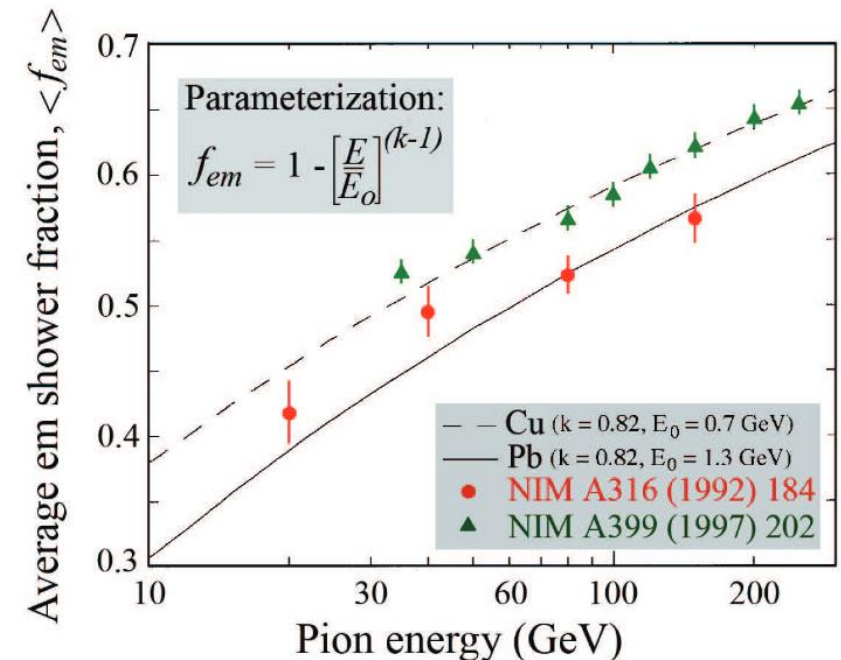


# **JShowerReconstruction**

(from a newcomer's perspective)

# Why not one all-purpose shower fitter?

- Low-energy (1-20 GeV) showers are difficult to detect!
  - Few hit coincidences (**low L1 hit probability**)
- Relatively **large hadronic components**
  - Vast majority (~90%) of hadronic shower particles are pions
  - Neutral pions decay into  $2\gamma$ 's  
--> EM showers
  - At lower shower energies, smaller EM component (from ~30% at 10 GeV to ~50% at 100 GeV)
  - The higher the hadronic component, the higher the fraction of **invisible energy** (due to nuclear interactions)



R. Wigmans – Lecture series on *Calorimetry* (2008)

# The reconstruction chain

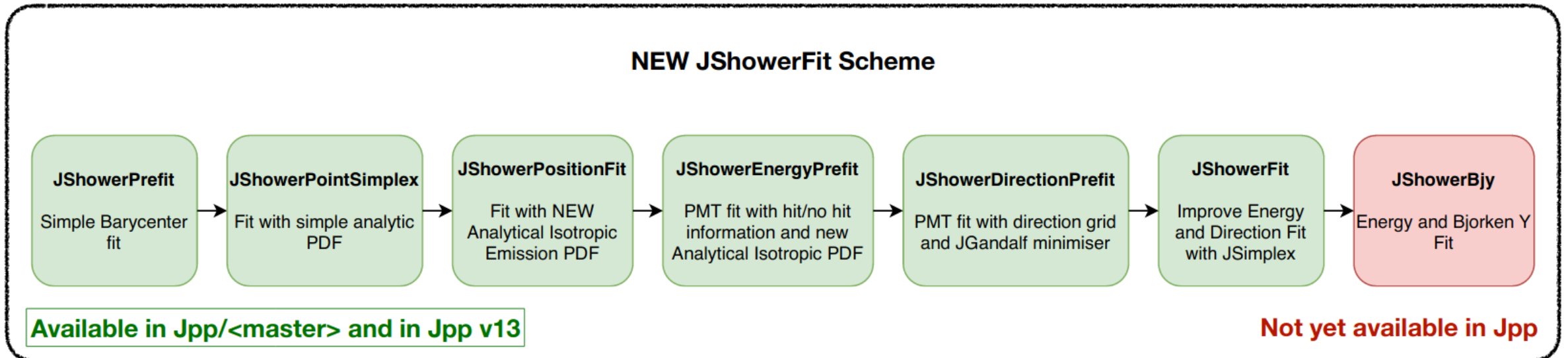
- At present, six parts:

- I. Vertex + direction prefit
- II. 1st vertex likelihood minimization (Powell's method)
- III. 2nd vertex likelihood minimization (Levenberg-Marquardt)

Vertex

- 
- IV. Energy prefit
  - V. 2nd direction prefit
  - VI. Energy and direction likelihood minimization (Powell's method)

