

Parallelization and vectorization of the Fit

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ROOT

Data Analysis Framework

<https://root.cern>

My poster

Parallelization and Vectorization of ROOT Fitting classes



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Introduction

In order to take full advantage of new computer architectures and to satisfy the requirement of maximizing the CPU usage with increasing amount of data, analysis parallelization and vectorization have been introduced in the ROOT mathematical and statistical libraries, requiring minimal changes in user code.



CERN Computing News Group Storage (CNGS)

As part of this effort, new generic classes supporting a task based parallelization mode have been defined in ROOT, which can be used for a wide range of computational tasks in the field of High Energy Physics. The support for different SIMD libraries has also been included.

Tools for parallelism

Task level parallelism: TThreadExecutor

- Task-oriented, multithreaded MapReduce for ROOT
- Provides operations Map, Reduce, Foreach and even chunked mapping with partial reduction.
- Used in main fitting: MINOS/Binned-Decoder Tree evaluation, Deep Neural Networks processing, implicit multithreading operators in I/O handling, deserialization and decompression of tree branches in parallel, parallel webbing and for parallel execution of functional classes in TGlobalFrame.

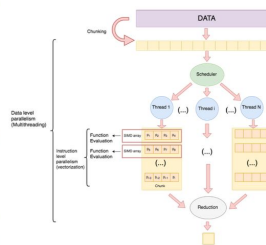
```
#include "TThreadExecutor.h"
using namespace ROOT;

// Example of task-oriented multithreaded MapReduce
// for fitting a Gaussian function
int main() {
    TThreadExecutor te(4);
    TThreadExecutor::MapReduce(te, ...);
    return 0;
}
```

Instruction level parallelism: VecCore

- Provides efficient vectorization on all platforms by writing abstract, architecture generic code that will map to each of the optimal hardware: SSE, AVX or FMA instructions. Includes a wrap library for the case when SIMD operations are not available.
- Use the faster "speeding-up software with VecCore, a portable SIMD library" by Guilherme Almeida

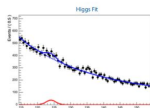
Fitting Parallelization



References

1. VecCore Library <https://github.com/root-project/veccore>
2. Wt Library <https://github.com/WtDev/wt>
3. LMB-SIMD <https://github.com/leleu/SIMD>
4. ROOT Data Analysis Framework <https://root.cern>

CASE EXAMPLE: HIGGS FIT



```
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```

Current implementation

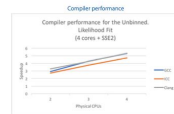
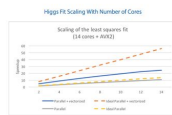
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Vectorized plus parallelized implementation

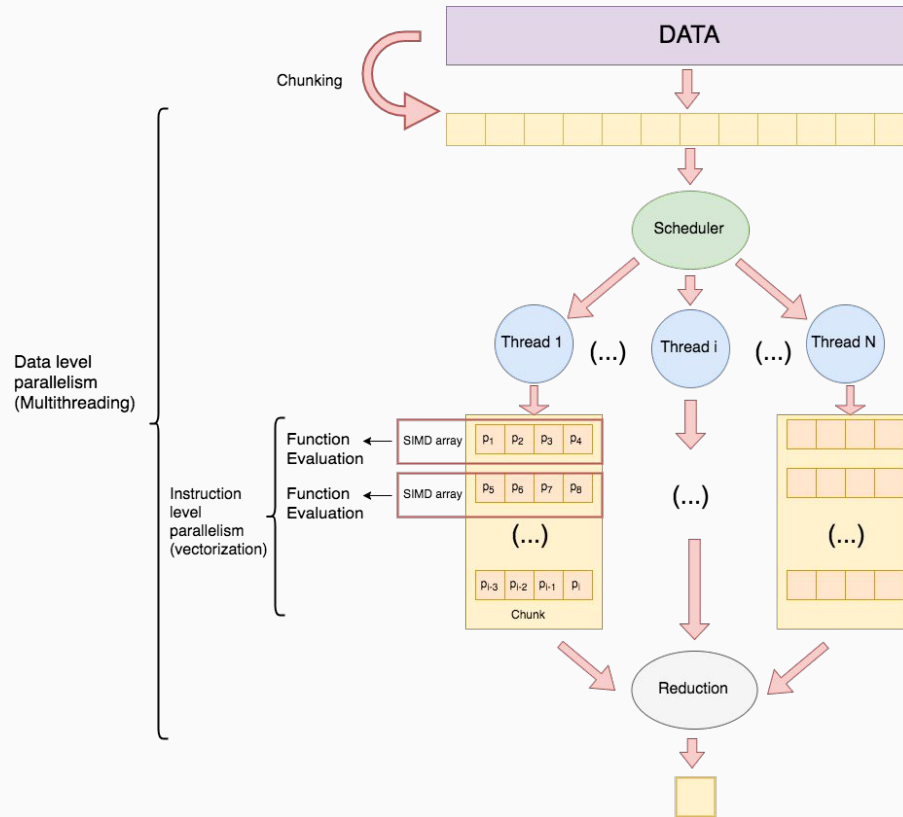
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// Example of task-oriented multithreaded MapReduce
// for fitting a Gaussian function
int main() {
    TThreadExecutor te(4);
    TThreadExecutor::MapReduce(te, ...);
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Fitting parallelization





Tools for parallelism

Task level: **ROOT::TThreadExecutor**

Instruction level: **VecCore**



Backwards compatibility!

Spot the differences

