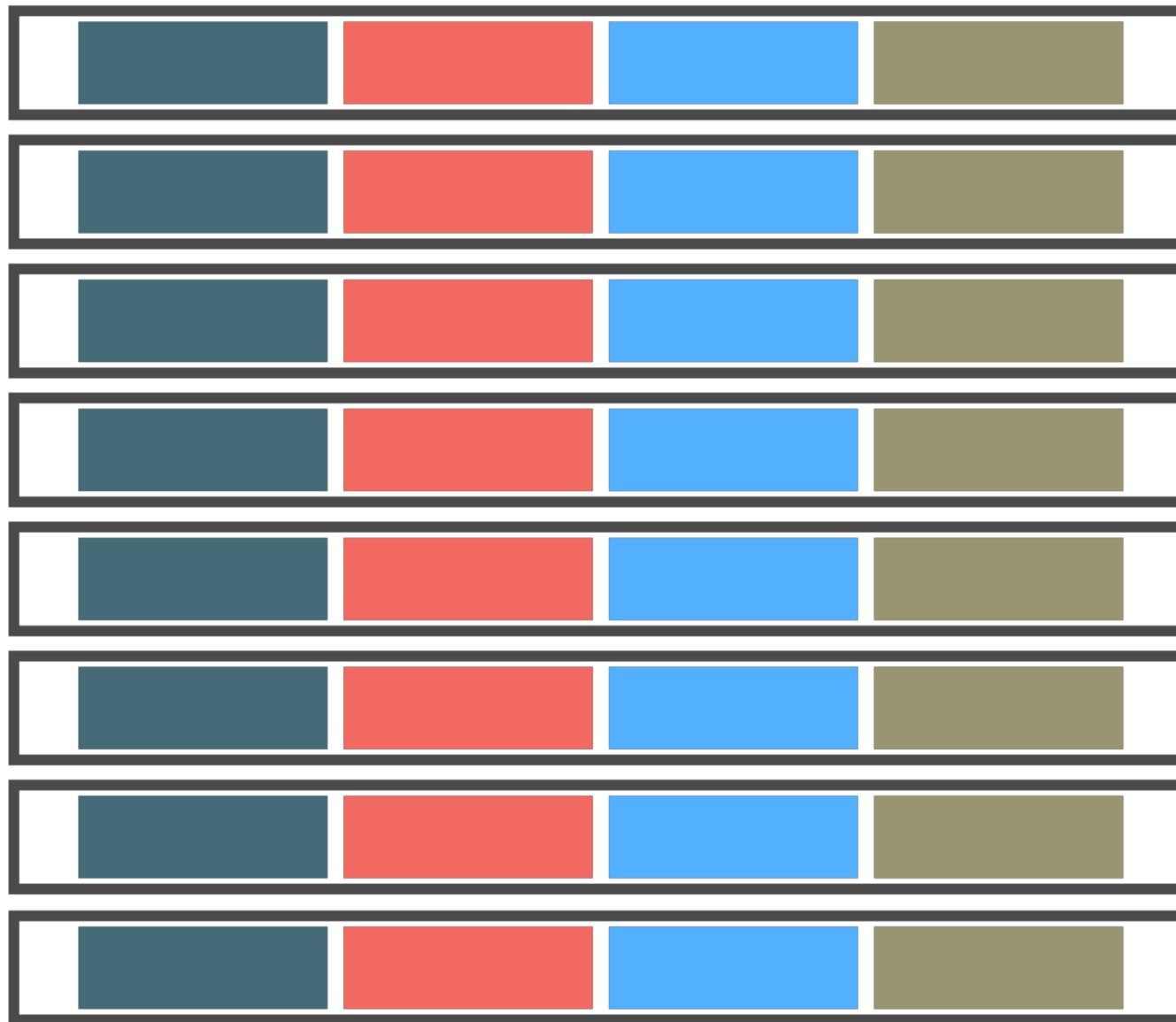


- Difficult to fit stream computing
- Better to use a Structure of Arrays
  - (column-wise storage)
- OO can wrap SoA as the AoS
  - Move abstraction higher
  - Expose data layout to the compiler
- Explicit copy in many cases more efficient
  - (notebooks vs whiteboard)



