

# NMSU Update

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# Qt App fro Hodoscope Monitoring

- Qt5 framework(python/c++) was used.
- Qt frmaework is;
  - Cross-platform software for creating graphical user interfaces (KDE Plasma DE).
  - Easy-to-read code
  - Rich choice of modules.
  - Multiple libraries and high level of control.
  - API for easier development.
- Drawback: can be slow with `ssh`. Better to use textual.
- github

# CoDaS-HEP: Columnar Data Analysis

- A common feature among all array-oriented languages (except `Fortran 90`) is that they are also interactive languages.
- Typically, you perform one operation on a whole dataset, see what that does to the distribution, then apply another (easy to debug).
- ROOT :
  - `RDataFrame`
  - `RVec`
- python
  - `numpy`
  - `pandas`
  - `Awkward Array`
- `github`
- New kid in the block: `julia` (<https://github.com/JuliaLang/julia>)

## ■ Vectorization;

- Vectorization is effectively loop unrolling
  - In effect, the compiler unrolls by 4 iterations, if 4 elements fit into a vector register

```
for (i=0; i<N; i++) {  
    c[i]=a[i]+b[i];  
}
```



```
for (i=0; i<N; i+=4) {  
    c[i+0]=a[i+0]+b[i+0];  
    c[i+1]=a[i+1]+b[i+1];  
    c[i+2]=a[i+2]+b[i+2];  
    c[i+3]=a[i+3]+b[i+3];  
}
```



Load a(i..i+3)  
Load b(i..i+3)  
Do 4-wide a+b->c  
Store c(i..i+3)

- Parallel processing: OpenMP for multi-threading.
- github

## Results\*

- Original Serial pi program with 100000000 steps ran in 1.83 seconds.

```
#include <omp.h>
static long num_steps = 100000;    double step;
#define NUM_THREADS 2
void main ()
{  int i, nthreads; double pi, sum[NUM_THREADS];
  step = 1.0/((double) num_steps;
  omp_set_num_threads(NUM_THREADS);
  #pragma omp parallel
  {
    int i, id, nthrds;
    double x;
    id = omp_get_thread_num();
    nthrds = omp_get_num_threads();
    if (id == 0)  nthreads = nthrds;
    for (i=id, sum[id]=0.0; i< num_steps; i=i+nthrds) {
      x = (i+0.5)*step;
      sum[id] += 4.0/(1.0+x*x);
    }
    for(i=0, pi=0.0; i<nthreads; i++) pi += sum[i] * step;
  }
}
```

threads	1 <sup>st</sup> SPMD*
1	1.86
2	1.03
3	1.08
4	0.97

Intel compiler (icpc) with default optimization level (O2) on Apple OS X 10.7.3 with a dual core (four HW thread) Intel® Core™ i5 processor at 1.7 Ghz and 4 Gbyte DDR3 memory at 1.333 Ghz.

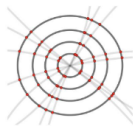
\*SPMD: Single Program Multiple Data

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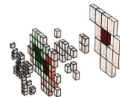
■ github

## Low-Level Reconstruction Tasks

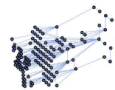
Combine detector signals  
to form "building blocks"  
for various particle types



Track Reconstruction

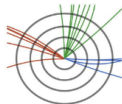


Calorimeter Segmentation

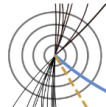


## Higher Level Particle-based Tasks

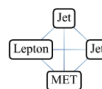
Given a set of particles, can  
we combine them to  
represent a specific decay?  
Can we identify a physics  
signal?



Jet Identification



Event Classification



GNNS AT THE LHC  
COMMON APPLICATIONS

# GNNS AT THE LHC

## WHY GNNS?

### **Common Justifications** (task-dependent)

- Many LHC datasets have inherent relational structure and/or no inherent ordering
- Grids, sequences, etc. cannot naturally represent irregular detector geometries
  - A small fraction of sensors are activated in any given event → data is sparse
  - Many different data sizes (particle counts, sensor readings, etc.)
- LHC data is heterogeneous
  - Data recorded from multiple subdetectors
  - Different types of particles
- Excellent performance
  - Relational inductive bias
  - Message passing leverages low-level detector info in addition to global (or otherwise human-devised) info
  - Generally smaller architectures (qualitatively speaking)

# For SpinQuest

- Vectorize/Parallelize the KTracker.
- Use GNN for track building.