```
//Higgs Fit: Implementation of the scalar function
double func(const double *data, const double *params)
{
    return params[0] * exp(-(*data + (-130.)) * (*data + (-130.)) / 2) +
        params[1] * exp(-(params[2] * (*data * (0.01)) - params[3] *
            ((*data) * (0.01)) * ((*data) * (0.01))));
}

TF1 *f = new TF1("fScalar", func, 100, 200, 4);
f->SetParameters(1, 1000, 7.5, 1.5);
TH1D h1f("h1f", "Test random numbers", 12800, 100, 200);
h1f.FillRandom("fvScalar", 1000000);
h1f.Fit(f);
```

```
//Higgs Fit: Implementation of the vectorized function
ROOT::Double_v func(const ROOT::Double_v *data, const double *params)
{
    return params[0] * exp(-(*data + (-130.)) * (*data + (-130.)) / 2) +
        params[1] * exp(-(params[2] * (*data * (0.01)) - params[3] *
            ((*data) * (0.01)) * ((*data) * (0.01))));
}

//This code is totally backwards compatible
TF1 *f = new TF1("fvCore", func, 100, 200, 4);
f->SetParameters(1, 1000, 7.5, 1.5);
TH1D h1f("h1f", "Test random numbers", 12800, 100, 200);
h1f.FillRandom("fvCore", 1000000);