# Array of Structures

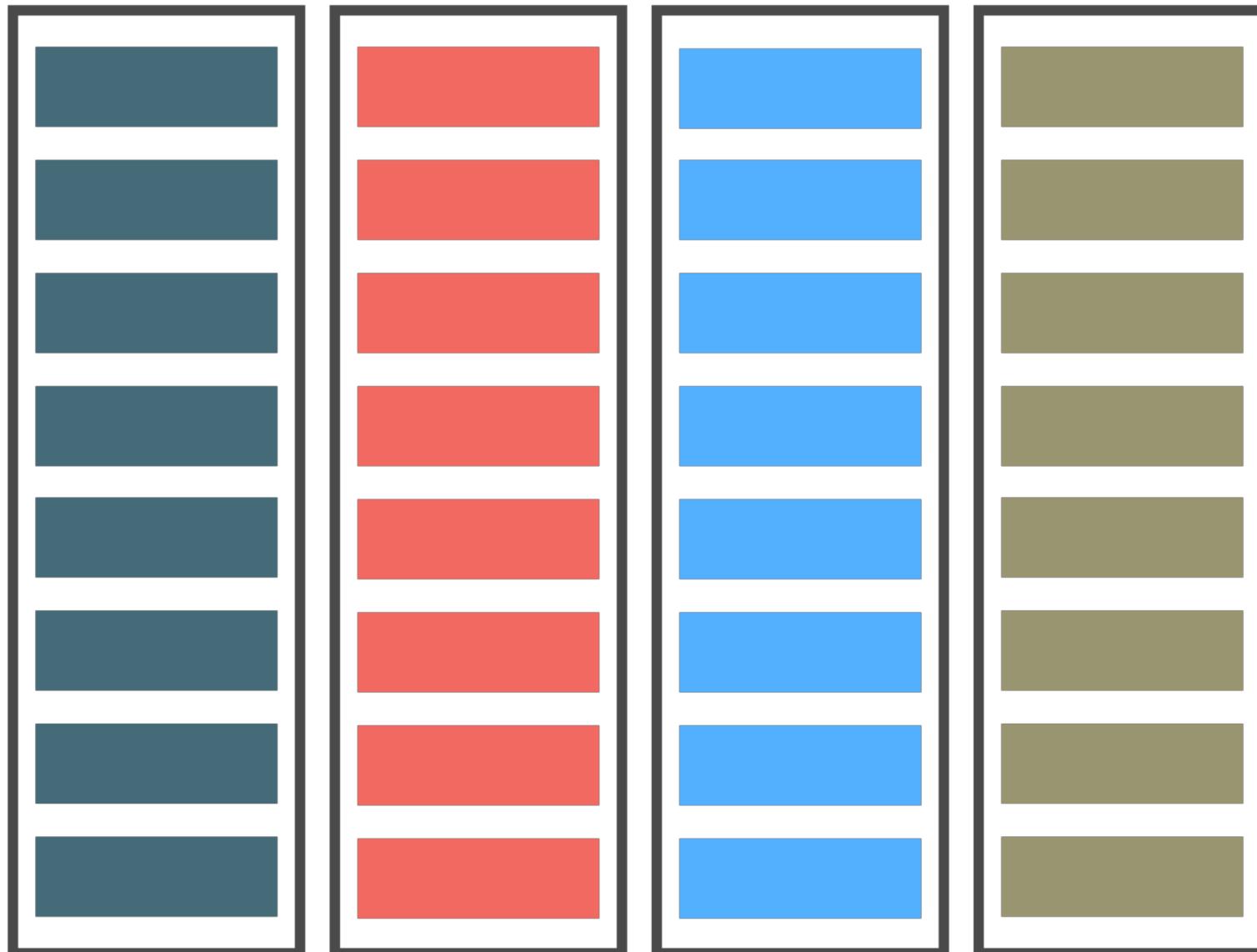
T matrix[N][4];



