

Kinematic fitting in the presence of ISR

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DESY

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Introduction
MarlinKinfIt
Modelling ISR
Application to SM 4-jet events
Summary & Outlook



Motivation

- ▶ kinematic fitting uses known constraints of the events (e.g. $\sum p_x = 0$) to improve measured quantities and/or determine unmeasured quantities
- ▶ big improvement of jet energy resolution at LEP
- ▶ problem at ILC: ISR and Beamstrahlung¹
- ▶ with naive approach: loose lots of events, bias di-jet masses for remaining ones

¹„ISR“ refers to both effects throughout the talk

Kinematic Fitting

- ▶ minimize $\chi^2 = (\vec{a} - \vec{\eta})^T Cov^{-1}(\vec{a} - \vec{\eta})$,
 \vec{a} : fitted parameters, $\vec{\eta}$: measured parameters
- ▶ apply “hard” constraints: $G_k(\vec{a}, \vec{\xi}) = 0$,
 $\vec{\xi}$: unmeasured parameters
- ▶ method: Lagrange multipliers: $\chi^2_T = \chi^2 + \sum \lambda_k G_k$,
 seek stationary point:
 - $\frac{\partial \chi^2_T}{\partial a_i, \xi_j} = 0 \quad \rightarrow$ best values of parameters
 - $\frac{\partial \chi^2_T}{\partial \lambda_k} = G_k(\vec{a}, \vec{\xi}) = 0 \quad \rightarrow$ constraints are fulfilled
- ▶ needs multidimensional equation solver \rightarrow iterative method

MarlinKinfit: Software Structure

3 basic concepts:

- ▶ FitObject: e.g. JetFitObject
 - ▶ has several parameters, measured or unmeasured
 - ▶ can calculate χ^2 contribution and derivatives
 - ▶ can calculate 4-vector
- ▶ Constraint: e.g. MassConstraint
 - ▶ is formulated in terms of 4-vectors (from FitObjects)
 - ▶ can calculate constraint function G_k and derivatives
- ▶ Fitter: e.g. OPALFitter
 - ▶ manages the fitting problem, assigns numbers to parameters and constraints
 - ▶ sets up and solves system of equations

Code (with example) available as Marlin processor

Example Code

```
//          E  theta phi    dE dtheta dphi mass
JetFitObjet jet1 (44., 1.2, 0.087, 5.0, 0.2, 0.1, 0.);
JetFitObjet jet2 (46., 1.8, 3.120, 5.0, 0.2, 0.1, 0.);

// Constraint 0*sum(E) + 1*sum(px) + 0*sum(py) + 0*sum(pz) = 0
Momentumconstraint pxconstraint (0, 1, 0, 0, 0);
pxconstraint.addToFOList (jet1);
pxconstraint.addToFOList (jet2);

OPALFitter fitter;

fitter.addFitObject (jet1);
fitter.addFitObject (jet2);

fitter.addConstraint (pxconstraint);

double prob = fitter.fit();
```

Soft Constraints

- ▶ “soft” constraints: Add additional terms to total χ^2
- ▶ example: $\Delta m(\vec{a}, \vec{\xi})$ is mass difference of two dijet-pairs
→ add $(\Delta m/\sigma)^2$ to total χ^2
- ▶ allows to take into account constraints that are not exactly fulfilled
- ▶ useful to avoid very small fit probabilities
- ▶ if unmeasured quantities are only constrained via “soft” constraints, some methods to solve equation system fail (e.g. OPALFitter)

Available Fit Engines

- ▶ OPALFitter
 - ▶ implementation of kinematic fit used at OPAL, (see L. Lyons: *Statistics for nuclear and particle physicists*, Cambridge Univ. Pr. 1986)
 - ▶ well tested, robust → our reference implementation
 - ▶ cannot treat case of soft constraints with unmeasured parameters
- ▶ NewtonFitter
 - ▶ new development, uses multidimensional Newton-Raphson method
 - ▶ can treat soft constraints with unmeasured parameters
 - ▶ needs (and uses!) 2nd derivatives of constraint functions → slightly better convergence
 - ▶ the plots in the rest of the talk are made with the NewtonFitter

