

# MarlinKinfFit — Kinematic Fitting in MarlinReco

Jenny List

DESY

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Introduction — What & Why?

Basic Abstractions — How is it done?

Usage Example — How can it be used?

Summary — So what?

## Why a kinematic fit?

### **We plan to optimize the final physics performance!**

„improve resolution by exploiting a priori knowledge“

- ▶ „resolution “: (jet)energy, invariant masses, jet pairing, ...
- ▶ „a priori knowledge “: energy and momentum conservation, particle masses, vertices ....

### **How does it work?**

- ▶ Vary measured quantities within errors (sensible error parametrisation important!)
- ▶ enforce hard constraints (e.g.  $\sum E_j = \sqrt{s}$ ) by Lagrange multipliers
- ▶ „soft constraints “: additional  $\chi^2$  term

# Goals of MarlinKinfIt

## Would like

- ▶ not to reinvent the wheel!
- ▶  $\Rightarrow$  use well tested minimisation backend ( $\rightarrow$  LEP)
- ▶ to provide flexible machinery adjustable to many different configurations
- ▶ easy to use within ILC software  $\Rightarrow$  C++, Marlin

## Beyond this scope

- ▶ one processor which can do everything ( $WW$ ,  $ZZ$ ,  $ZH$ ,  $ZHH$ ,  $t\bar{t}$ , SUSY, ...)
- ▶ error parametrisation for particle momenta

## BaseFitObject: „reconstructed particles“

- ▶ four-momenta: measured / unmeasured parameters
- ▶ covariance matrix
- ▶ first partial derivatives of  $p_x, p_y, p_z, E$  w.r.t. to parameters
- ▶ contribution to global  $\chi^2$
- ▶ currently implemented:
  - ▶ JetFitObject: measured  $E, \theta, \phi$
  - ▶ NeutrinoFitObject: unmeasured,  $p_{x,y,z}, E$  from constraints
- ▶ new classes with different parametrisations welcome! (write them or ask me...)

## BaseConstraint: „a priori knowledge“

- ▶ Value: = 0 if constraint fulfilled
- ▶ list of FitObjects which contribute
- ▶ first partial derivatives w.r.t.  $p_x$ ,  $p_y$ ,  $p_z$ ,  $E$
- ▶ currently implemented: Px/y/zConstraint, EnergyConstraint, MassConstraint
- ▶ again, further contributions are welcome!

## BaseFitter: the fit engine itself

- ▶ list of `FitObjects`
- ▶ list of `Constraints`
- ▶ assembles global  $\chi^2$  and covariance matrix
- ▶ actual  $\chi^2$  minimisation
- ▶ currently implemented: `WWFitter` one-to-one  $F \rightarrow C$  translation of OPAL kinematic fit engine `WWFG0`. Results have to be checked to be identical down to machine precision.
- ▶ probably soon to come: `cernlib` free version (matrix inversion)

## Usage Example: $W^+ W^- \rightarrow 4$ jets, equal mass constraint

- ▶ Energy and momentum conservation:
  - ▶  $\Sigma E - \sqrt{s} = 0$
  - ▶  $\Sigma p_{x/y/z} = 0$
- ▶ equal mass constraint:  $M_{12} = M_{34}$
- ▶ try all three possible jet pairings!
- ▶ plots to come (just proof of principle):
  - ▶ 200 full hadronic  $WW$  events
  - ▶ LDC00Sc, 4T, Mokka 5.4
  - ▶ TrackCheater, TrackwiseParticleFlow (O.Wendt)

## Usage Example: $W^+ W^- \rightarrow 4$ jets, equal mass constraint

```
JetFitObject fitjets[NJETS];
for (int i = 0; i < 4; i++) {

    ReconstructedParticle* j = jetcol->getElementAt( i );
    lvec = HepLorentzVector ((j->getMomentum())[0], (j->getMomentum())[1],
(j->getMomentum())[2], j->getEnergy());

    fitjets[i] = new JetFitObject (lvec.e(), lvec.theta(), lvec.phi(), erre, errtheta, errphi);
}

FourJetPairing pairing (fitjets);
JetFitObject *permutedjets[NJETS];

for (int iperm = 0; iperm < pairing.getNPerm(); iperm++) {

    pairing.nextPermutation (permutedjets);
    PxConstraint pxc;
    for (int i = 0; i < NJETS; ++i) pxc.addToFOList (*(permutedjets[i]));

    PyConstraint pyc;
    for (int i = 0; i < NJETS; ++i) pyc.addToFOList (*(permutedjets[i]));

    PzConstraint pzc;
    for (int i = 0; i < NJETS; ++i) pzc.addToFOList (*(permutedjets[i]));

    EConstraint ec(500.);
    for (int i = 0; i < NJETS; ++i) ec.addToFOList (*(permutedjets[i]));
```