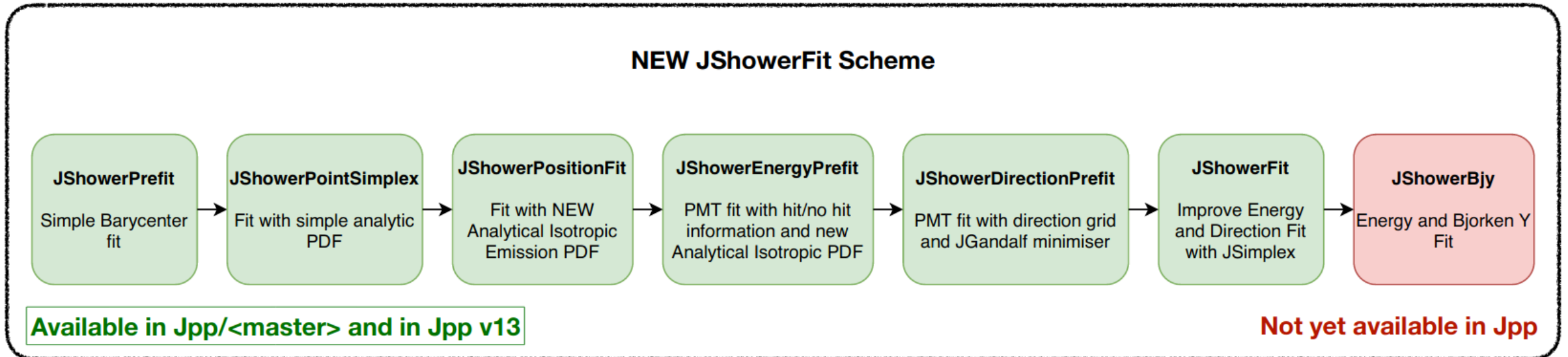
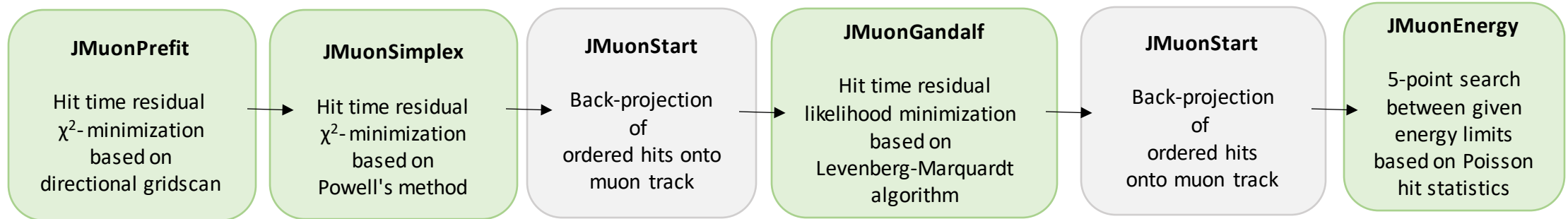
Energy + direction



*From Alba's June 2020 collaboration meeting presentation*

# The reconstruction chain

- At present, six parts
- C.f. the JMuonReconstruction chain:



*From Alba's June 2020 collaboration meeting presentation*

# Vertex prefit

- The vertex prefit consists of roughly 2 steps:

## I. Hit selection

- Find a cluster of causally related hits originating from a single emission point (vertex)
- Minimize background

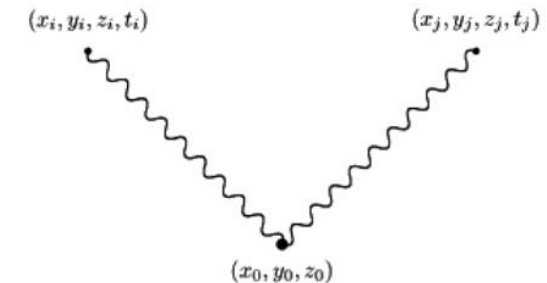
## II. Least squares vertex (barycenter) fit based on **time of flight**

- Find the neutrino vertex which minimizes squared hit time residuals

template<>

class JFIT::JEstimator< JPoint4D >

Linear fit of bright point (position and time) between hits (objects with position and time).



$$t_j = t_0 + \frac{c}{n} \times \sqrt{(x_j - x_0)^2 + (y_j - y_0)^2 + (z_j - z_0)^2}$$

where:

$x_0$  = x position of vertex (fit parameter)  
 $y_0$  = y position of vertex (fit parameter)  
 $z_0$  = z position of vertex (fit parameter)  
 $t_0$  = time at vertex (fit parameter)

$c$  = speed of light (in vacuum)  
 $n$  = index of refraction

Defining:

$$\begin{aligned} t'_j &\equiv nct_j \\ t'_0 &\equiv nct_0 \end{aligned}$$

$$\Rightarrow (t'_j - t'_0)^2 = (x_j - x_0)^2 + (y_j - y_0)^2 + (z_j - z_0)^2$$

The parameters  $\{x_0, y_0, z_0, t_0\}$  are estimated in the constructor of this class based on consecutive pairs of equations by which the quadratic terms in  $x_0, y_0, z_0$  and  $t_0$  are eliminated.