std::vector< >

# Structure of Array

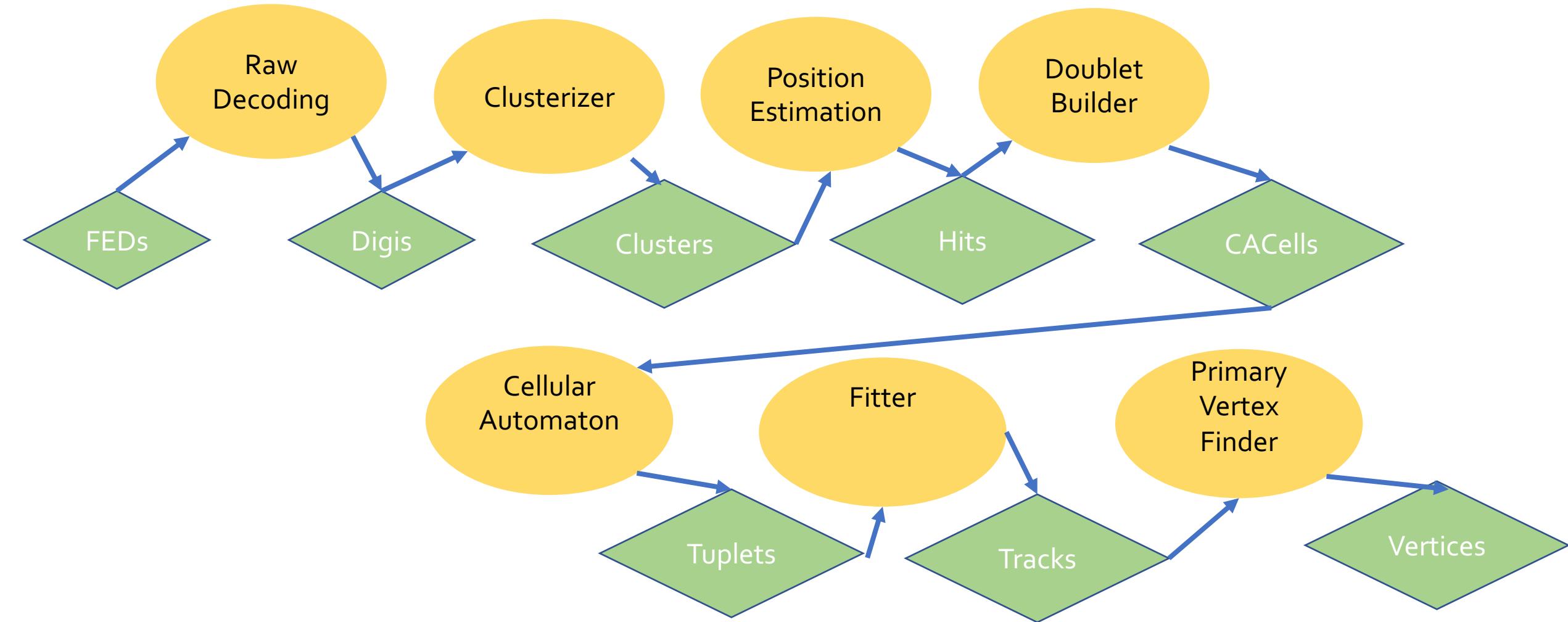
## T matrix[4][N];



# Why SOA?

- SOA naturally fits SIMD/SIMT architectures (stream computing)
- No padding or alignment issue: one can choose optimized types
- Its cache friendly in particular if only few elements are accessed
- Very Large SOA (in both dimensions) may affect TLB though

# Pixel Tracks/Seed Workflow



# Digis SOA

$x_{21}$	...	$x_{41}$ $x_{42}$ $x_{44}$ $x_{45}$ $x_{46}$ $x_{47}$ $x_{48}$ $x_{49}$ $x_{50}$ $x_{51}$ $x_{52}$ $x_{53}$	...	x (row)
$y_{21}$	...	$y_{41}$ $y_{42}$ $y_{44}$ $y_{45}$ $y_{46}$ $y_{47}$ $y_{48}$ $y_{49}$ $y_{50}$ $y_{51}$ $y_{52}$ $y_{53}$	...	y (column)
$c_{21}$	...	$c_{41}$ $c_{42}$ $c_{44}$ $c_{45}$ $c_{46}$ $c_{47}$ $c_{48}$ $c_{49}$ $c_{50}$ $c_{51}$ $x_{52}$ $c_{53}$	...	adc (charge)
42	...	42 42 42 42 42 42 42 42 42 42 42 42	...	moduleId
2	...	2 1 3 2 2 1 3 1 1 1 1 2	...	clusterId

The result of the clusterizer is just the assignment of a clusterId to each digi (and the number of clusters per module as necessary side-effect)

# Few Parallel Axioms on filling data structures

- Never Resize:
  - either size is known in advance
  - or fixed to some Max
  - or: first count, then fill
- Never delete: prefer masking
- Prefer *many-to-one* to *one-to-many* references
- Avoid sorting: is highly parallel unfriendly
  - Try to get it right from start!
- Atomic increment at SIMD/SIMT level is cheap:
  - Once per vector/warp (cuda: implemented in compiler, X86: do it yourself)