SM 4-jet events

SLAC Whizard SM sample

- ▶ $e^+e^- \rightarrow u\bar{d}d\bar{u}$, 500 fb^{-1}
(mostly W^+W^- , no ν from semileptonic c/b decays)
- ▶ full simulation of LDCprime
- ▶ full reconstruction, Particle Flow: Pandora
- ▶ force event into four jets (Durham4Jets on DST)
- ▶ require for each jet:

| | events | |
|--------------------------------------|--------|--------|
| N_{tot} | 52490 | |
| $ \cos\theta_{\text{jet}} < 0.989$ | 39940 | 76.1 % |
| $N_{\text{track}}/\text{jet} \geq 1$ | 38925 | 74.2 % |
| $E_{\text{jet}} > 4.5 \text{ GeV}$ | 38912 | 74.1 % |

ISR classification

select (MC info) events with either

- ▶ no significant ISR:

$$\sum E_{\gamma} < 5 \text{ GeV}$$

→ 24731 ev. (64 %)

- ▶ with significant ISR:

$$\geq 1 \gamma \text{ with } E_{\gamma} > 25 \text{ GeV, } \cos \theta_{\gamma} > 0.999 \rightarrow 6699 \text{ ev. (17 \%)}$$

- ▶ later:

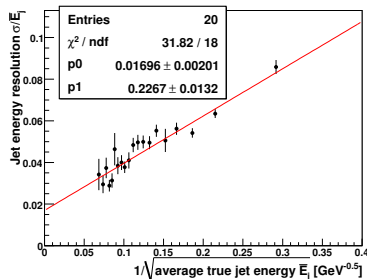
→ 7482 ev. (19 %)

- ▶ photons within detector acceptance
- ▶ „soft“ ISR

Jet Parametrisation

roughly determined from „no ISR“ sample:

- ▶ Jet energy scale: $E = 1.01 \cdot E_{\text{rec}}$
- ▶ $\delta E = 23\% \cdot \sqrt{E} + 1.7\% E$
(from $u\bar{d}d\bar{u}$ sample, not single hadrons)
- ▶ $\delta\theta = 0.01$ rad
- ▶ $\delta\phi = 0.01$ rad



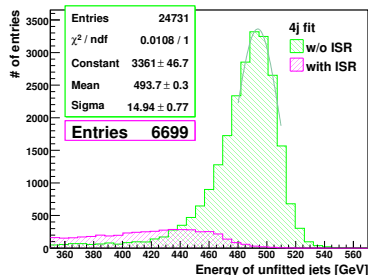
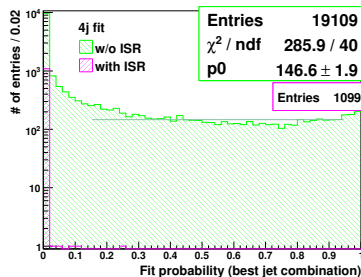
4 Jet Hypothesis

Constraints

- ▶ $\sum E^{\text{jet}} = E_{\text{CM}}$
- ▶ $\sum p_x^{\text{jet}} = 0$
- ▶ $\sum p_y^{\text{jet}} = 0$
- ▶ $\sum p_z^{\text{jet}} = 0$
- ▶ equal di-jet masses

...as expected:

- ▶ works well on events w/o ISR
- ▶ hopeless for ISR sample....



Modelling ISR

Soft Constraints

- ▶ don't force $E_{\text{CM}} = 500 \text{ GeV}$ and $\sum p_z = 0$
- ▶ use „soft“ constraints instead,
i.e. no Lagrange multiplier but just additional χ^2 contribution depending on deviation from naive beam constraint
- ▶ problem: need to keep E_{CM} and $\sum p_z$ consistent!
- ▶ could constrain $E_{\text{CM}} \pm \sum p_z \dots$
- ▶ in combination with unmeasured parameters (ν , LSP):
minimisation get's more involved, need special fit engine!
- ▶ currently not pursued

Modelling ISR

PhotonFitObject, „unmeasured“

- ▶ treat photon parameters as unmeasured
- ▶ this „costs“ constraints
- ▶ usually not applicable for final states with true E_{miss} (ν , LSP)

PhotonFitObject, measured

- ▶ treat photon parameters as measured
- ▶ choose error parametrisation such that fitted E_γ spectrum reproduces ISR spectrum
- ▶ advantage: no loss of constraints!