# Vertex prefit (hit selection)

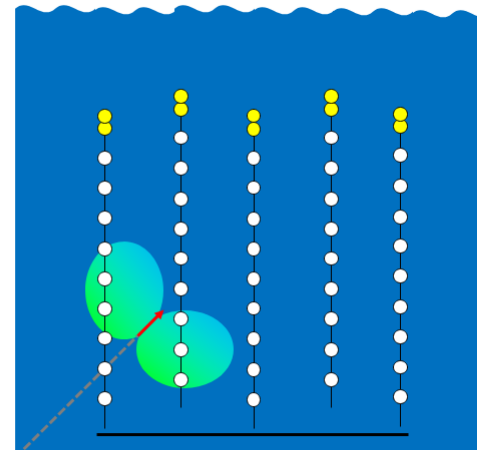
- Based on the MX (mix) trigger
  - Uses a mix of L0 and L1 hits

Hit rates:

- RL0  $\sim 115 \times 18 \times 31 \times 5 \text{ kHz} = 320 \text{ MHz}$
- RL1  $\sim 115 \times 18 \times 1 \text{ kHz} = 2 \text{ MHz}$

In typical time window of  $\sim 100 \text{ ns}$ ,  
expect  $\sim \mathbf{O(1)}$  L1 hit and  $\sim \mathbf{O(10-100)}$  L0 hits

- Which L0's are causally related to a L1 hit?
  - Low energy showers are well-contained  
--> **use distance** as selection criterion



size of shower  
much smaller than  
size of detector

↓

limit distance  
between  
L1/L0 hits

See Maarten's presentation on JTriggerMXShower

# Vertex prefit (hit selection)

- Start with a cluster of causally connected L2s ( $\Delta t < 20$  ns and  $\alpha < 135$  deg) that satisfy:

$$|t_i - t_j| \leq \boxed{|\vec{x}_i - \vec{x}_j|/c_{water}} + \boxed{T_{extra}} \quad \text{with } |\vec{x}_i - \vec{x}_j| = 50 \text{ m and } T_{extra} = 20 \text{ ns.}$$

i.e.: hit time difference smaller than **TOF between the two hit DOMs**  
+ **some extra time** (to account for scattering and hit time resolution)

$$c_{water} \sim 0.29979 / 1.34 \sim 0.22 \text{ m/ns} \longrightarrow |t_i - t_j| \leq 220 \text{ ns}$$

- Subsequently check if any surrounding L0s are causally connected with either of the L2 hits (using same condition) and add them to the time-sorted L2 cluster
- Matching operator defined in **JMatch3G**

```
/**
 * Match operator.
 * \param first hit
 * \param second hit
 * \return match result
 */
virtual bool operator()(const JHit_t& first, const JHit_t& second) const
{
    t = fabs(first.getT() - second.getT());

    if (t > TMax_ns) {
        return false;
    }

    x = first.getX() - second.getX();
    y = first.getY() - second.getY();
    z = first.getZ() - second.getZ();
    d = sqrt(x*x + y*y + z*z);

    if (d <= 0.5 * DMax_m)
        return t <= d * getIndexOfRefraction() * getInverseSpeedOfLight() + TMaxExtra_ns;
    else if (d <= DMax_m)
        return t <= (DMax_m - d) * getIndexOfRefraction() * getInverseSpeedOfLight() + TMaxExtra_ns;

    return false;
}
```

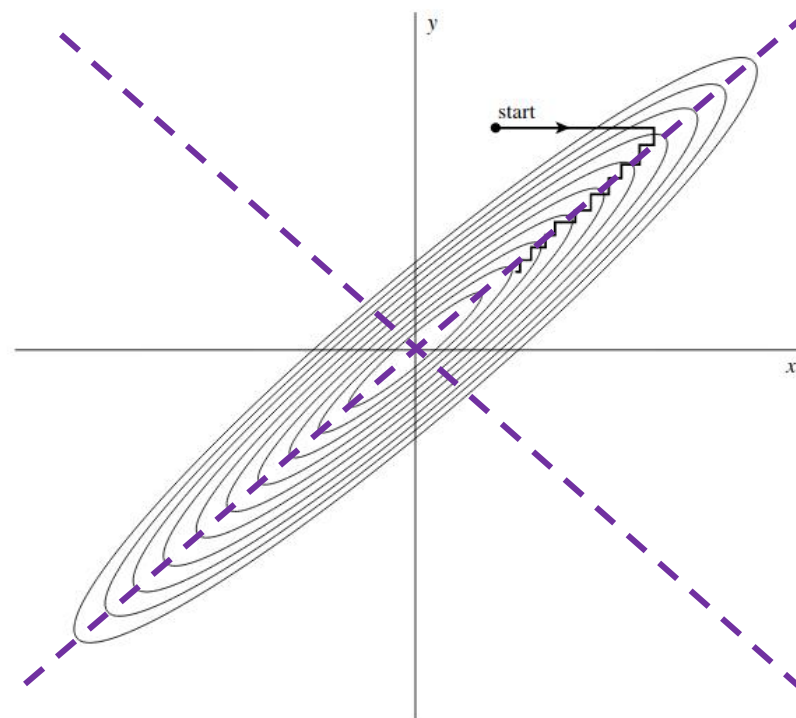
## Vertex prefit (fitting algorithm)