//Added multithreaded fit option
h1f.Fit(f, "MULTITHREAD");
```

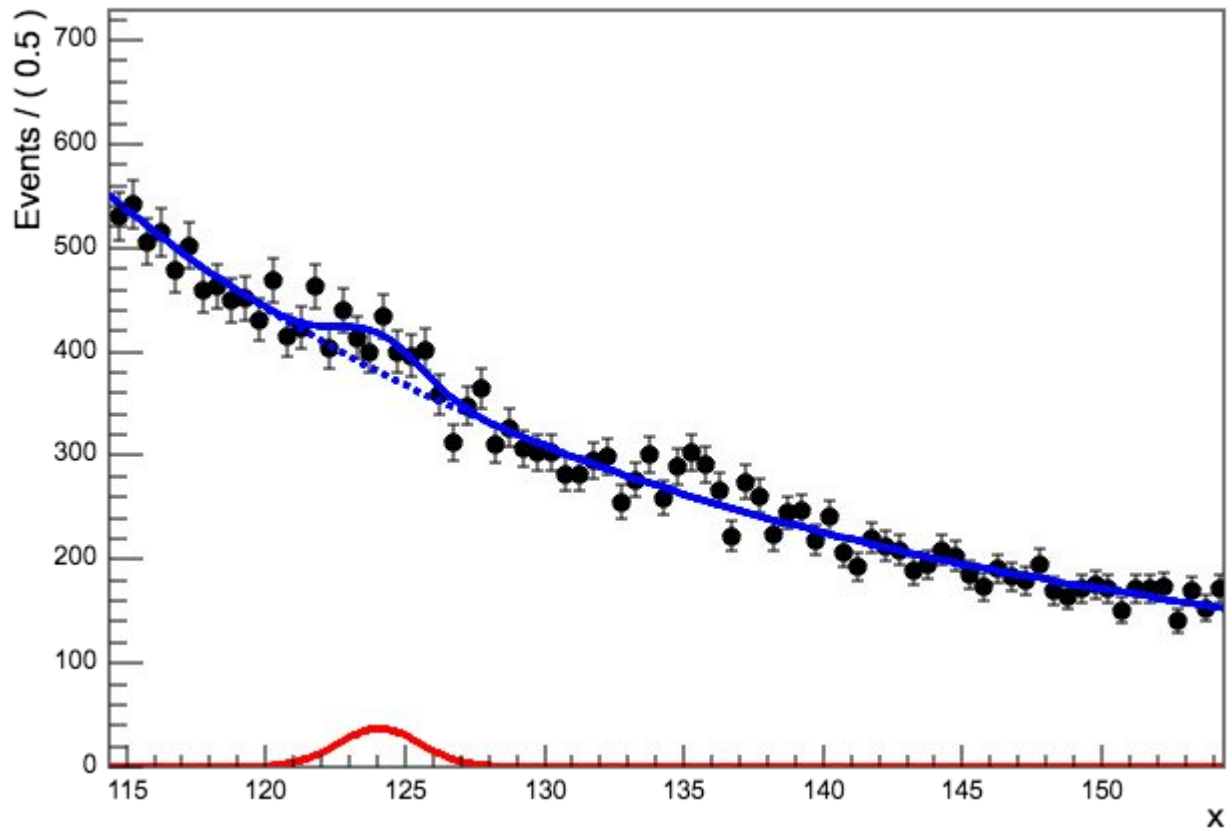
Backwards compatibility!

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//Added multithreaded fit option
h1f.Fit(f, "MULTITHREAD");
```

← Won't be needed with ImplicitMT!