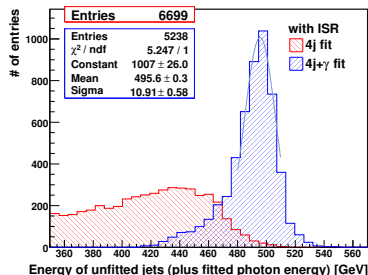
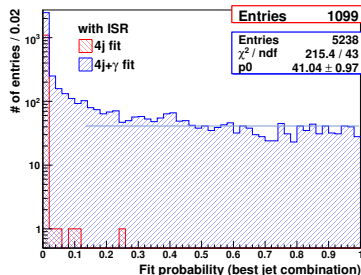
Application to SM 4-jet events

Photon Parametrisation

- ▶ one parameter: p_z
- ▶ others fixed via
 $E = |p_z|, p_x = p_y = 0$
- ▶ treat p_z as *measured!*
- ▶ with currently Gaussian error: $p_z = 0 \pm 100$ GeV

„with ISR “ sample

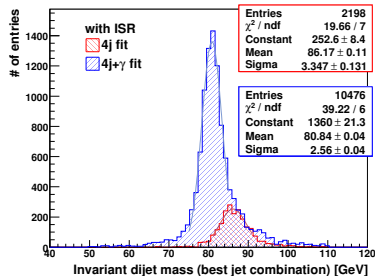
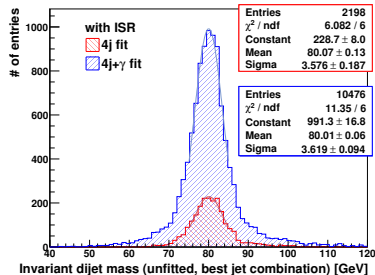
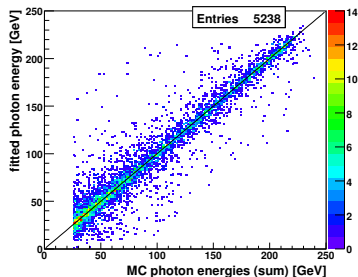
- ▶ vast improvement!
- ▶ converged fits 78%
- ▶ fit prob flat, $E_{\text{tot}} \simeq 500$ GeV



4 Jet + Photon Hypothesis

„with ISR “ sample

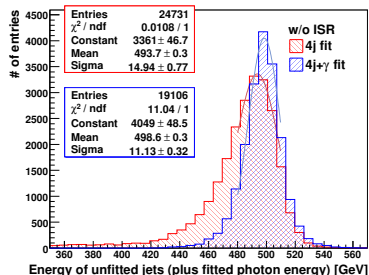
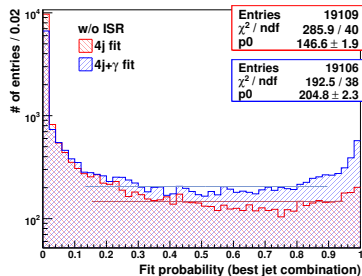
- ▶ nice correlation between E_{γ}^{rec} and E_{γ}^{MC}
- ▶ no shift in mass peak
- ▶ $\sigma_{M_{jj}}$: **3.6 GeV** \rightarrow **2.6 GeV**



4 Jet + Photon Hypothesis

„no ISR “ sample

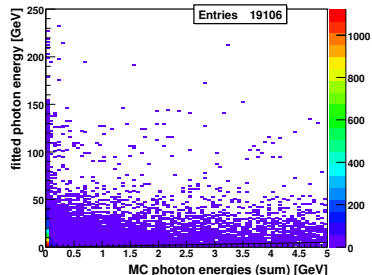
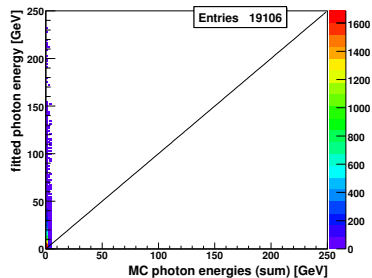
- ▶ higher fit probabilities
- ▶ total reconstructed energy shifts by $\simeq 5$ GeV
→ not too bad, since up to 5 GeV ISR allowed also in this sample