- The least squares fit for the arrival time residuals is largely analogous to what is done in JMuonPrefit
  - See also the appendix of Brían's track reconstruction document
- But it involves additional iterative steps, meant to enhance the precision (whilst limiting CPU time!)
  1. An extra distinction based on L2 cluster size:
    - If total L2+L0 hits,  $N_{\text{tot}} < 41$ :
      - i. Repeat the chi2-minimization  $N_{\text{tot}} - 4$  times, removing 1 L0 hit in each iteration
      - ii. Choose the fit which yields the least chi2
    - If total L2+L0 hits,  $N_{\text{tot}} > 40$ :
      - i. Perform the chi2-minimization just once, using all hits within the cluster



# JSimplex

- JSimplex performs a first, simple likelihood minimization, using **Powell's method**:
  - Iteratively finds a set of  $N$  mutually **conjugate directions**, with which to minimize the likelihood landscape
  - Does not require any gradients!
- For each step, the likelihood is re-evaluated:



∏  
hits

```
double JFIT::JRegressor< JPoint4D, JSimplex >::operator() ( const JPoint4D & vx,  
                                                         const JHit_t & hit  
                                                         ) const inline
```

Definition at line 50 of file JPoint4DRegressor.hh.

```
51 {  
52     using namespace JPP;  
53     const double dt = hit.getT() - vx.getT(hit.getPosition());  
54     const double u = dt / sigma;  
55     return estimator->getRho(u) * hit.getW();  
56 }  
57  
58  
59
```

- Terminate when fractional improvement becomes negligible (default 1e-4)

## 10.7.2 Powell's Quadratically Convergent Method

Powell first discovered a direction set method that does produce  $N$  mutually conjugate directions. Here is how it goes: Initialize the set of directions  $\mathbf{u}_i$  to the basis vectors,

$$\mathbf{u}_i = \mathbf{e}_i \quad i = 0, \dots, N - 1 \quad (10.7.6)$$

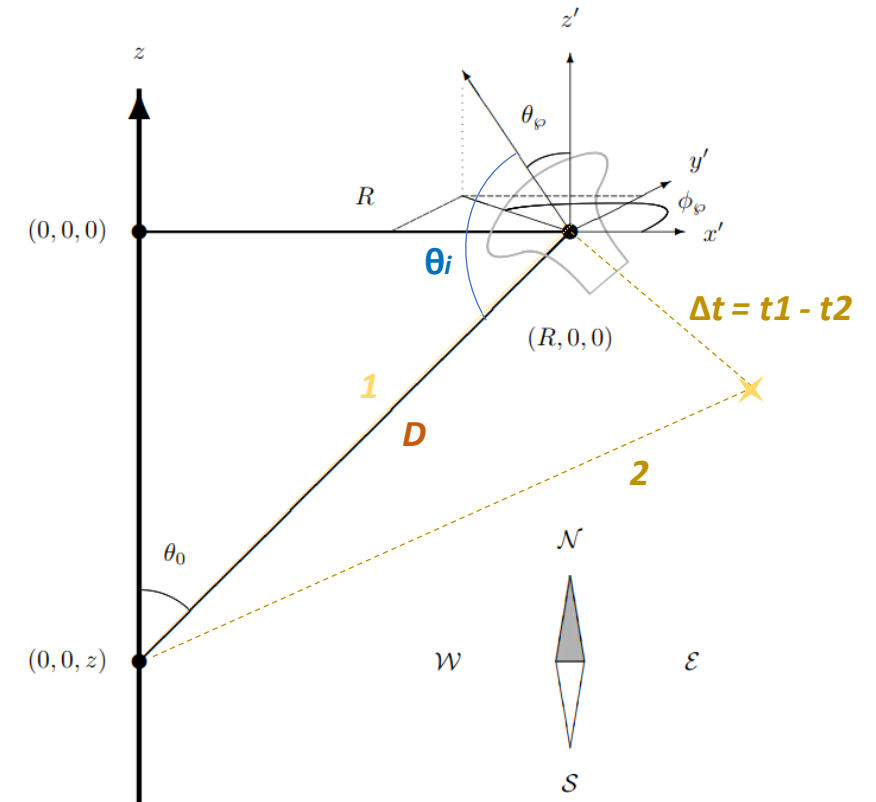
Now repeat the following sequence of steps ("basic procedure") until your function stops decreasing:

- Save your starting position as  $\mathbf{P}_0$ .
- For  $i = 0, \dots, N - 1$ , move  $\mathbf{P}_i$  to the minimum along direction  $\mathbf{u}_i$  and call this point  $\mathbf{P}_{i+1}$ .
- For  $i = 0, \dots, N - 2$ , set  $\mathbf{u}_i \leftarrow \mathbf{u}_{i+1}$ .
- Set  $\mathbf{u}_{N-1} \leftarrow \mathbf{P}_N - \mathbf{P}_0$ .
- Move  $\mathbf{P}_N$  to the minimum along direction  $\mathbf{u}_{N-1}$  and call this point  $\mathbf{P}_0$ .

Powell, in 1964, showed that, for a quadratic form like (10.7.1),  $k$  iterations of the above basic procedure produce a set of directions  $\mathbf{u}_i$  whose last  $k$  members are mutually conjugate. Therefore,  $N$  iterations of the basic procedure, amounting to

# JShowerPositionFit

- The fits from JPrefit and JSimplex should now be sufficiently close to the true optimum, to allow for a final, full MLE, including non-linear effects such as scattering
- JShowerPositionFit constitutes this final minimization step
  - Exploits the analytical arrival time PDFs, dependent on:
    - i.  $\Delta t$ : **the arrival time residual**
    - ii.  $D$ : the **(nu-vertex, PMT)-distance**
    - iii.  $\cos(\theta_i)$ : **cosine angle of incidence**
- The fit probability is defined as:
  - $P = E * \text{PDF}(D, \cos(\theta_i), \Delta t)$
- Minimization is based on the **Levenberg-Marquard Algorithm**
  - Requires analytical calculation of the Hessian matrix



JPDF documentation – Maarten de Jong (see [doxygen](#))

# JShowerPositionFit

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  - $P = E * \text{PDF}(D, \cos(\theta_i), \Delta t)$
- Minimization is based on the **Levenberg-Marquard Algorithm**
  - Requires analytical calculation of the Hessian matrix

```
JPDF_t::result_type JFIT::JRegressor< JPoint4D, JGandalf >::getH1 ( const double D,  
                                                                const double ct,  
                                                                const double t,  
                                                                const double E  
                                                                ) const
```

Get signal hypothesis value for bright point emission PDF.

#### Parameters

D hit distance from shower vertex [m]  
ct cosine of the HIT angle  
t arrival time of the light  
E shower energy [GeV]

#### Returns

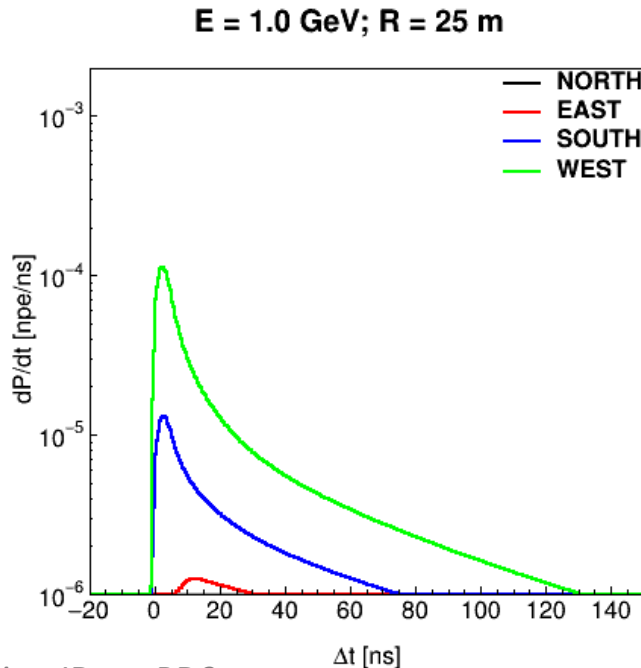
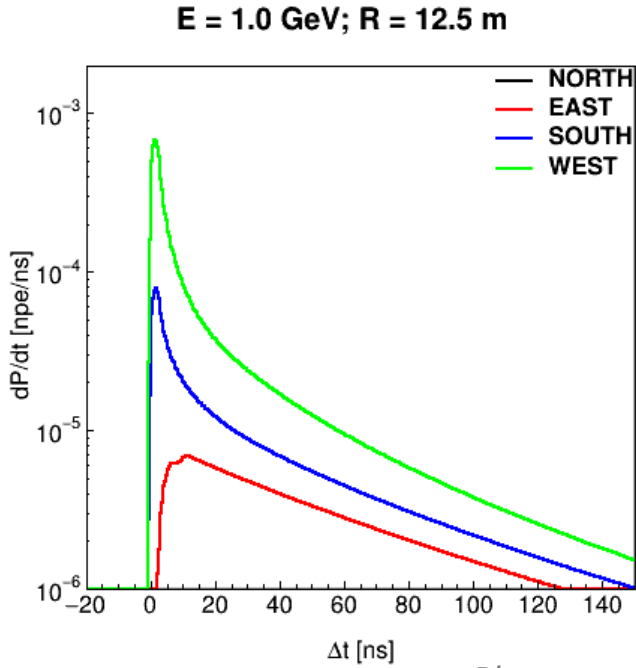
hypothesis value