# Example of use of SOA

```
// compute cluster charge
__global__
void clusterCharge(int16_t const * clusterid, uint8_t const * adc,
                   int * charge, int n) {
    int first = blockDim.x * blockIdx.x + threadIdx.x;
    for (int i = first; i < n; i += gridDim.x*blockDim.x)
        if (clusterId(i) >= 0)
            atomicAdd(&charge[clusterId(i)], adc[i]);
}

int threadsPerBlock = 256; // a magic number
int blocksPerGrid =
    (maxNumOfDigis+ threadsPerBlock - 1) / threadsPerBlock;
clusterCharge<<<blocksPerGrid, threadsPerBlock>>>(clId,adc, charge,nDigis);
```

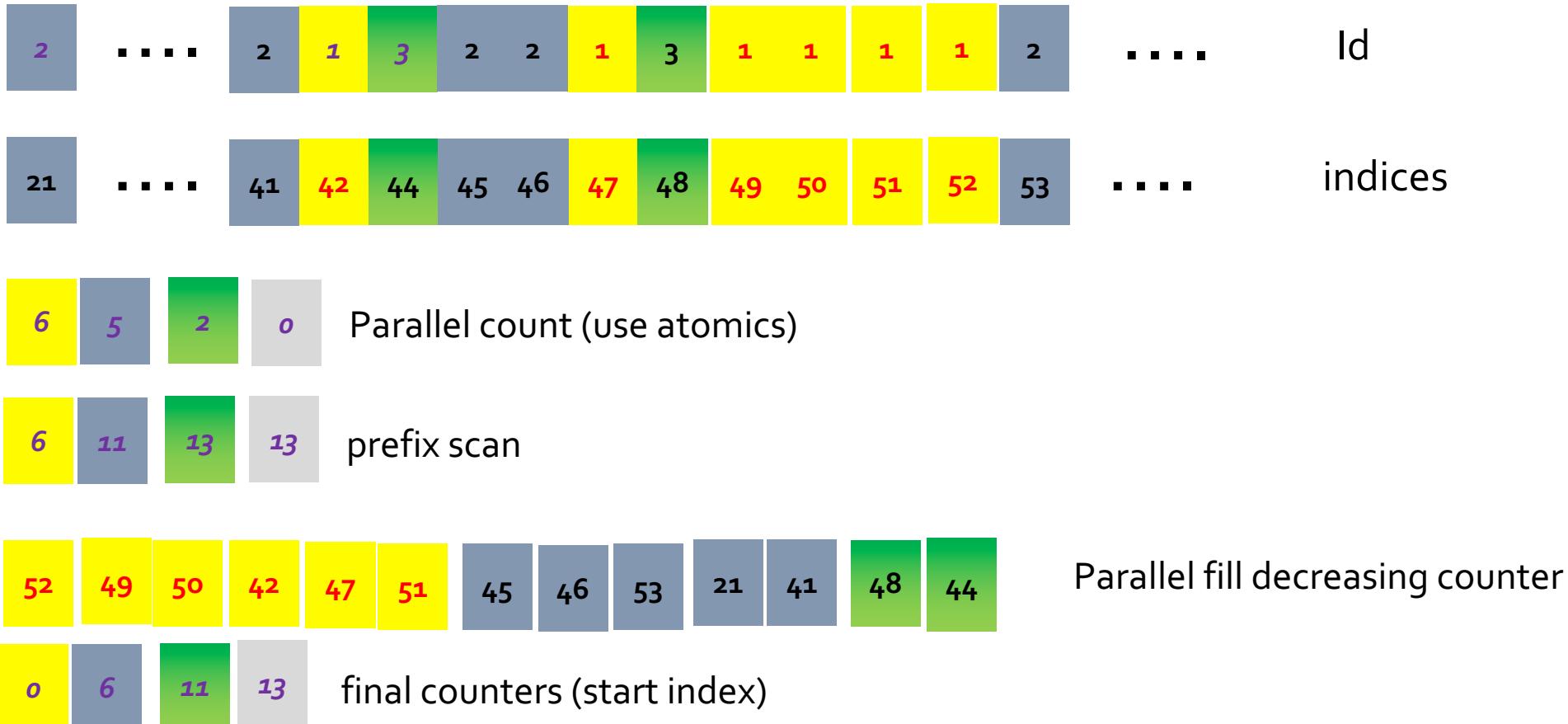
“loop” on digis

Accumulate in Cluster SOA

# Position (and Error) estimation

- Loop on Digits (not cluster) (blocked by Module as clusterizer)
  - Loop in CUDA means one per thread...
- Fills standard object per cluster in shared memory with “CPE” info
  - Fully concurrent, use atomics
- Synchronize!
- Loop on “clusters” (CPE objects) and fill hit-SOA
  - local  $x, y, x_{err}, y_{err}$
  - global  $x_g, y_g, z_g, r_g, \phi$  ( $\phi$  is an int16)
    - Beamspot subtracted!