Usage Example:  $W^+ W^- \rightarrow 4$  jets, equal mass constraint

```

MassConstraint w(0.);
w.addToFOList (*(permutedjets[0]), 1);
w.addToFOList (*(permutedjets[1]), 1);
w.addToFOList (*(permutedjets[2]), 2);
w.addToFOList (*(permutedjets[3]), 2);

WWFitter fitter;
for (int i = 0; i < NJETS; ++i) fitter.addFitObject (*(permutedjets[i]));
fitter.addConstraint (pxc);
fitter.addConstraint (pyc);
fitter.addConstraint (pzc);
fitter.addConstraint (ec);
fitter.addConstraint (w);

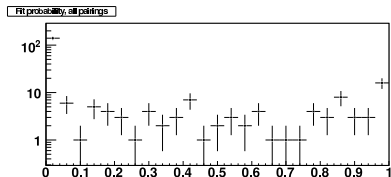
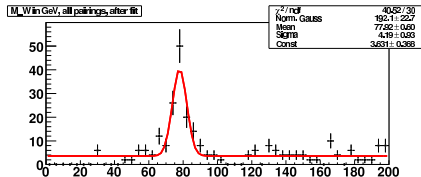
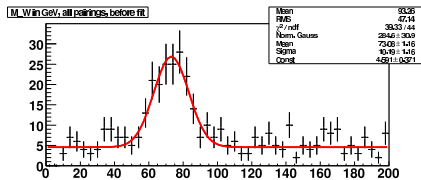
double prob = fitter.fit();
cout << " fit probability = " << prob << endl;
cout << " error code: " << fitter.getError() << endl;
for (int i = 0; i < NJETS; ++i) cout << "final four-vector of jet " << i << " = " <<
*(permutedjets[i]) << endl;

cout << "final mass of W 1:" << w.getMass(1) << endl;
cout << "final mass of W 2:" << w.getMass(2) << endl;
}

```

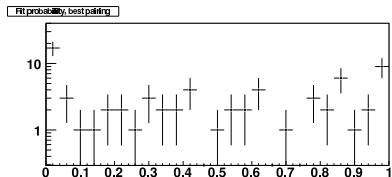
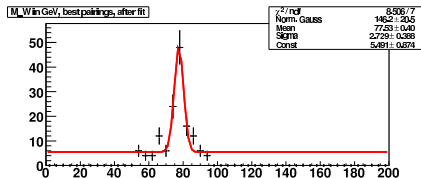
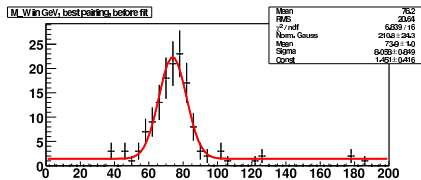
## Results: All jet pairings

- ▶  $M_W$  before fit:  
 mean =  $(73.1 \pm 1.2)$  GeV,  
 $\sigma = (10.2 \pm 1.2)$  GeV
- ▶  $M_W$  after fit:  
 mean =  $(77.9 \pm 0.6)$  GeV,  
 $\sigma = (4.2 \pm 0.9)$  GeV
- ▶ Fit probability: resonably flat  
 (caveat: jet errors put  
 „by hand“ to  $100\%/\sqrt{E}$   
 and 10 mrad for  $\theta$  and  $\phi$ )



## Results: Best jet pairing

- ▶  $M_W$  before fit:  
mean =  $(73.9 \pm 1.0)$  GeV,  
 $\sigma = (8.1 \pm 0.9)$  GeV
- ▶  $M_W$  after fit:  
mean =  $(77.5 \pm 0.4)$  GeV,  
 $\sigma = (2.7 \pm 0.4)$  GeV
- ▶ Fit probability: flat  
(caveat: jet errors put  
„by hand“ to  $100\%/\sqrt{E}$   
and 10 mrad for  $\theta$  and  $\phi$ )



## Summary

- ▶ MarlinKinfitter provides a flexible interface to an established fitter
- ▶ it is waiting to be used!
- ▶ check out from MarlinReco in `MarlinReco/[version]/Analysis/MarlinKinfitter` including
  - ▶ README
  - ▶ doxygen documentation
  - ▶ note on the minimisation mathematics behind
  - ▶ and the *WW* example
- ▶ or look at: <http://www.desy.de/~blist/kinfitter>
- ▶ feed-back, questions, complaints, more examples, fit objects welcome: `jenny.list@desy.de`

**... and happy fitting!**