4 Jet + Photon Hypothesis

„no ISR “ sample

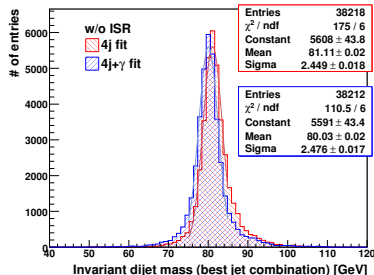
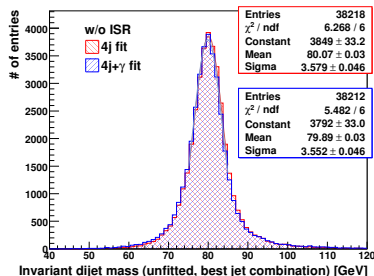
- ▶ majority of events gets only little fake photon energy
- ▶ some events with huge fitted photon energies
→ investigate...
- ▶ improvement expected with
 - ▶ proper probability density describing ISR spectrum
 - ▶ optimized jet error parametrisation



4 Jet + Photon Hypothesis

„no ISR “ sample

- ▶ convergence: 77% for both fits
- ▶ mostly same jet pairing „best“
- ▶ 4 jet fit: $\langle M_{jj} \rangle = 80.1$ GeV
→ 81.1 GeV
- ▶ nearly no bias by $4j+\gamma$ fit:
 $\langle M_{jj} \rangle = 80.0$ GeV
- ▶ $\sigma_{M_{jj}} = 2.5$ GeV for both fits

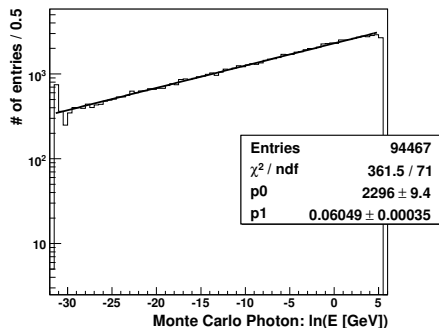


Summary

- ▶ di-jet mass resolution can be significantly improved by kinematic fits even in the presence of ISR
- ▶ already with a very basic photon parametrisation significant amounts ISR energy can be well reconstructed
- ▶ many details not yet optimized
⇒ possibilities for improvement!
- ▶ SM $u\bar{d}d\bar{u}$ events with and without ISR:
$$\sigma_{M_{jj}} = 3.6 \text{ GeV} \rightarrow 2.6 \text{ GeV}$$
- ▶ fitting with ISR hypothesis:
same resolution gain as w/o ISR,
but **without any bias** on reconstructed dijet mass!
- ▶ PhotonFitObject will be in next MarlinReco release

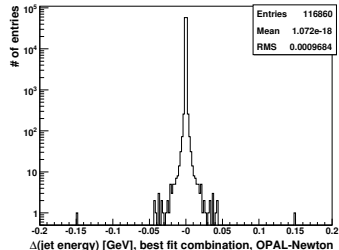
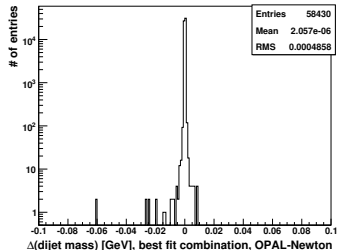
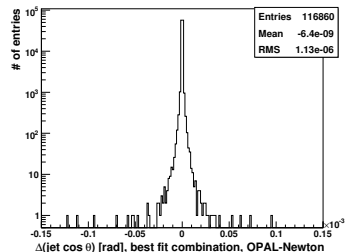
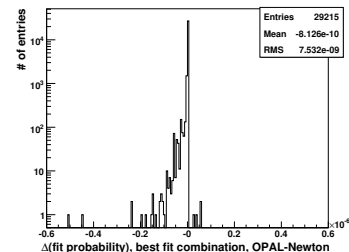
Outlook

- ▶ serious photon parametrisation
- ▶ include events with little ISR
- ▶ look at events with photon in detector (5 jet collection?)
- ▶ try other quark flavours
- ▶ optimize jet energy scale, jet error parametrisation, convergence criteria, ...
 - ▶ maybe look at soft constraint approach for comparison
 - ▶ apply to Chargino / Neutralino separation
 - ▶ write LCNote and Dipoma Thesis (Moritz :-)



BACKUP

OpalFitter vs NewtonFitter



OpalFitter vs NewtonFitter