# Filling One to Many Association (aka vector of vector)



# AoS, SoA, Eigen...

## AoS

```
struct Point {  
    float x,y,z;  
};  
std::vector<Point> points;
```

## EigenAoS

```
Map<Vector3f> point(p+3*i);
```

## SoA

```
struct Points {  
    std::vector<float> x,y,z;  
};
```

```
struct Points {  
    float x[N],y[N],z[N];  
    uint32_t size;  
};
```

## Unified-View

```
struct Points {  
    float * p;  
    uint32_t size;  
};
```

## SoA-View

```
struct Points {  
    float * x,y,z;  
    uint32_t size;  
};
```

## EigenSoA

```
Map<Vector3f,al, InnerStride<maxPoints>> point(p+i);
```

# More complex

```
struct Hit {  
    Point p; // x,y,z  
    CovXYZ err; // 6 elements  
};  
  
struct Track {  
    Hit hits[4];  
};  
  
std::vector<Track> tracks;
```



```
constexpr uint32_t maxNumberOfTracks() { return 10*1024; }  
constexpr uint32_t stride() { return maxNumberOfTracks();}  
  
using Matrix3x4f = Eigen::Matrix<float,3,4>; // each hit in a column  
using Map3x4f = Eigen::Map<Matrix3x4f,0,Eigen::Stride<3*stride(),stride()>>;  
  
using Matrix6x4f = Eigen::Matrix<float,6,4>;  
using Map6x4f = Eigen::Map<Matrix6x4f,0,Eigen::Stride<6*stride(),stride()>>;
```

# More general

```
using V3 = Eigen::Vector3f;
using V15 = Eigen::Matrix<float,15,1>;
struct TSOS {
    V3 position;
    V3 momentum;
    V15 covariance;
};
```

```
template<typename M, int S>
struct alignas(128) MatrixSoA {
    using Scalar = typename M::Scalar;
    using Map = Eigen::Map<M, 0,
                           Eigen::Stride<M::RowsAtCompileTime*S,S>>;
    using CMap = Eigen::Map<const M, 0,
                           Eigen::Stride<M::RowsAtCompileTime*S,S>>;
    constexpr Map operator()(uint32_t i) { return Map(data+i);}
    constexpr CMap operator()(uint32_t i) const { return CMap(data+i);}

    Scalar data[S*M::RowsAtCompileTime*M::ColsAtCompileTime];
    static_assert(isPowerOf2(S),"stride not a power of 2");
    static_assert(sizeof(data)%128==0, "size not a multiple of 128");
};
```



```
template<int S>
struct TSOSSoa {
    static constexpr int stride() { return S;}
    MatrixSoA<V3,S> position;
    MatrixSoA<V3,S> momentum;
    MatrixSoA<V15,S> covariance;
};
```

# Dynamic??

```
using DynStride = Eigen::Stride<Eigen::Dynamic,Eigen::Dynamic>;
template<typename M>
struct MatrixDynSoA {
    using Scalar = typename M::Scalar;
    using Map = Eigen::Map<M, 0, DynStride>;
    using CMap = Eigen::Map<const M, 0, DynStride>;
    constexp auto eStride() const { return DynStride(M::RowsAtCompileTime*stride,stride);}

    constexpr Map operator()(uint32_t i) { return Map(data+i, estride());}
    constexpr CMap operator()(uint32_t i) const { return CMap(data+i, eStride());}

    Scalar * data;
    int stride;
};
```

```
using V3 = Eigen::Vector3f;
using V15 = Eigen::Matrix<float,15,1>;
struct TSOS {
    V3 position;
    V3 momentum;
    V15 covariance;
};
```



```
template<int S>
struct TSOSSoa {
    MatrixDynSoA<V3> position;
    MatrixDynSoA<V3> momentum;
    MatrixDynSoA<V15> covariance;
    TSOSSoa(void *, int);//?????
};
```