Definition at line 200 of file JShowerBrightPointRegressor.hh.

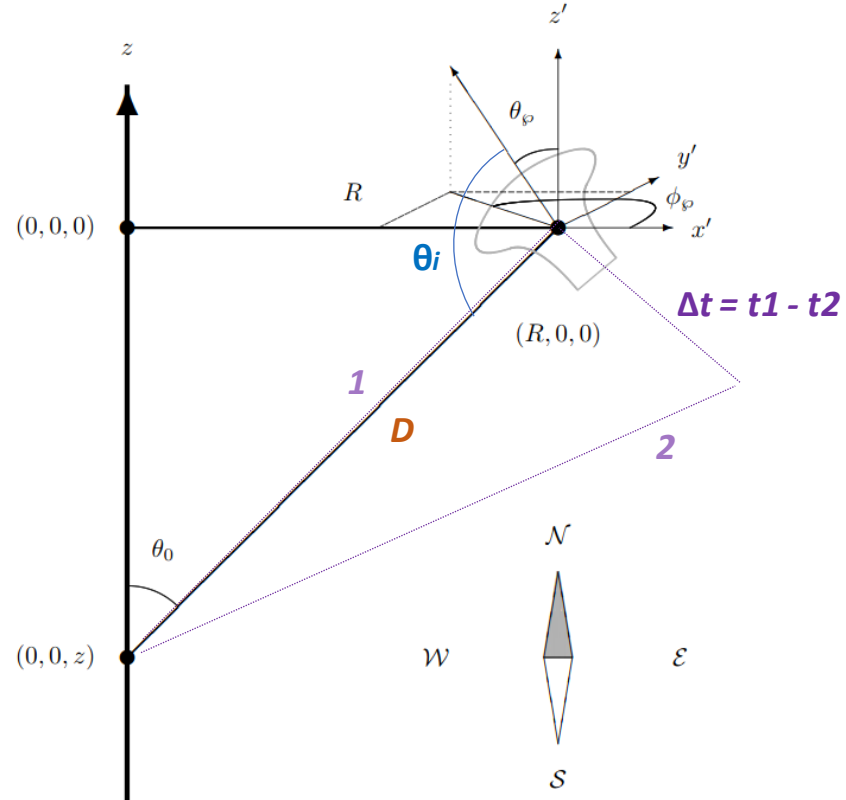
```
204 {  
205     using namespace JPP;  
206     JPDF_t::result_type h1 = JMATH::zero;  
207     for (int i = 0; i != NUMBER_OF_PDFS; ++i) {  
208         if (!pdf[i].empty() && D <= pdf[i].getXmax()) {  
209             try {  
210                 JPDF_t::result_type y1 = E * pdf[i](std::max(D, pdf[i].getXmin()), ct, t);  
211                 if (get_value(y1) > 0.0) {  
212                     h1 += y1;  
213                 }  
214             }  
215             catch(JLANG::JException& error) {  
216                 ERROR(error << std::endl);  
217             }  
218         }  
219     }  
220     return h1;  
221 }  
222 }  
223 }  
224 }  
225 }  
226 }  
227 }  
228 }  
229 }
```

# JShowerPositionFit

- The fits from JPrefit and JSimplex should now be sufficiently close to the true optimum, to allow for a final, full MLE, including non-linear effects such as scattering
- JShowerPositionFit constitutes this final minimization step



Plots created using JDrawPDO



JPDF documentation – Maarten de Jong (see [doxygen](#))

# JShowerEnergyPrefit

- Similar in working to JMuonEnergy
  - Energy estimation based on a binomial hit/no-hit likelihood estimation
- Uses the  $J_{pp}$  isotropic emission PDFs to calculate the hit probabilities, given a PMT at distance  $D$  from the vertex, with photon incidence angle  $\theta_i$ , originating from a shower with total energy  $E$
- An iterative five-point search is performed between a given energy range
  - In each iteration, pick the point with maximal hit/no-hit likelihood and refine the five points