VECTORIZATION



MAX 4

PARALLELIZATION



MAX 4

**PARALLELIZATION +
VECTORIZATION**

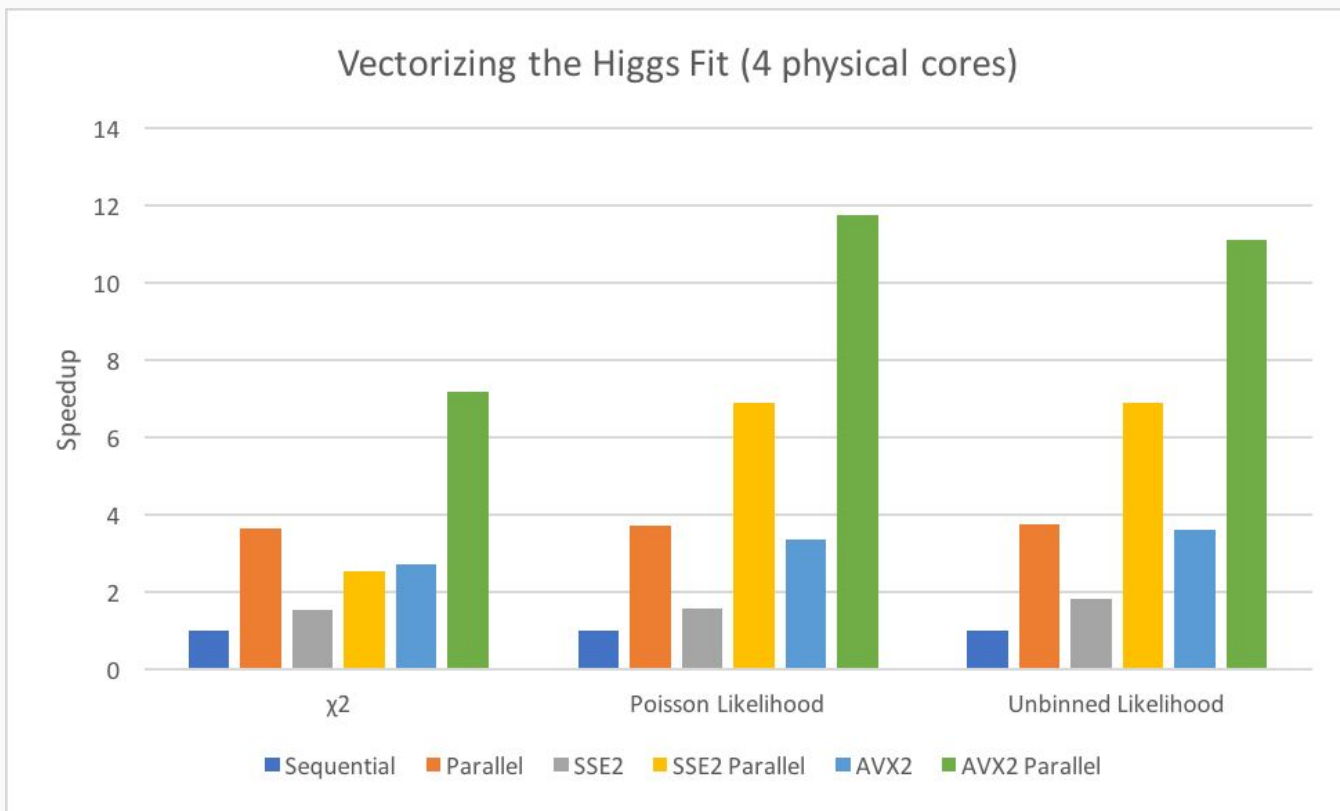


MAX 16

On your PC!

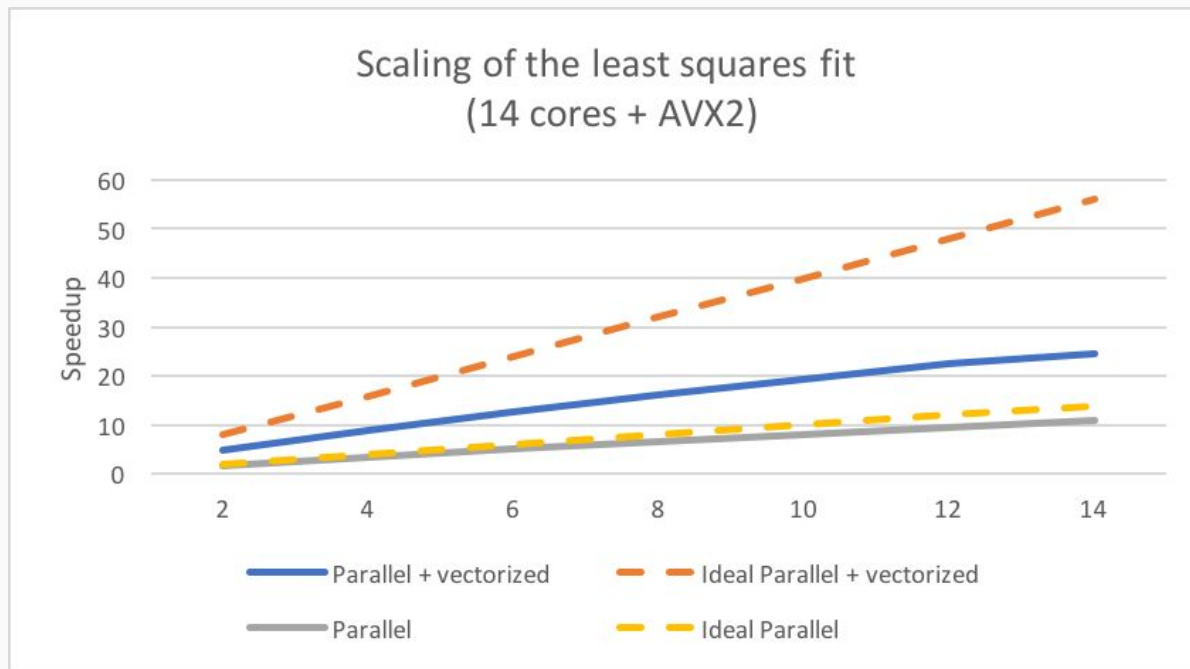


Big images, big results





Multithreaded Speed up



Compiler performance

