



Update on Kinematic Fits

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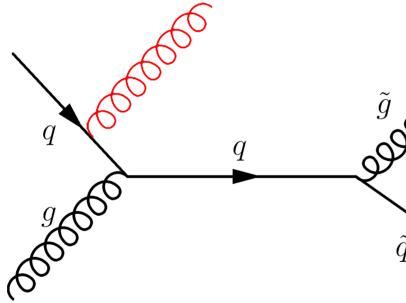
Uni HH - Susy Group Meeting - 21st February 09



Potential Problems



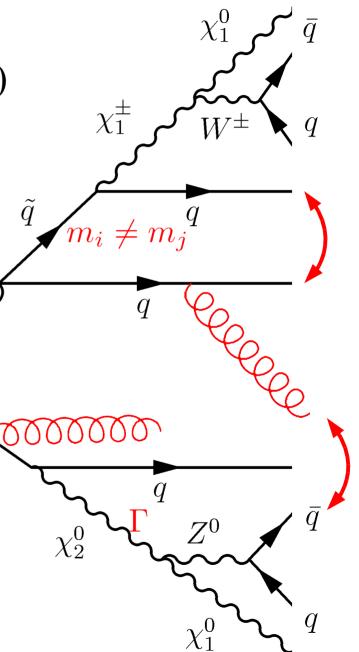
- Suppression of SM and SUSY background
- 7 jets in final state (huge combinatorial bg)
 - all reconstructed
- No perfect mass degeneration



Width of virtual particles

additional jets important for momentum balance

• +ISR





Commonly Used Approach



Constrained fitting via Lagrangian Multiplier

$$\sum_{i=1}^{N_m} \left(\frac{m_i - t_i}{\sigma_i}\right)^2 + 2 \cdot \sum_{i=1}^{N_c} \lambda_i \cdot f_i$$

- Find extremum where all derivatives vanish
- If constraints are linear find perfect solution within one Gauss-Newton step
- Invariant mass constraints are (highly) non-linear → linearization and iterative approach
- General problem: Fit can converge at local (and not global) minimum



Alternative Approach



• Formulation of constraints as additional χ^2 term \rightarrow "cost function"

• To interpret cost function as χ^2 all correlations have to be taken

into account

$$\left(\frac{M_{\rm inv}(j_1,j_2,j_3)-M}{\sigma}\right)^2$$

with

$$\sigma^2 = \sum_{i=1}^{N_m} \left(\frac{\partial M_{\text{inv}}}{\partial i} \right)^2 \cdot \sigma_i^2 + \Gamma_m^2$$

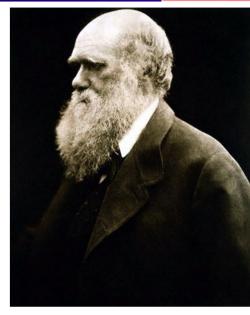
 Minimize cost function: many possible algorithms (gradient, simplex, LBFGS, simulated annealing ... genetic algorithm)



Genetic Algorithm



- Final state 4-momenta are properties (genes) of individuum; jet combination is one additional gene
- A fitness function (here χ^2) defines if a *individuum* is *fittest*
- 1) Starting from starting values create a first *generation* of *individua* (starting *population*): use all possible jet combinations (1260 for 7 jets)
- 2) Select N best fitting individua (here 20)
- 3) Create M (here 1000) new *individua* by selecting randomly two *parents* and take randomly the *genes* from either one or the other *parent*
- 4) Mutate (variation within the measurement errors) each gene (except jet combination) with a given probability (here 10 %)
- 5) Back to step 2) until convergence is reached (here: no change within 3 *generations*) or fixed iteration number is reached (here 200)

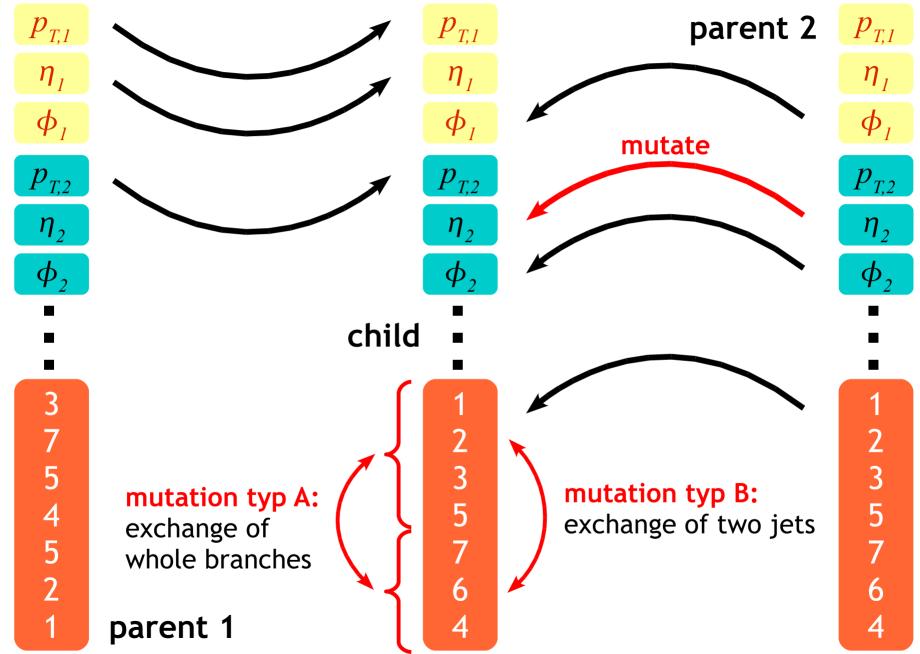


Charles Darwin (1809 - 1882) On the origin of species (1859)



Genetic Algorithm (cont.)







Starting Values for Neutralinos



- In typical Susy scenarios: $m_{\chi_1^\pm} m_{\chi_1^0} \gtrsim m_W$
- \rightarrow small relative momentum of W and χ^0
- Assume same direction of W and χ^0 and adjust χ^0 momentum to fulfill mass constraint

$$0 \stackrel{!}{=} f(x) = m_{\chi_1^{\pm}}^2 - \left(\left(E_W + \sqrt{m_{\chi_0}^2 + (x \cdot p_W)^2} \right)^2 - (1+x)^2 p_W^2 \right)$$

Solve analytically or numerically by Newtons method

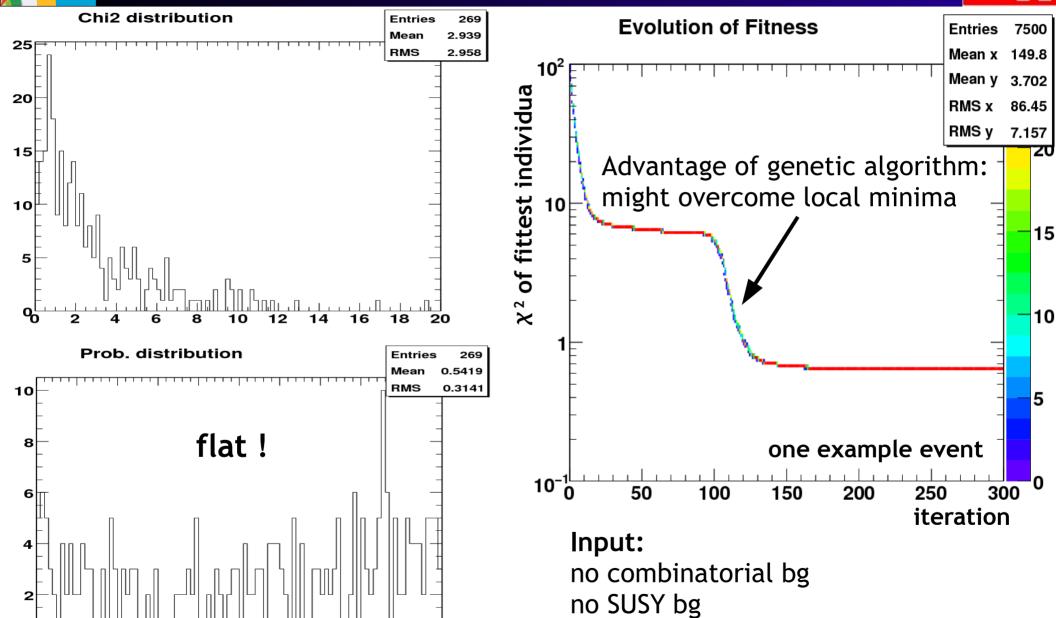
$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

CMS

Fundamental Behaviour

ISR for momentum balance from MC



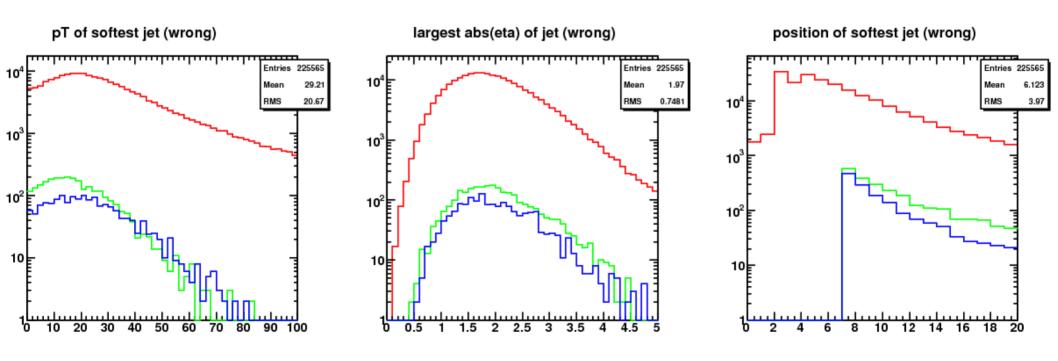




Preselection



- Demand exactly 7 jets with:
 - $p_T > 20 \text{ GeV}$
 - $|\eta| < 3.0$
- This will reject many signal events with hard ISR or FSR, but combinatorics is reduced





constraints)

Dependence on Starting Values



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Before genetic algorithm:
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Chi2 (staring values, no combinatorics) = 18.6535
probability = 1.56763e-05
Chi2 (KinFitter, no combinatorics) = 4.09197
probability = 0.251702
After genetic algorithm:
Chi2 (fitted values) = 8.03493
                                        KinFitter result (and convergence)
                                        depends strongly on starting values
probability = 0.00458838
Chi2 (KinFitter, starting values and combinatorics from gen.
 fit) = 1.15671
probability = 0.763404
\chi^2 of genetic Fit and KinFitter not directly comparable (different treatment of mass
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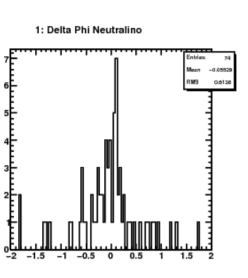
21st April 09 Kinematic Fits 10

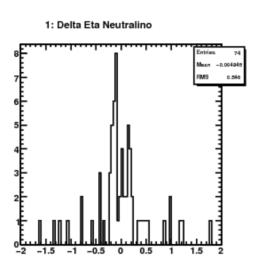


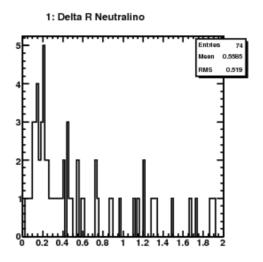
Resolution of Unmeasured Particles

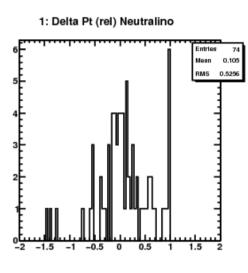


• Genetic algorithm: RMS ($\triangle R$) = 0.56

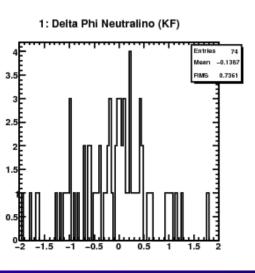


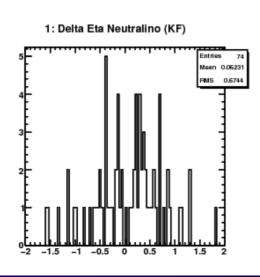


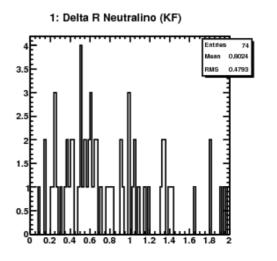


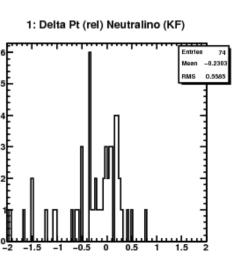


• KinFitter (with starting values from genetic fit): RMS ($\triangle R$) = 0.50











Fit Results Including Bgs



- Included backgrounds:
 - SUSY

signal cascades but not all signal jets selected (at least one ISR or FSR jet is

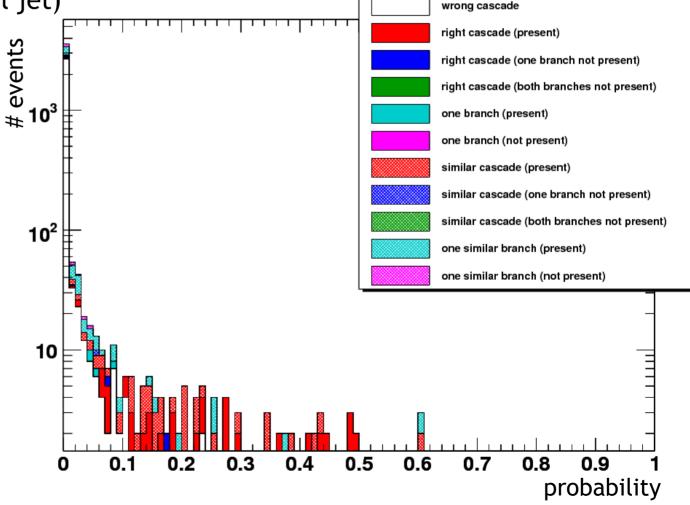
harder than one signal jet)

Signal cascade: only left handed squarks of first two generations

Similar cascades: other squarks

→ Prob. distribution not flat due to missing ISR

→ SUSY bg peaks at small probabilities





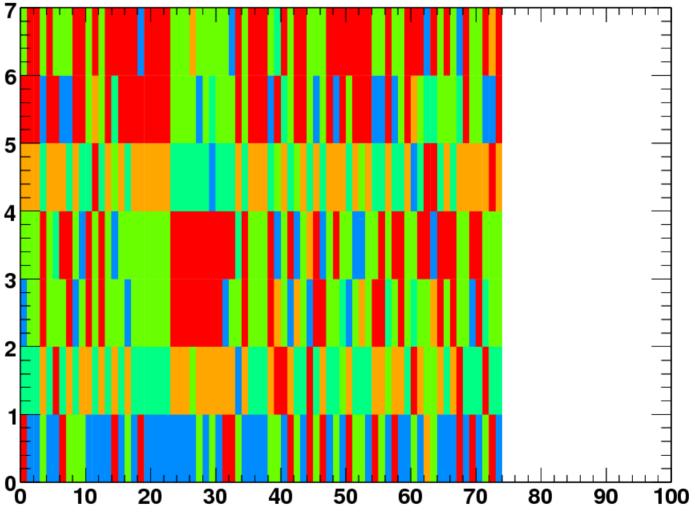
Visualization of Best Combinations



~ 20% complete right cascade most wrong combinations are exchange of the two branches









Plans and Ideas



- Scan over mass hypothesis or mSUGRA parameters
 - Use NAF —
 - First try with: $M_1: M_2: M_3 = 1:2:6$



- Use leptonic Z-decays (smaller combinatorics, bg, and statistics)
- Include ISR jets in event selection (treat higher statistics vs. larger combinatorial bg) (maybe not for SUSY09)
- Include systematic uncertainties (likely not for SUSY09)
- Think about "dynamic hypothesis selection" (surely not for SUSY09)