# JShowerEnergyPrefit

- Similar in working to JMuonEnergy
  - Energy estimation based on a binomial hit/no-hit likelihood
- Uses the Jpp isotropic emission PDFs to calculate the hit probability at distance D from the vertex, with photon incidence angle  $\theta_i$ ,
- An iterative five-point search is performed between a given energy range
  - In each iteration, pick the point with maximal hit/no-hit likelihood and refine the five points

```
// 5-point search between given limits
const int N = 5;
JResult result[N];
for (int i = 0; i != N; ++i) {
    result[i].x = log10(Emin_GeV + i * (Emax_GeV - Emin_GeV) / (N-1));
}
do{
    int j = 0;
    for (int i = 0; i != N; ++i) {
        if (!result[i]) {
            result[i].chi2 = (*this)(result[i].x, data.begin(), data.end());
        }
        if (result[i].chi2 < result[j].chi2) {
            j = i;
        }
    }
    // squeeze range
    switch (j) {
    case 0:
        result[4] = result[1];
        result[2] = JResult(0.5 * (result[0].x + result[4].x));
        break;
    case 1:
        result[4] = result[2];
        result[2] = result[1];
        break;
    case 2:
        result[0] = result[1];
        result[4] = result[3];
        break;
    case 3:
        result[0] = result[2];
        result[2] = result[3];
        break;
    case 4:
        result[0] = result[3];
        result[2] = JResult(0.5 * (result[0].x + result[4].x));
        break;
    }
    result[1] = JResult(0.5 * (result[0].x + result[2].x));
    result[3] = JResult(0.5 * (result[2].x + result[4].x));
} while (result[4].x - result[0].x > resolution);
```

# JShowerDirectionPrefit

- Similar in working to JMuonEnergy
  - Energy estimation based on a binomial hit/no-hit likelihood
- Uses the Jpp isotropic emission PDFs to calculate the hit probability at distance D from the vertex, with photon incidence angle  $\theta_i$ ,
- An iterative five-point search is performed between a given energy range
  - In each iteration, pick the point with maximal hit/no-hit likelihood and refine the five points

```
// 5-point search between given limits
const int N = 5;
JResult result[N];
for (int i = 0; i != N; ++i) {
    result[i].x = log10(Emin_GeV + i * (Emax_GeV - Emin_GeV) / (N-1));
}
do{
    int j = 0;
    for (int i = 0; i != N; ++i) {
        if (!result[i]) {
            result[i].chi2 = (*this)(result[i].x, data.begin(), data.end());
        }
        if (result[i].chi2 < result[j].chi2) {
            j = i;
        }
    }
    // squeeze range
    switch (j) {
    case 0:
        result[4] = result[1];
        result[2] = JResult(0.5 * (result[0].x + result[4].x));
        break;
    case 1:
        result[4] = result[2];
        result[2] = result[1];
        break;
    case 2:
        result[0] = result[1];
        result[4] = result[3];
        break;
    case 3:
        result[0] = result[2];
        result[2] = result[3];
        break;
    case 4:
        result[0] = result[3];
        result[2] = JResult(0.5 * (result[0].x + result[4].x));
        break;
    }
    result[1] = JResult(0.5 * (result[0].x + result[2].x));
    result[3] = JResult(0.5 * (result[2].x + result[4].x));
} while (result[4].x - result[0].x > resolution);
```