# “Managing” Dynamic SOAs

<https://github.com/cms-patatrack/cmssw/commit/8eaa29bd1066458obec5d0229da549c4cffad64d>

# What about my cute OO design?

```
struct Box {  
    Eigen::AffineCompact3f transform;  
    V3 halfWidth;  
  
    template<typename P3>  
    inline bool inside(P3 const & p) const {  
        return ((transform*p).array().abs() < halfWidth.array()).all();  
    }  
};
```

<https://godbolt.org/z/N-1D22>

```
void doAOS(std::vector<TSOS> &trajs, Box const & b,  
           std::vector<bool> & res) {  
    std::transform(trajs.begin(),trajs.end(),res.begin(),  
                 [&](auto const& t){ return b.inside(t.position);});  
}
```

```
void doSOA(TSOSSoa & trajSoa, Box const & b,  
           bool * res) {  
    #pragma GCC ivdep  
    for (auto i=0U; i<nTracks; ++i)  
        res[i] = b.inside(trajSoa.position(i));  
}
```

# More Flexible: ASoA (“bucketized SoA”)

```
template<uint32_t S>
struct alignas(128) SoA {
    static constexpr uint32_t stride() { return S; }
    static constexpr uint32_t mask() { return S-1; }
    static constexpr uint32_t shift() { return ilog2(S); }

    float a[S];
    float b[S];

    static_assert(isPowerOf2(S),
                  "stride not a power of 2");
    static_assert(sizeof(a)%128 == 0,
                  "size not a multiple of 128");
};

template< uint32_t S>
using ASoA = std::vector<SoA<S>>;
// std::vector<SoA<S>*>; std::list<SoA<S>>;?????
```

```
using V = SoA<myStride>;
void sum(V * psoa, uint32_t n) {
    for (uint32_t i=0; i<n; i++) {
        auto j = i/V::stride(); // i>>V::shift();
        auto k = i%V::stride(); // i&V::mask();
        auto & soa = psoa[j];
        soa.b[k] += soa.a[k];
    }
}
```

```
void sum(V * psoa , uint32_t n) {
    auto nb = (n+V::stride()-1)/ V::stride();
    for (uint32_t j=0; j<nb; j++) {
        auto & soa = psoa[j];
        auto kmax = std::min(V::stride(),n - j*V::stride());
        for(uint32_t k=0; k<kmax; k++) soa.b[k] += soa.a[k];
    }
}
```

# More Flexible: ASoA (“bucketized SoA”) cuda

```
template<uint32_t S>
struct alignas(128) SoA {

    static constexpr uint32_t stride() { return S; }
    static constexpr uint32_t mask() { return S-1; }
    static constexpr uint32_t shift() { return ilog2(S); }

    float a[S];
    float b[S];

    static_assert(isPowerOf2(S),
                  "stride not a power of 2");
    static_assert(sizeof(a)%128 == 0,
                  "size not a multiple of 128");
};


```

```
__global__
void sum(V * psoa, int n) {
    auto first = threadIdx.x + blockIdx.x*blockDim.x;
    for (auto i=first; i<n; i+=blockDim.x*gridDim.x) {
        auto j = i/V::stride();
        auto k = i%V::stride();
        auto & soa = psoa[j];
        soa.b[k] += soa.a[k];
    }
} __global__ // maps buckets to blocks
void sum(V * psoa, int n) {
    auto nb = (n+V::stride()-1)/V::stride();
    for (auto j=blockIdx.x; j<nb; j+=gridDim.x) {
        auto & soa = psoa[j];
        auto kmax = std::min(V::stride(),n - j*V::stride());
        for(uint32_t k=threadIdx.x; k<kmax; k+=blockDim.x) {
            soa.b[k] += soa.a[k];
        }
    }
}
```

# ASoA: a possible implementation

<https://github.com/VinInn/cmssw/blob/EigenSOA/HeterogeneousCore/CUDAUtilities/interface/ASoA.h>

<https://github.com/VinInn/cmssw/blob/EigenSOA/DataFormats/SiPixelDigi/interface/SiPixelDigisSoA.h>

# Advantaged of fixed size SoA

- Easy to manage and compose
  - Only one (de-)allocation and eventual copy for the whole SoA
- Correct alignment and sizing imposed/checked at compile time
- Compiler can take full advantage of knowing size and alignment
- Runtime sizing can be achieved using Arrays of SoAs
  - Bucketized SoA can help to mitigate pressure on TLB
  - Difficult to use with external utilities (memset!)