**Some slides by Brían on direction fit**



**Bonus**

# ToT vs number of hits

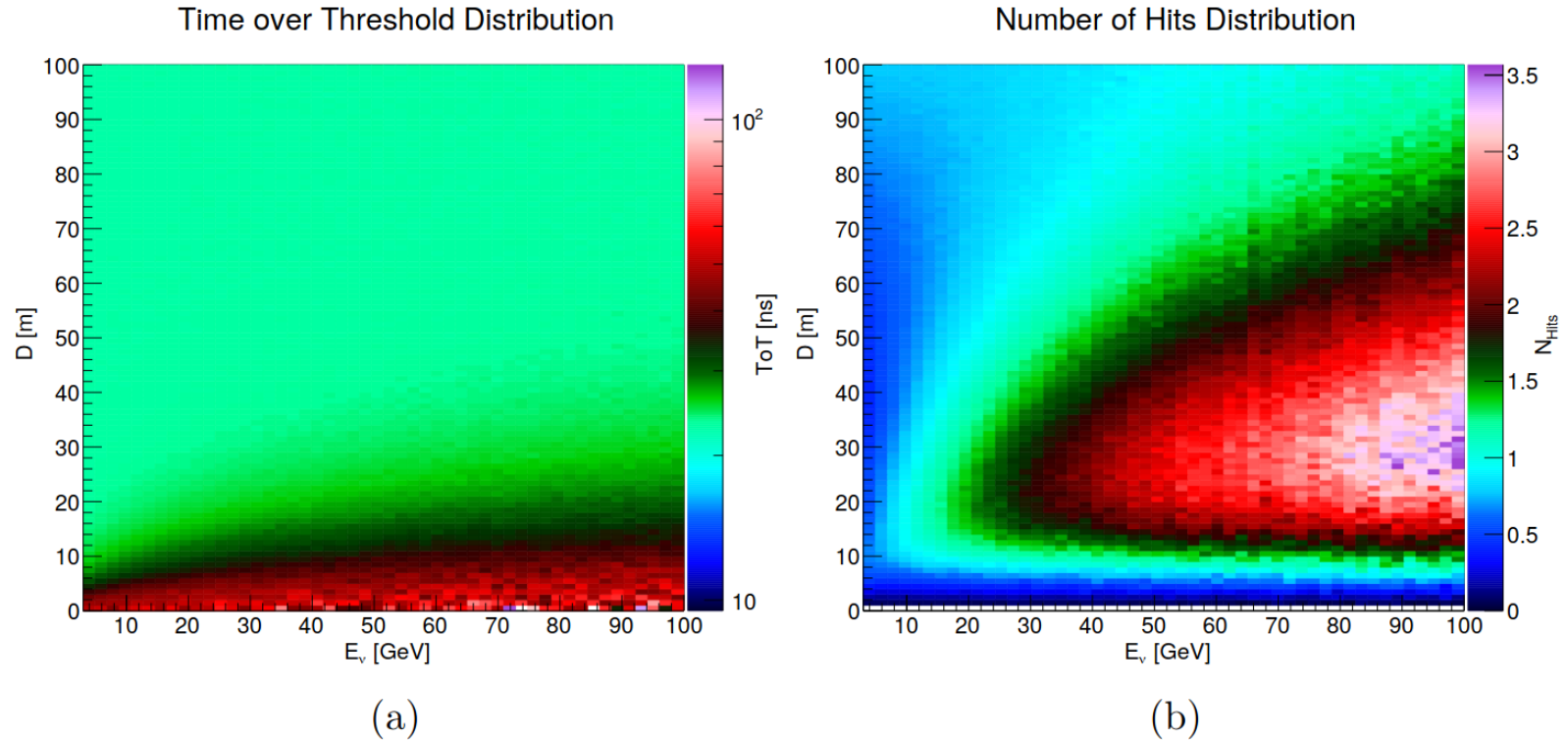


Figure 5.5: (a) ToT and (b) number of hits distributions in function of neutrino energy and distance from the neutrino interaction vertex.

# JPrefit – Mathematical foundation

- Define the difference between two hit arrival times:

$$t_j'^2 - t_i'^2 - 2(t_j' - t_i')t_0' = x_j^2 - x_i^2 - 2(x_j - x_i)x_0 + y_j^2 - y_i^2 - 2(y_j - y_i)y_0.$$

- Casting this into matrix form,  $H\Theta = Y$  :

$$H = \begin{pmatrix} 2(x_2 - x_1) & 2(y_2 - y_1) & -2(t_2' - t_1') \\ 2(x_3 - x_2) & 2(y_3 - y_2) & -2(t_3' - t_2') \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ 2(x_1 - x_n) & 2(y_1 - y_n) & -2(t_1' - t_n') \end{pmatrix}, \quad \Theta = \begin{pmatrix} x_0 \\ y_0 \\ t_0' \end{pmatrix} \longrightarrow Y = \begin{pmatrix} x_2^2 - x_1^2 + y_2^2 - y_1^2 - t_2'^2 + t_1'^2 \\ x_3^2 - x_2^2 + y_3^2 - y_2^2 - t_3'^2 + t_2'^2 \\ \cdot \\ \cdot \\ x_1^2 - x_n^2 + y_1^2 - y_n^2 - t_1'^2 + t_n'^2 \end{pmatrix},$$

- To which the least squares solution is:

$$\Theta = (H^T V^{-1} H)^{-1} \times H^T V^{-1} \times Y$$

$$V = J \times \begin{pmatrix} \sigma_1^2 & \cdot & \cdot & \cdot & 0 \\ \cdot & \sigma_2^2 & \cdot & \cdot & \cdot \\ \cdot & \cdot & \sigma_3^2 & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & \cdot & \cdot & \cdot & \sigma_n^2 \end{pmatrix} \times J^T, \quad J_{ij} = \delta Y_i / \delta t_j.$$

with V for the correlation matrix

## Vertex profit (fitting algorithm)

- Secondary grid scan of vertex hypotheses around the initial reconstructed vertex from step 1. (outcome of step 1. still very sensitive to inclusion of background hits)
  - From  $-20$  m to  $+20$  m in 3 steps for (x,y,z)
  - From  $-50$  ns to  $+50$  ns in 4 steps for (t)
- For each of these, make an L0-only hit cluster, with  $D < 80$  m and  $\Delta t < 100$  ns
- The resulting hit clusters are fitted using an analytical PDF (see [Janník's thesis](#)) which penalizes PMT hits not oriented towards the assumed vertex position:

