

An implement of Kalman filtering in MarlinTPC

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Equations

System equation:

$$\mathbf{x}_{k+1} = \Phi_k \mathbf{x}_k + \mathbf{w}_k$$

\mathbf{x}_k – system state vector at discrete time k

Φ_k – state transfer matrix

\mathbf{w}_k – process noise (e.g. multiple scattering)

Measurement equation:

$$\mathbf{z}_k = \mathbf{H}_k \mathbf{x}_k + \mathbf{v}_k$$

\mathbf{H}_k – measurement matrix (projection)

\mathbf{v}_k – measurement error

Covariance:

$$E[\mathbf{w}_i \mathbf{w}_j^T] = \mathbf{Q}_i \delta_{ij} \quad E[\mathbf{v}_i \mathbf{v}_j^T] = \mathbf{R}_i \delta_{ij} \quad E[\mathbf{w}_i \mathbf{v}_j^T] = 0$$

A simple deduction of Kalman Filter

- Prediction:

$$\hat{x}_k^- = \phi_{k-1} \hat{x}_{k-1}$$

$$\begin{aligned} P_k^- &= \text{cov}[\hat{x}_k - x_k] = \text{cov}[\phi_{k-1} \hat{x}_{k-1} - \phi_{k-1} x_{k-1} - w_{k-1}] \\ &= \text{cov}[\phi_{k-1} (\hat{x}_{k-1} - x_{k-1})] + \text{cov}[w_{k-1}] \\ &= \phi_{k-1} P_{k-1}^- \phi_{k-1}^T + Q_{k-1} \end{aligned}$$

- Filtering:

$$\begin{aligned} \text{Let } L(\hat{x}_k) &= (\hat{x}_k - \hat{x}_k^-)^T (P_k^-)^{-1} (\hat{x}_k - \hat{x}_k^-) + (z_k - H_k \hat{x}_k)^T R_k^{-1} (z_k - H_k \hat{x}_k) \\ &\equiv \min \end{aligned}$$

Then the least square estimator is,

$$\hat{x}_k = \hat{x}_k^- + K_k (z_k - H_k \hat{x}_k^-)$$

in which, $K_k = P_k^- H_k^T (H_k P_k^- H_k^T + R_k)^{-1}$ (Kalman gain matrix)

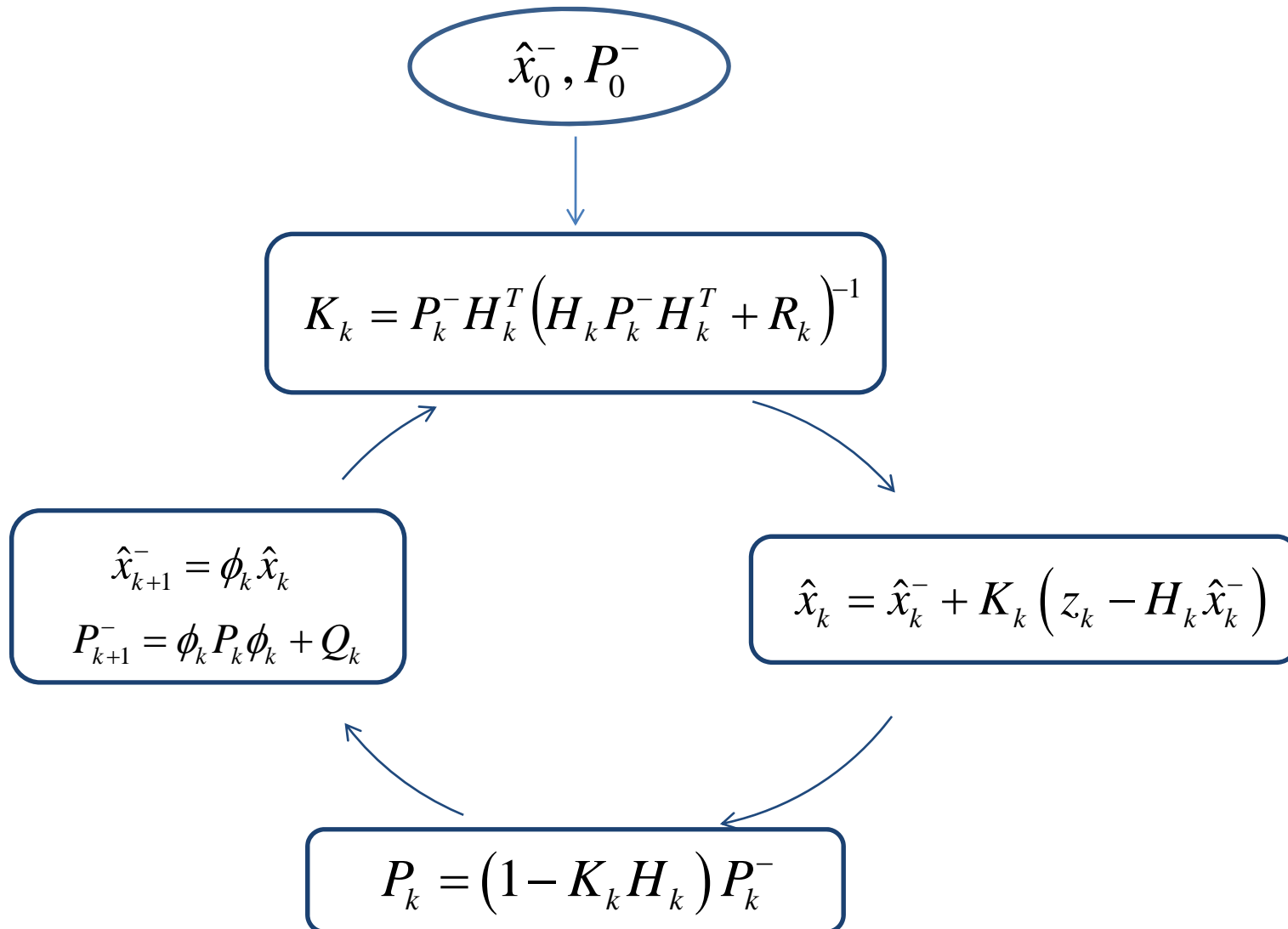
and $P_k = (1 - K_k H_k) P_k^-$

Filtered Residual $r_k = z_k - H_k \hat{x}_k = (1 - H_k K_k) (z_k - H_k \hat{x}_k^-)$

Its covariance $C_k = \text{cov}[\hat{r}_k] = \text{cov}[(1 - H_k K_k) (z_k - H_k \hat{x}_k^-)]$
 $= (1 - H_k K_k) \text{cov}[z_k - H_k \hat{x}_k^-]$
 $= (1 - H_k K_k) R_k$

χ^2 update $\chi_k^2 = \chi_{k-1}^2 + \chi_+^2 = \chi_{k-1}^2 + r_k^T C_k^{-1} r_k$

Kalman Filtering Loop



Straight model

The line equation is:

$$y_k = a_k x + b_k$$

re-writes it for kalman filtering:

$$X_k = \begin{bmatrix} a_k \\ b_k \end{bmatrix} \quad \phi_k = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad H_k = [k \quad 1]$$

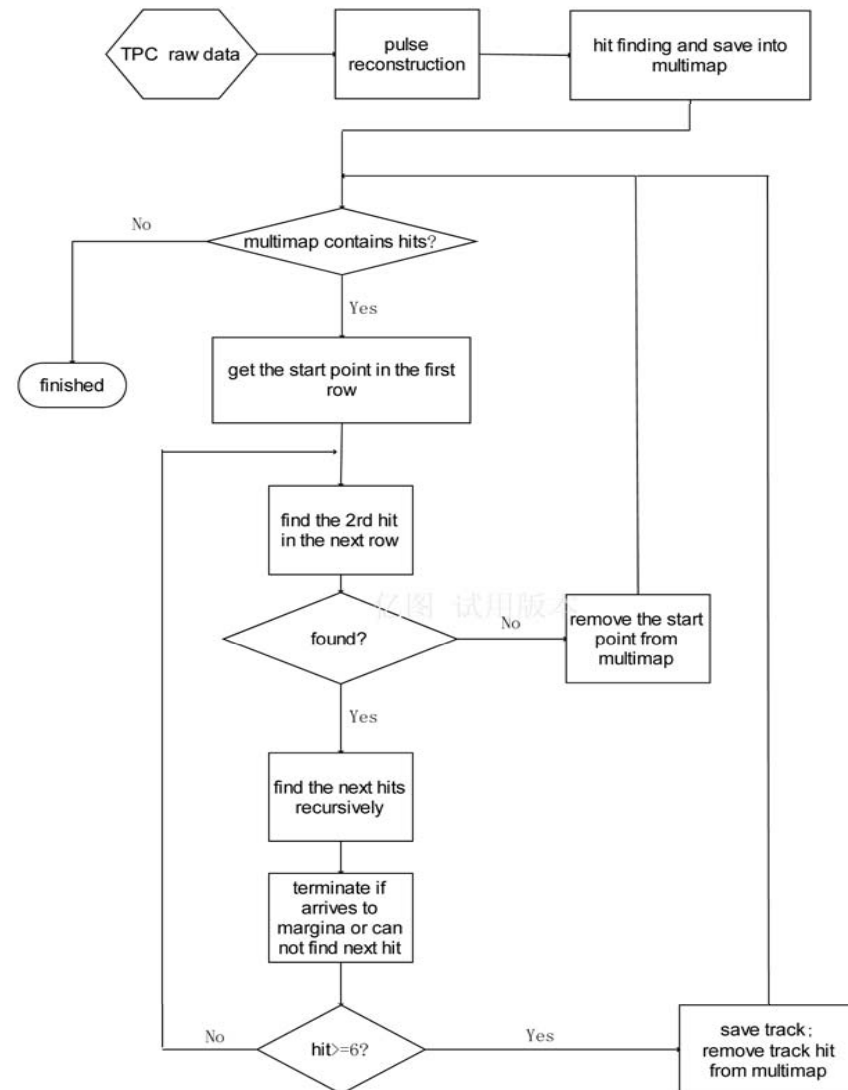
So both slope and intercept can be got.

In track following method in which kalman filter is used, track finding and track fitting are completely simultaneously. And the estimator

$$\hat{y}_k = \hat{a}_k x + \hat{b}_k$$

is used to discriminate wrong hit at every step.

Flowchart of Algorithm in MarlinTPC



Test condition

- The spatial resolution is not considered.
- Raw data is simulated by a C++ code
- Readout geometry is for TU-TPC(10 rows)
- Track definition: hit size \geq 6, MaxSkipRow=1.

Test 1

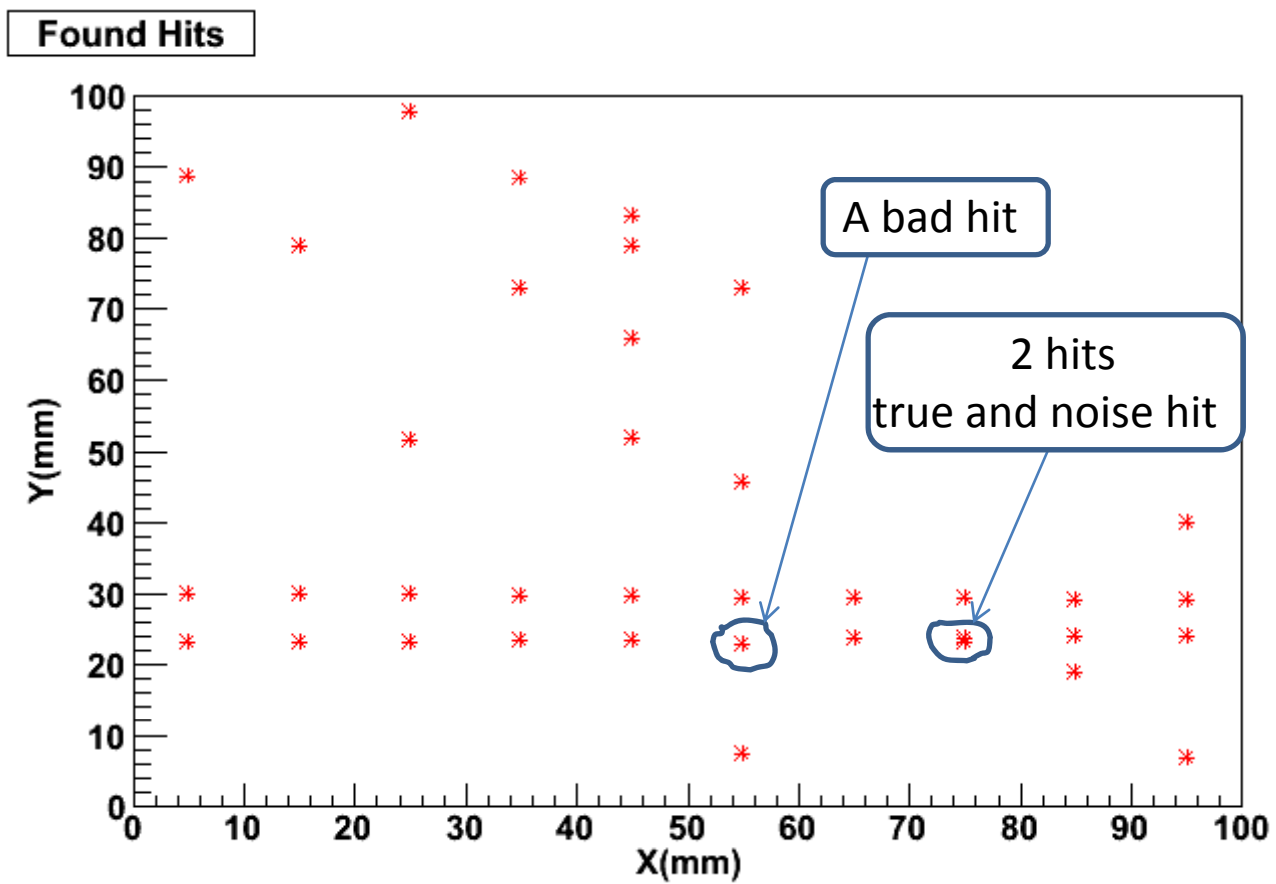
- Formula of two tracks:

$$y_1 = 0.01x + 23$$

$$y_2 = -0.01x + 30$$

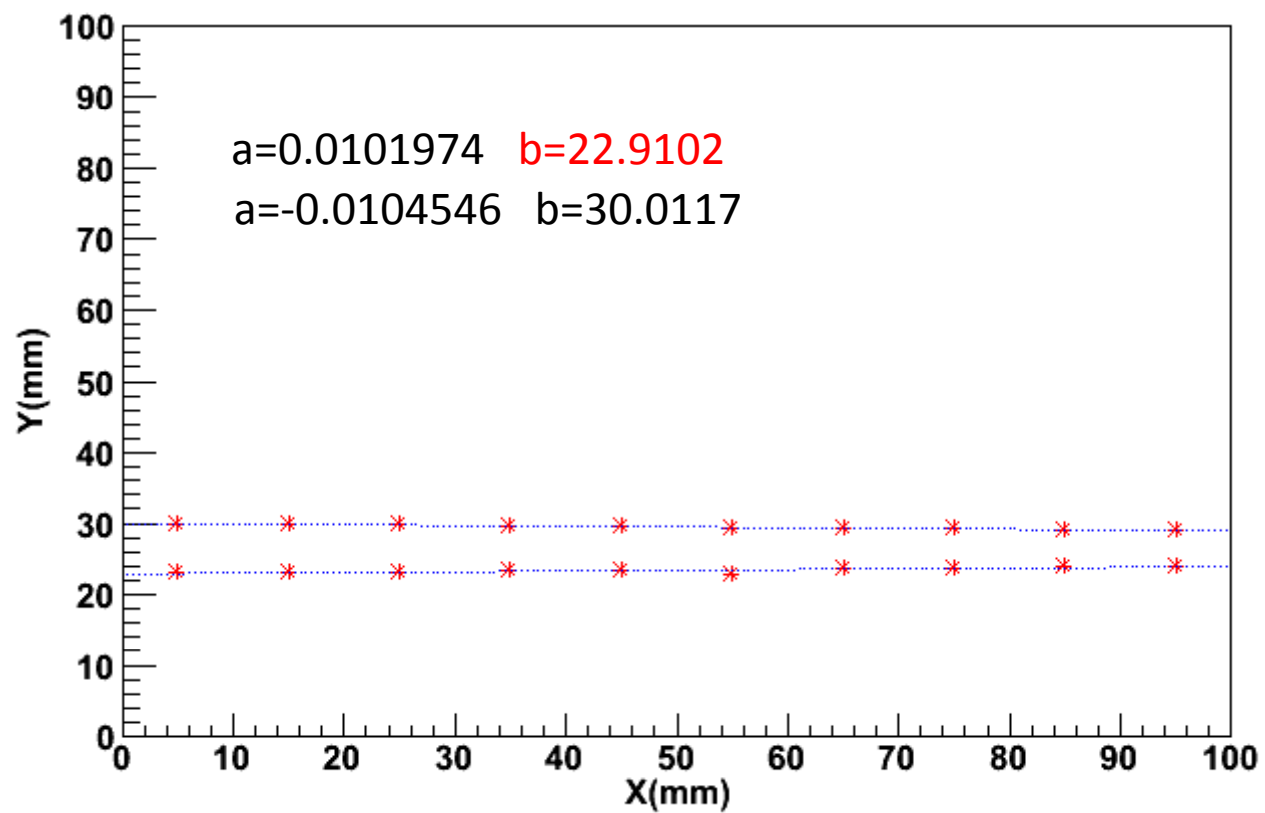
- 21 noise hits are added to the true hits.
- Discrimination value in the propagate step is 1.

Test 1-1



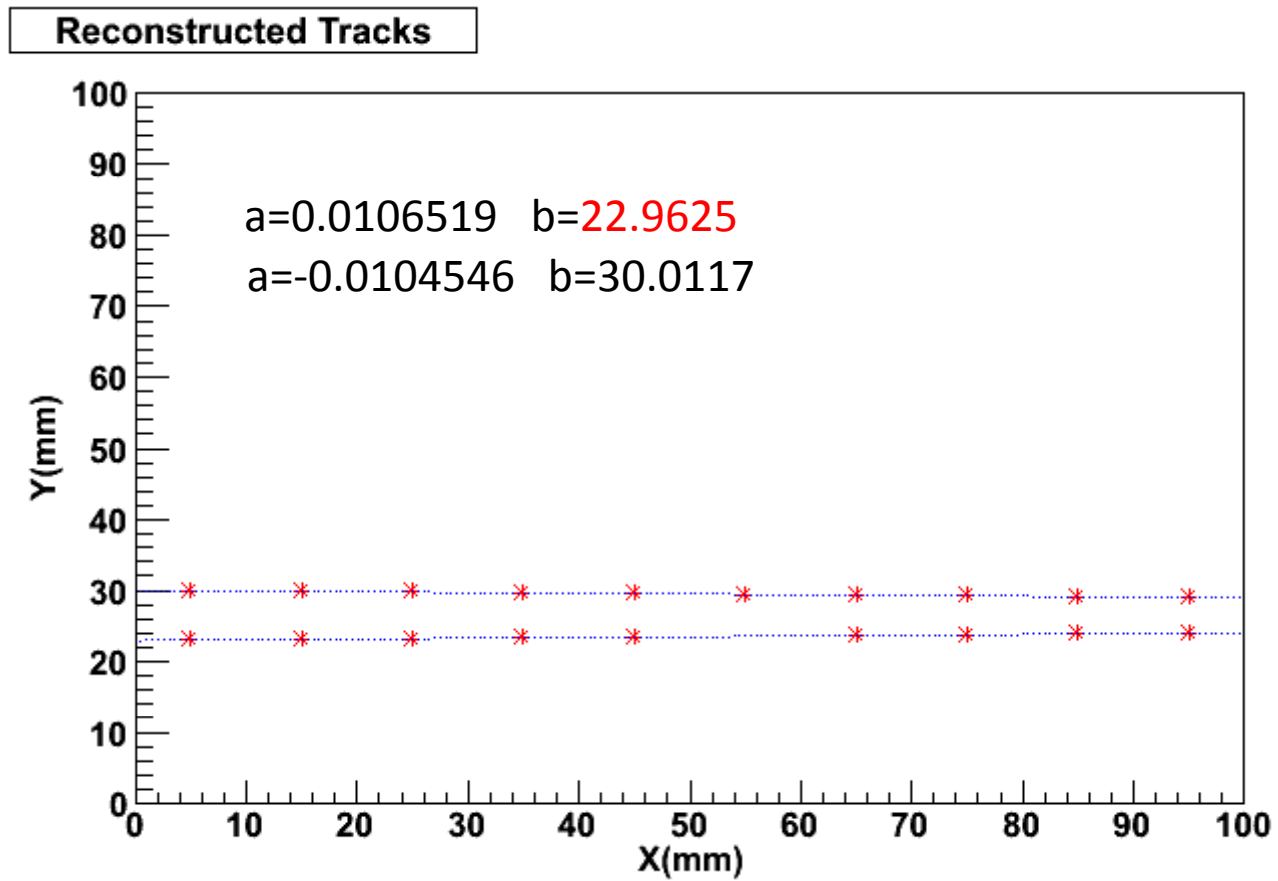
Test 1-1

Reconstructed Tracks



Test 1-2

Discrimination value is 0.1, other two conditions is the same as test I-I.

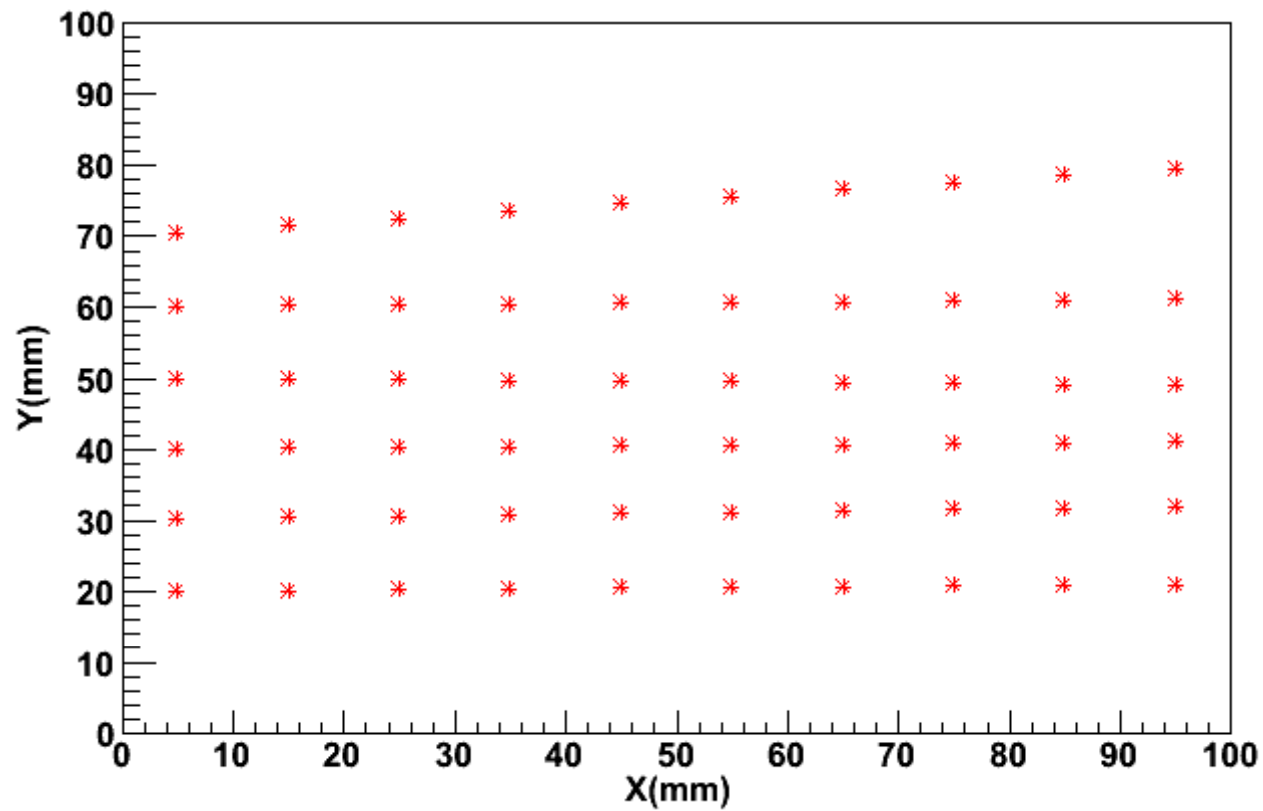


Conclusions from test 1

- The track finding code can have good capability to discriminate noise, if a proper value is chosen.
- Track fitting result is improved by abandoning bad hit.

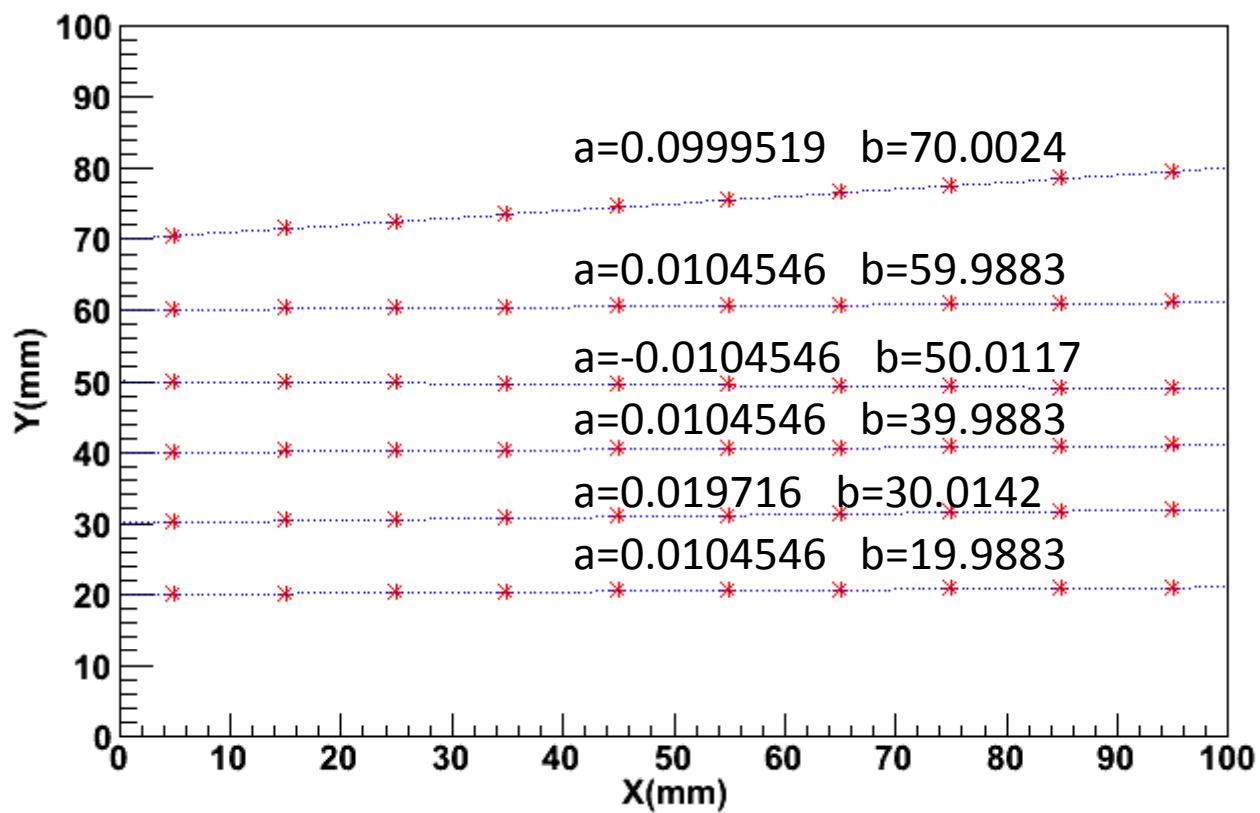
Test 2-1

Found Hits



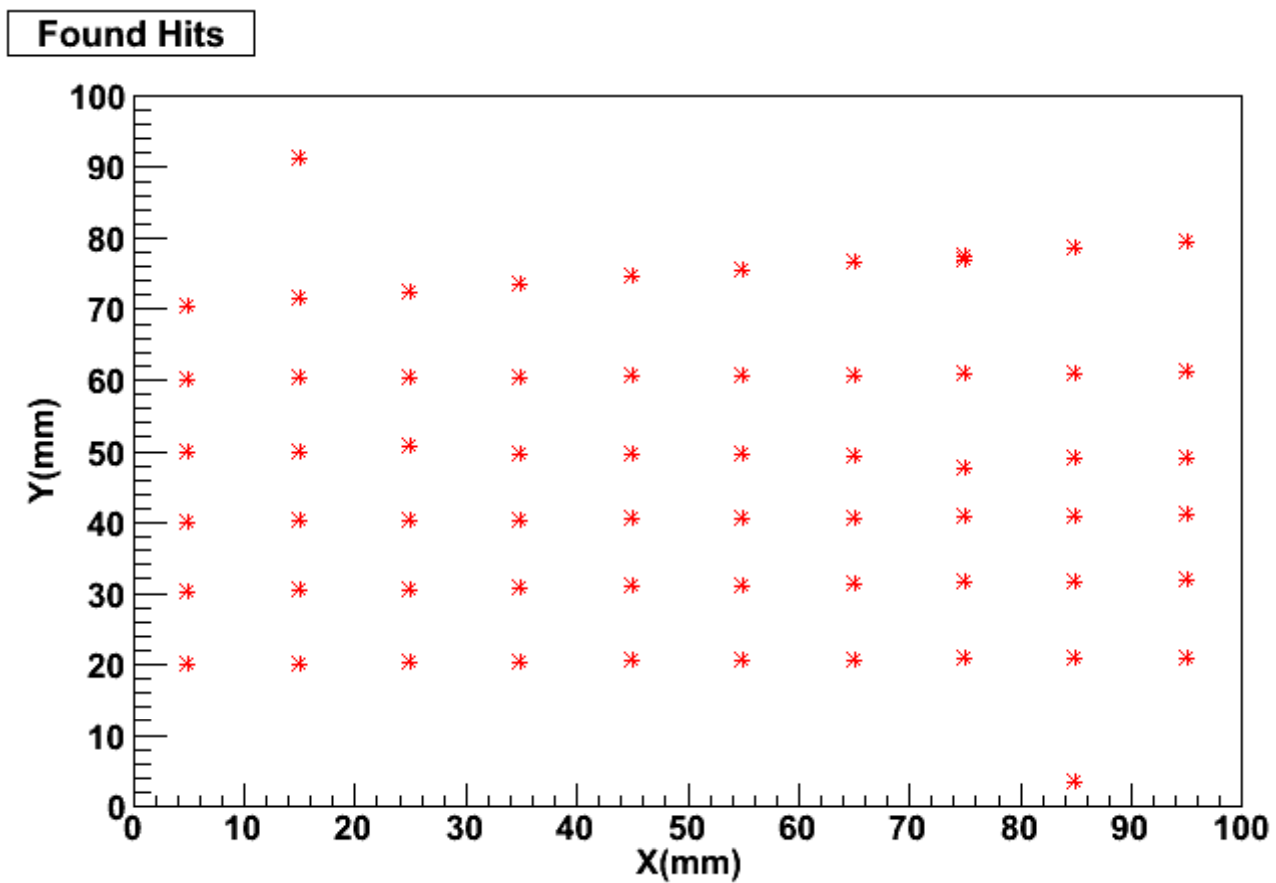
Test 2-1

Reconstructed Tracks



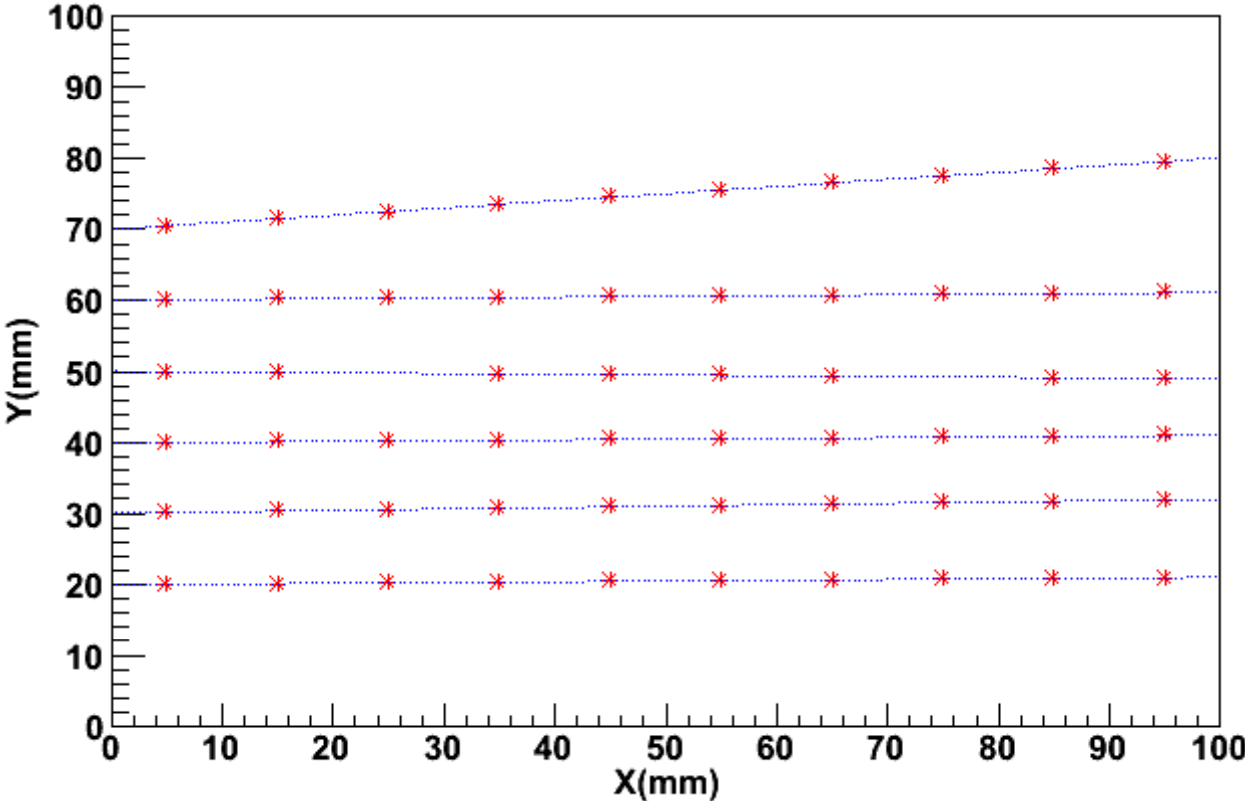
Test 2-2

Add 5 noise hits to test 2-1



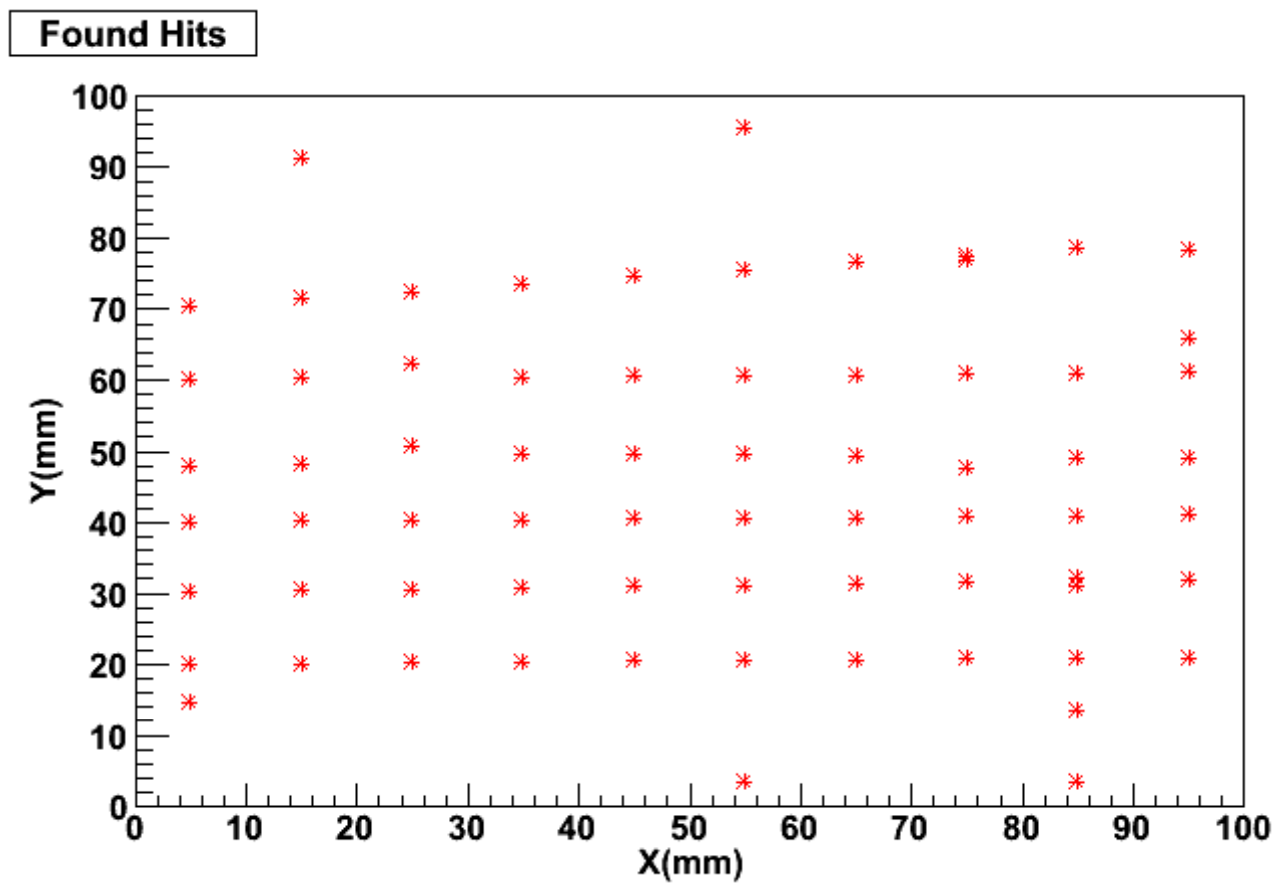
Test 2-2

Reconstructed Tracks



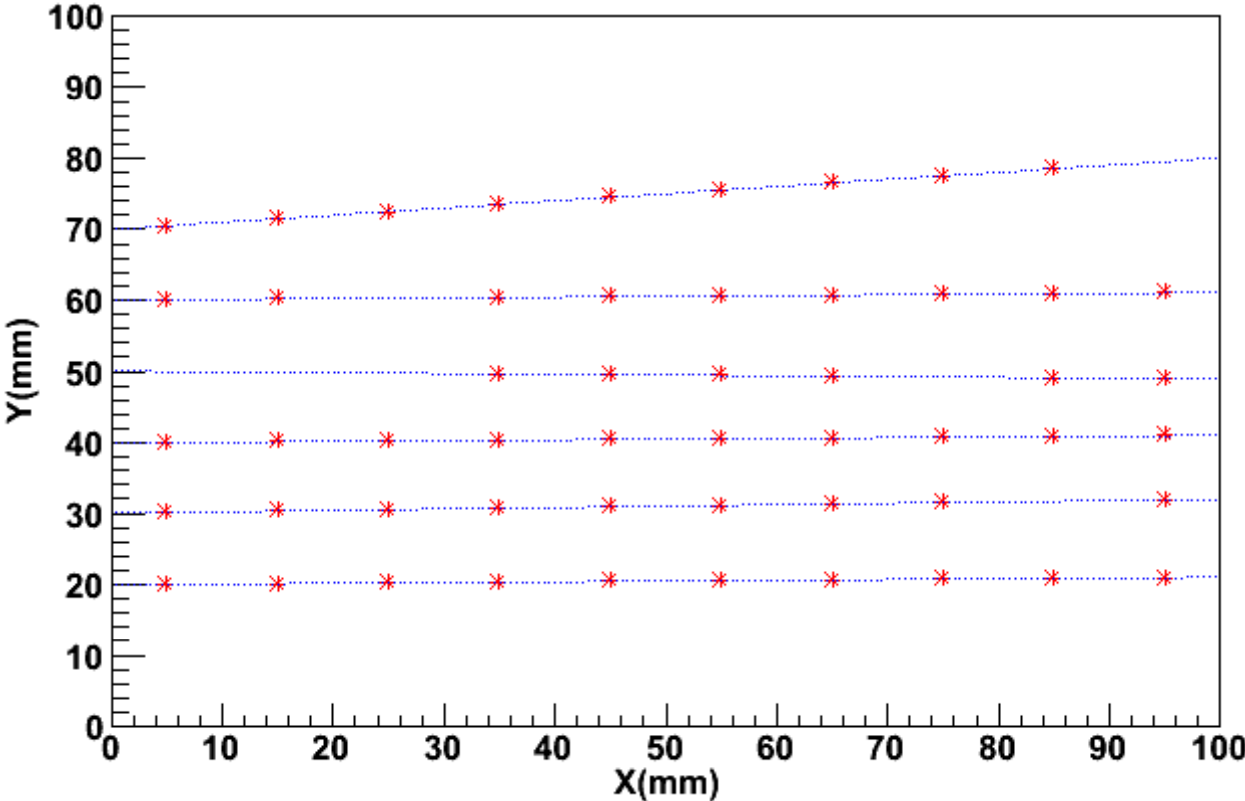
Test 2-3

Add 15 noise hits



Test 2-3

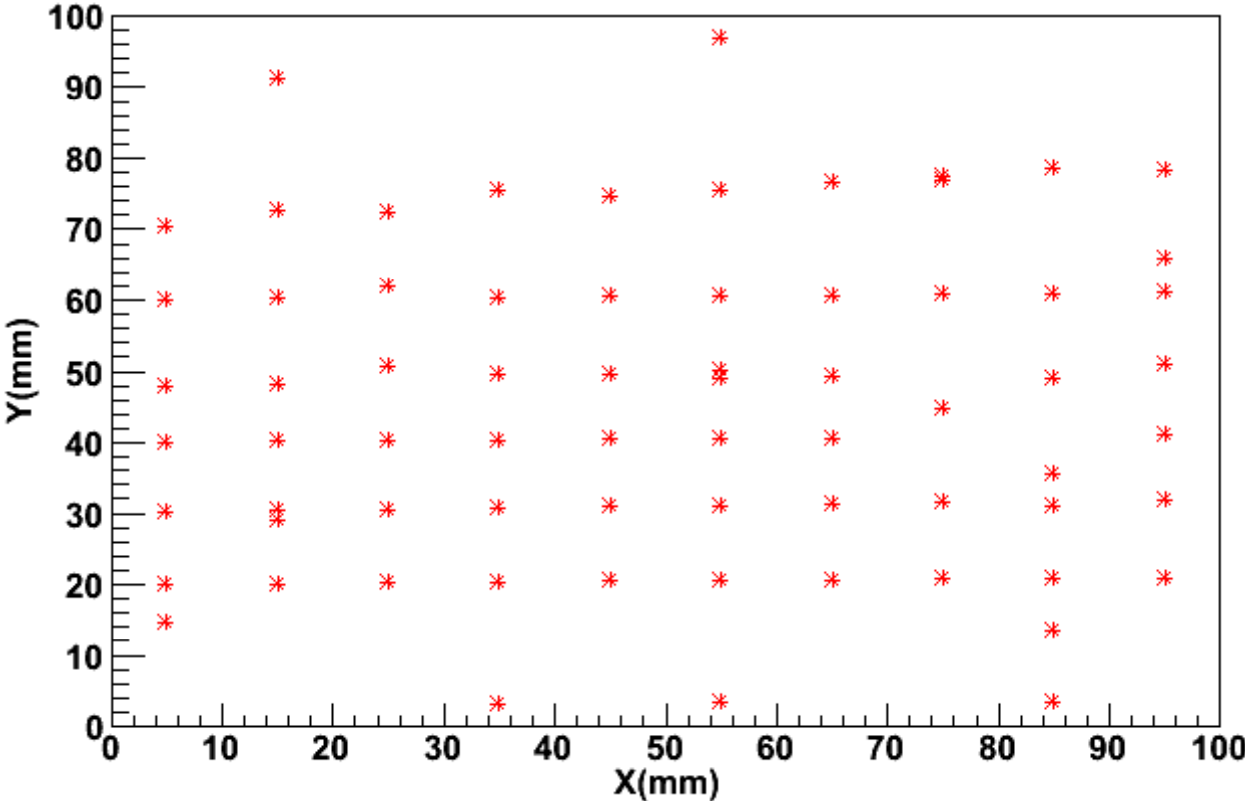
Reconstructed Tracks



Test 2-4

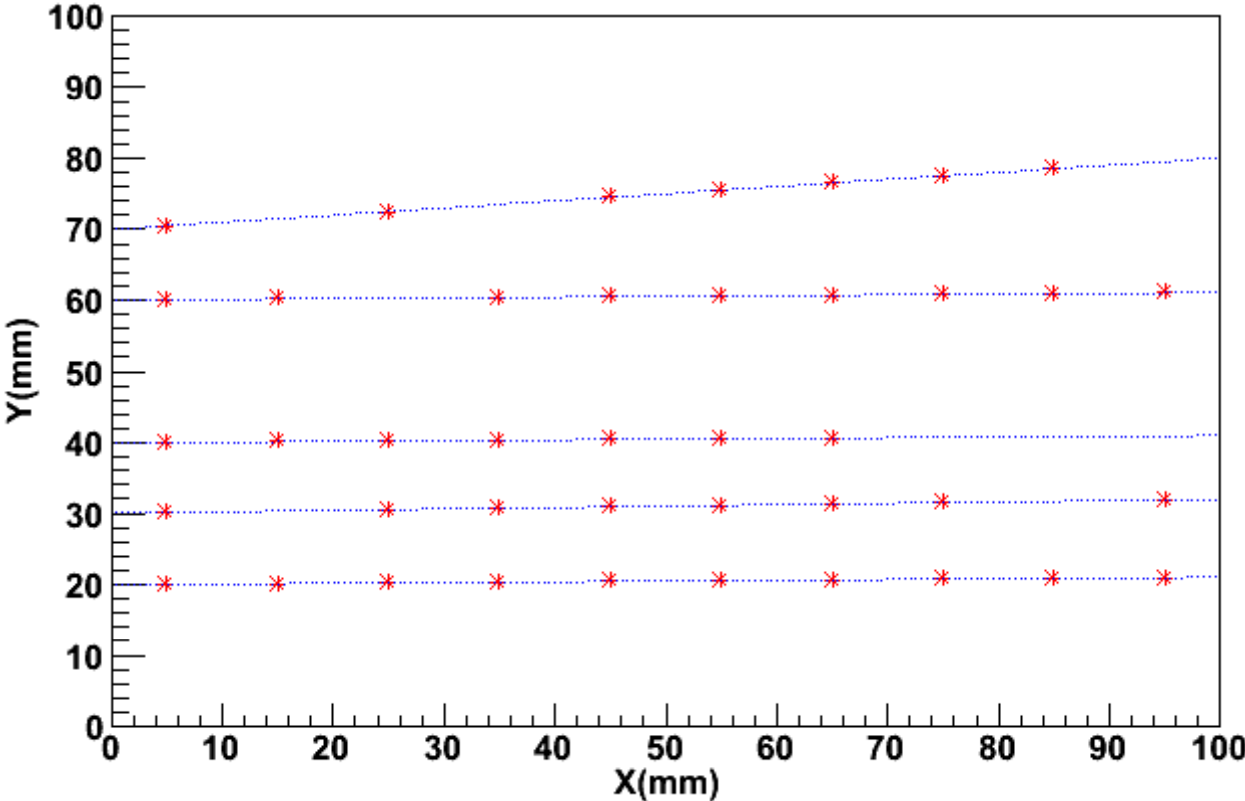
Add 25 noise hits

Found Hits



Test 2-4

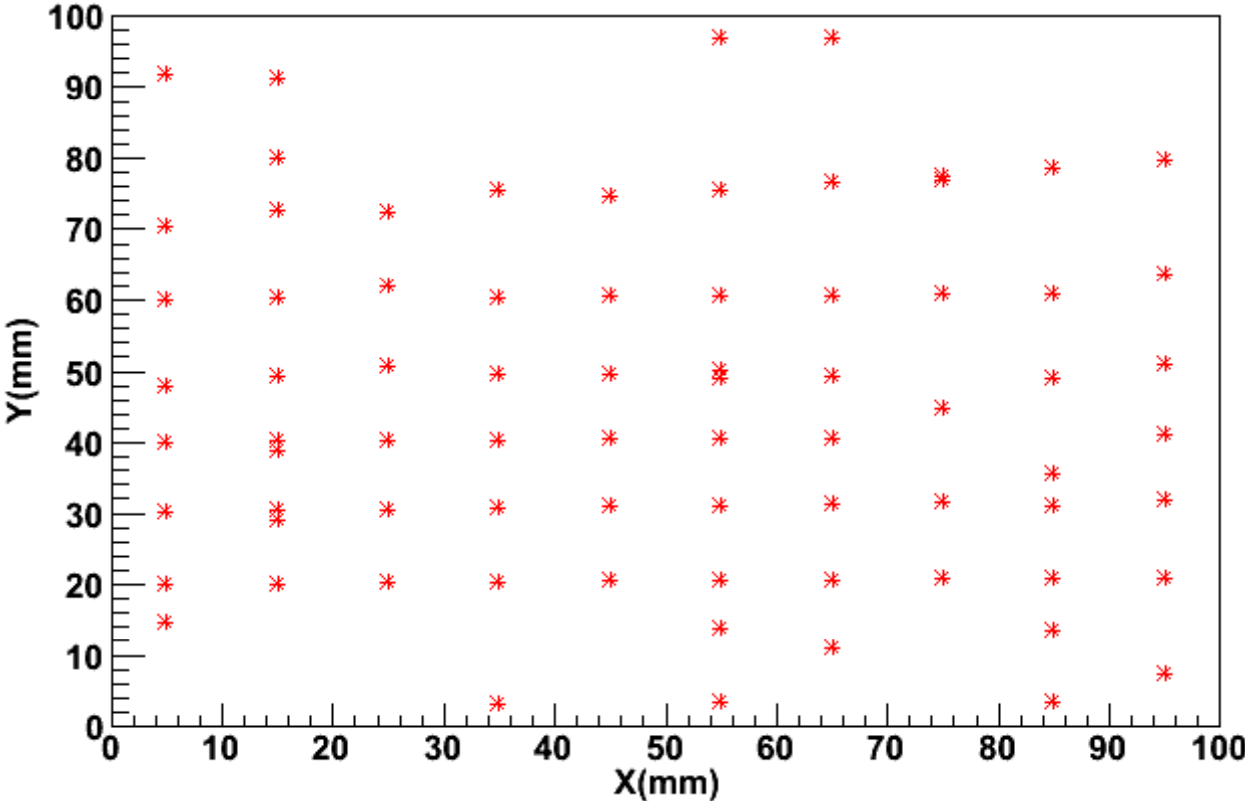
Reconstructed Tracks



Test 2-5

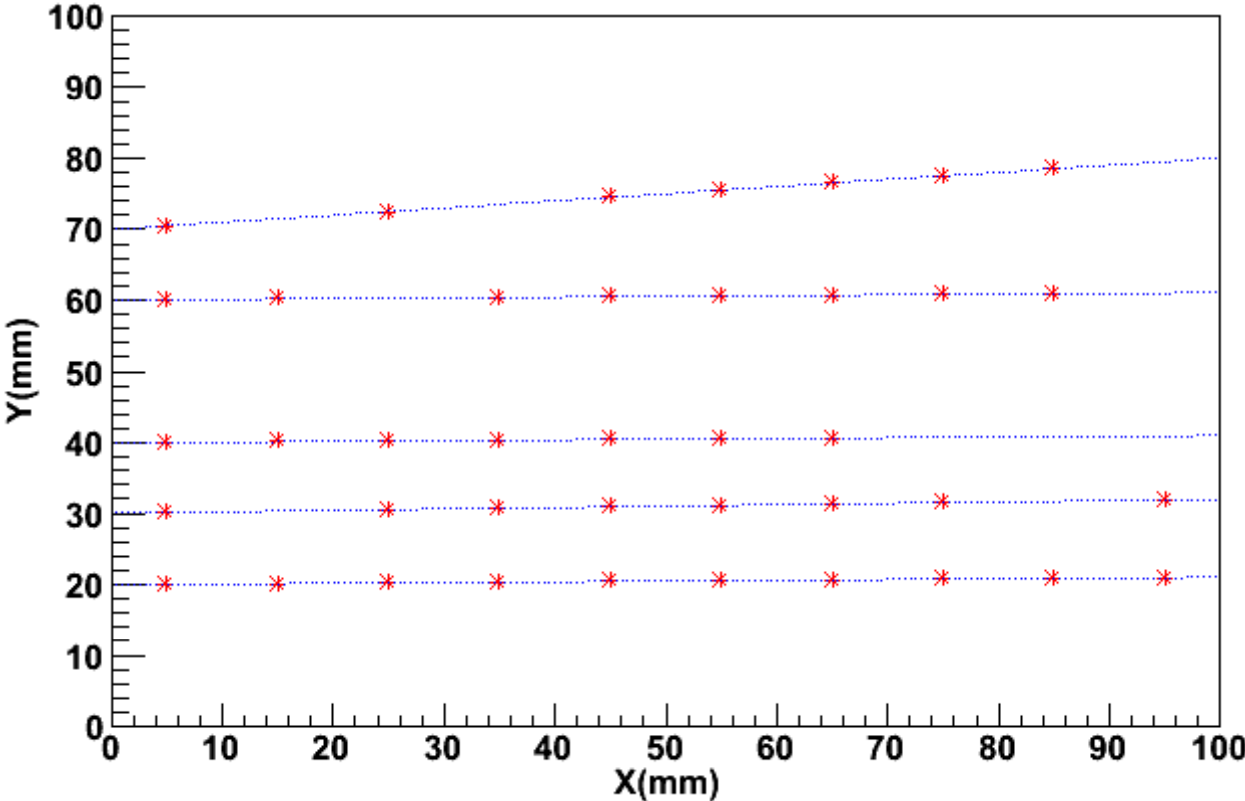
Add 35 noise hits

Found Hits



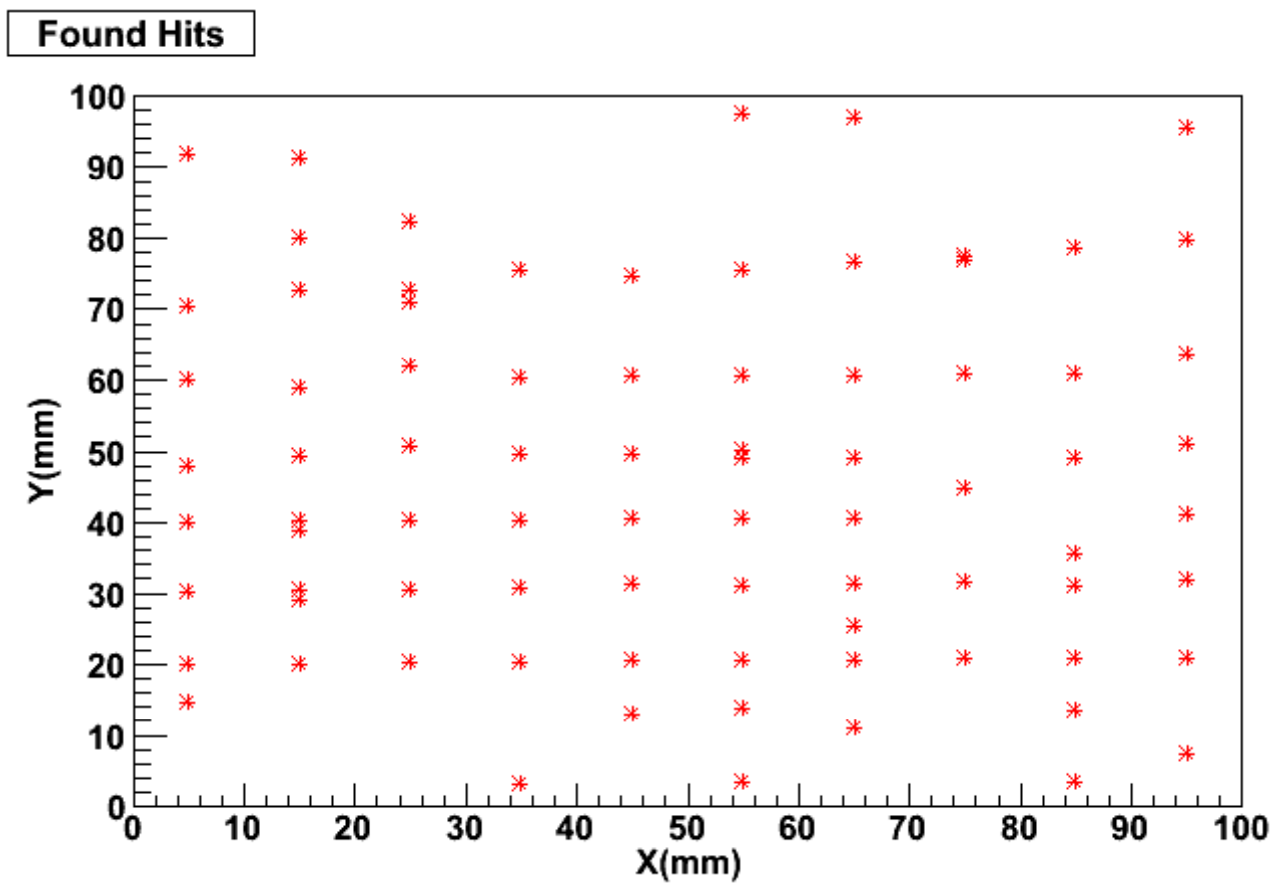
Test 2-5

Reconstructed Tracks



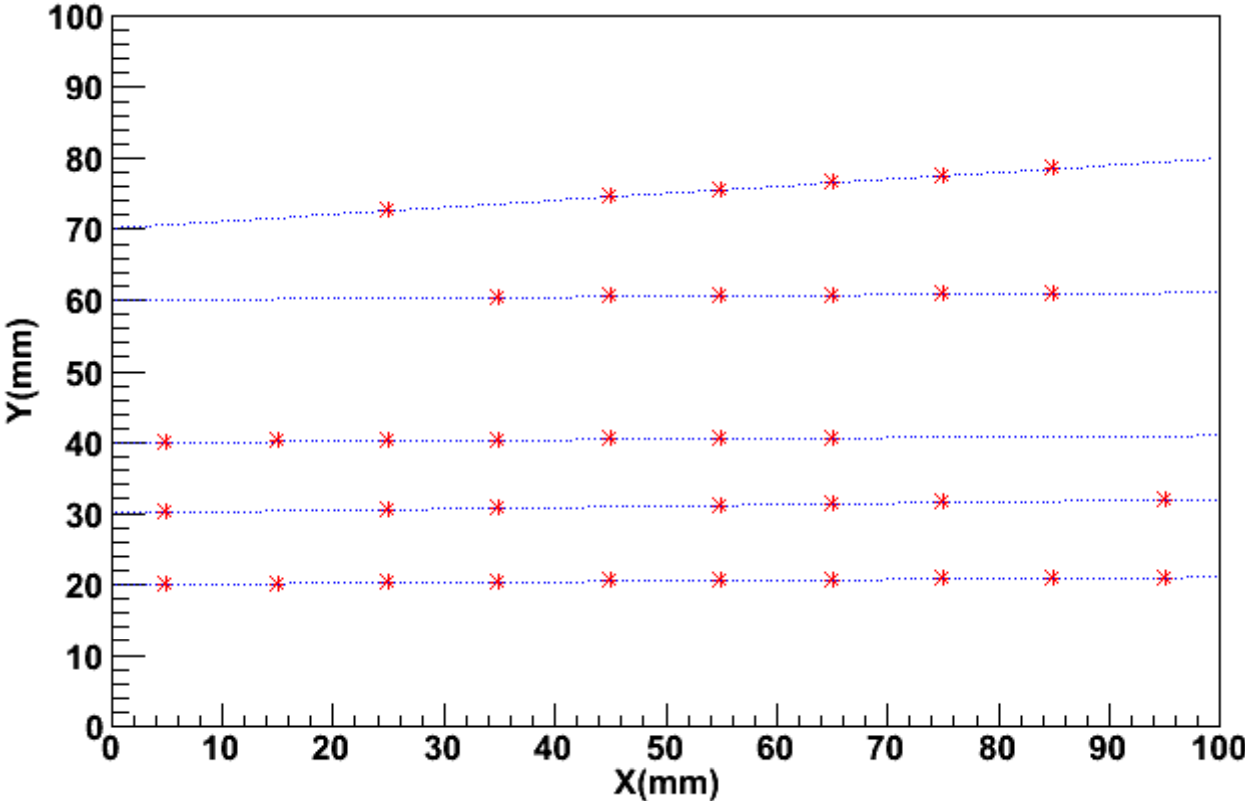
Test 2-6

Add 45 noise hits



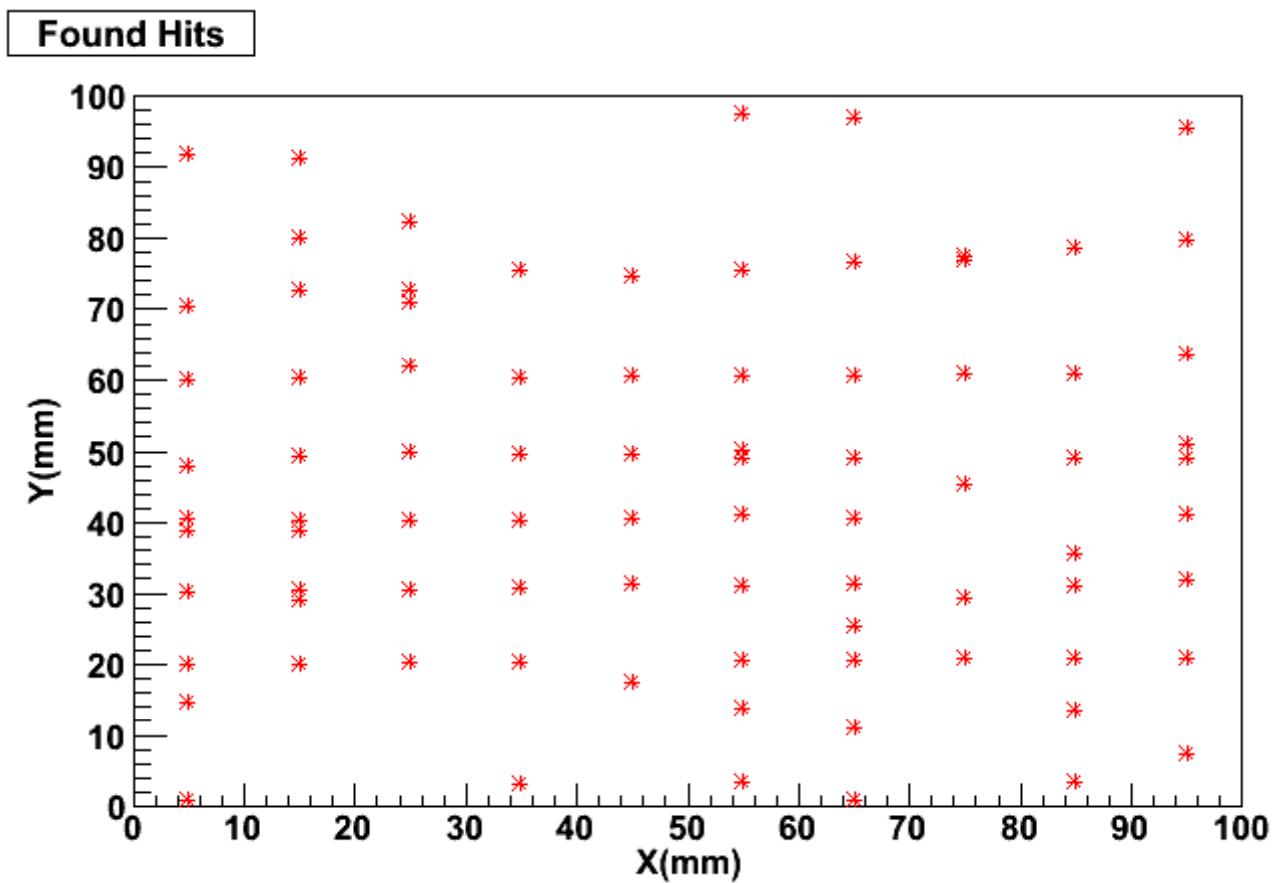
Test 2-6

Reconstructed Tracks



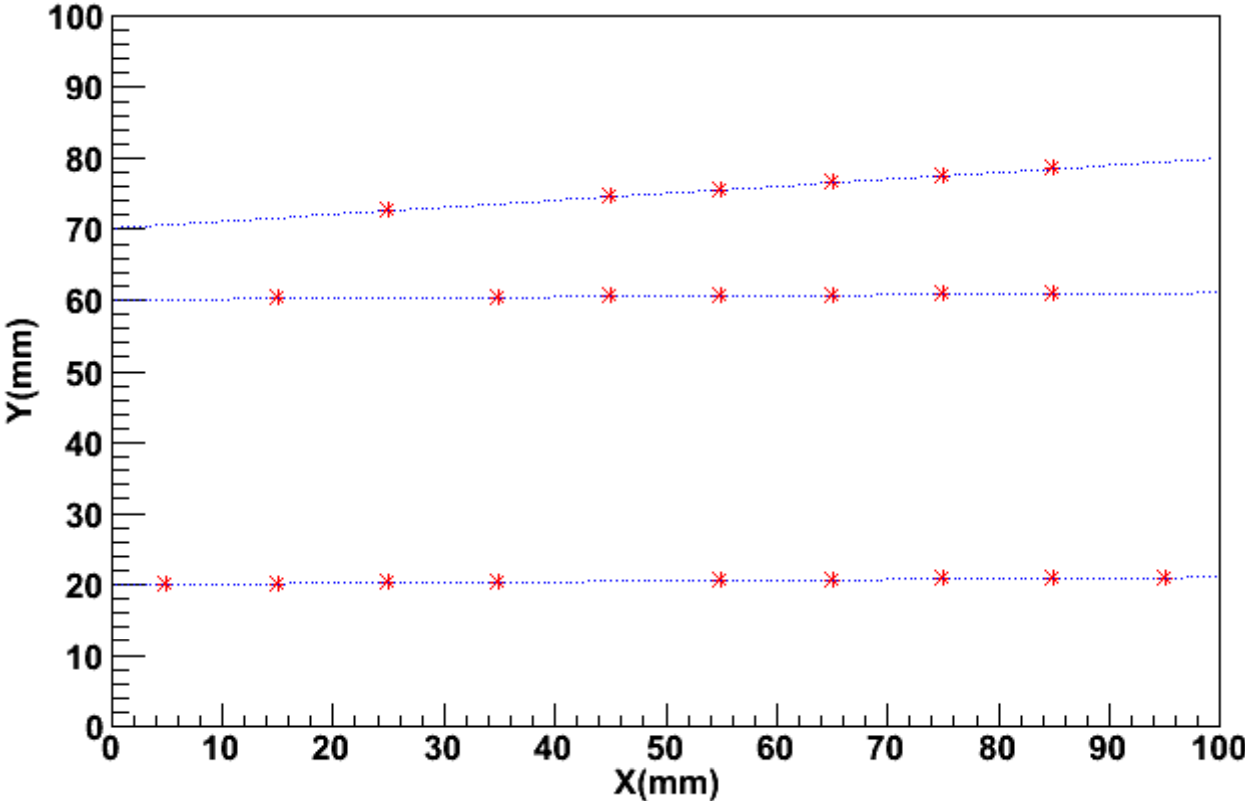
Test 2-7

Add 55 noise hits



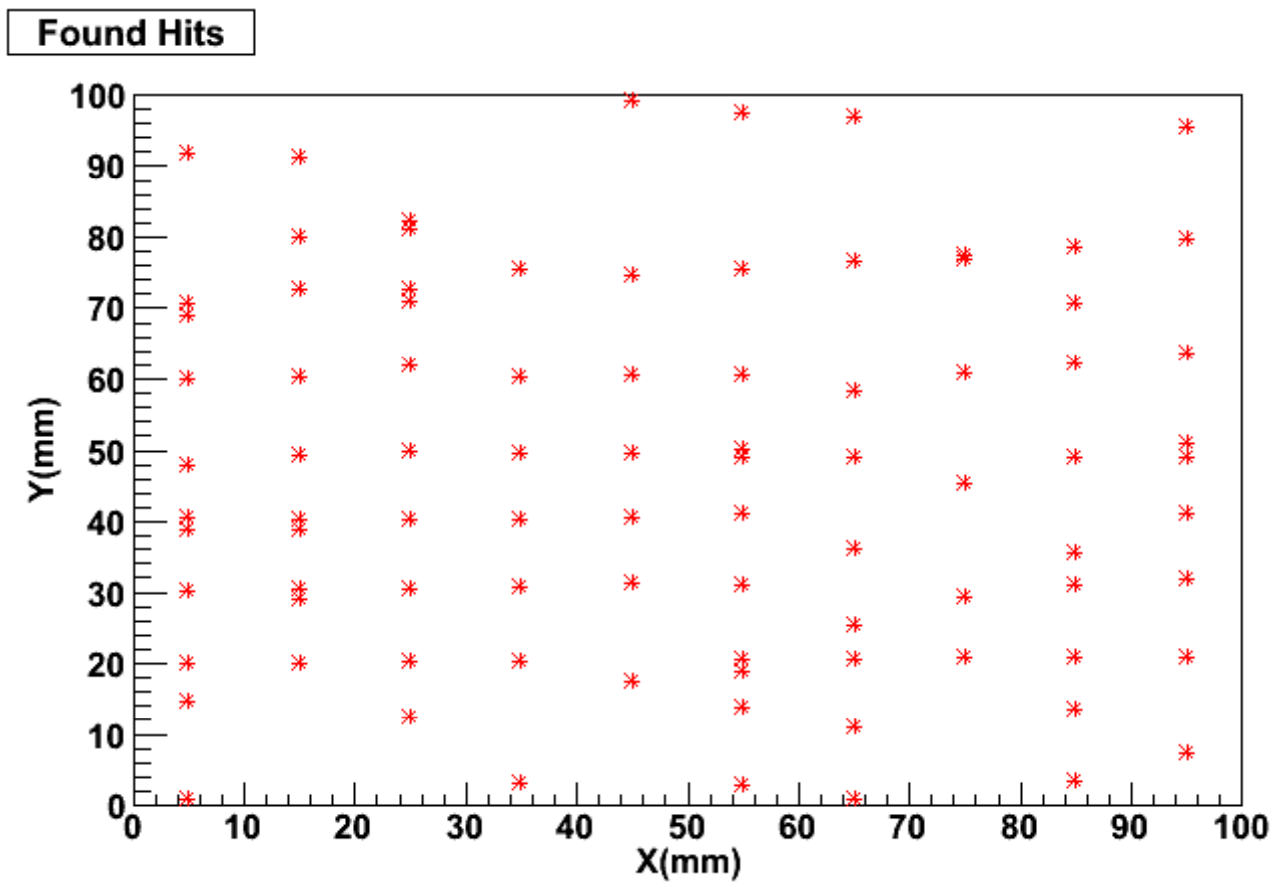
Test 2-7

Reconstructed Tracks



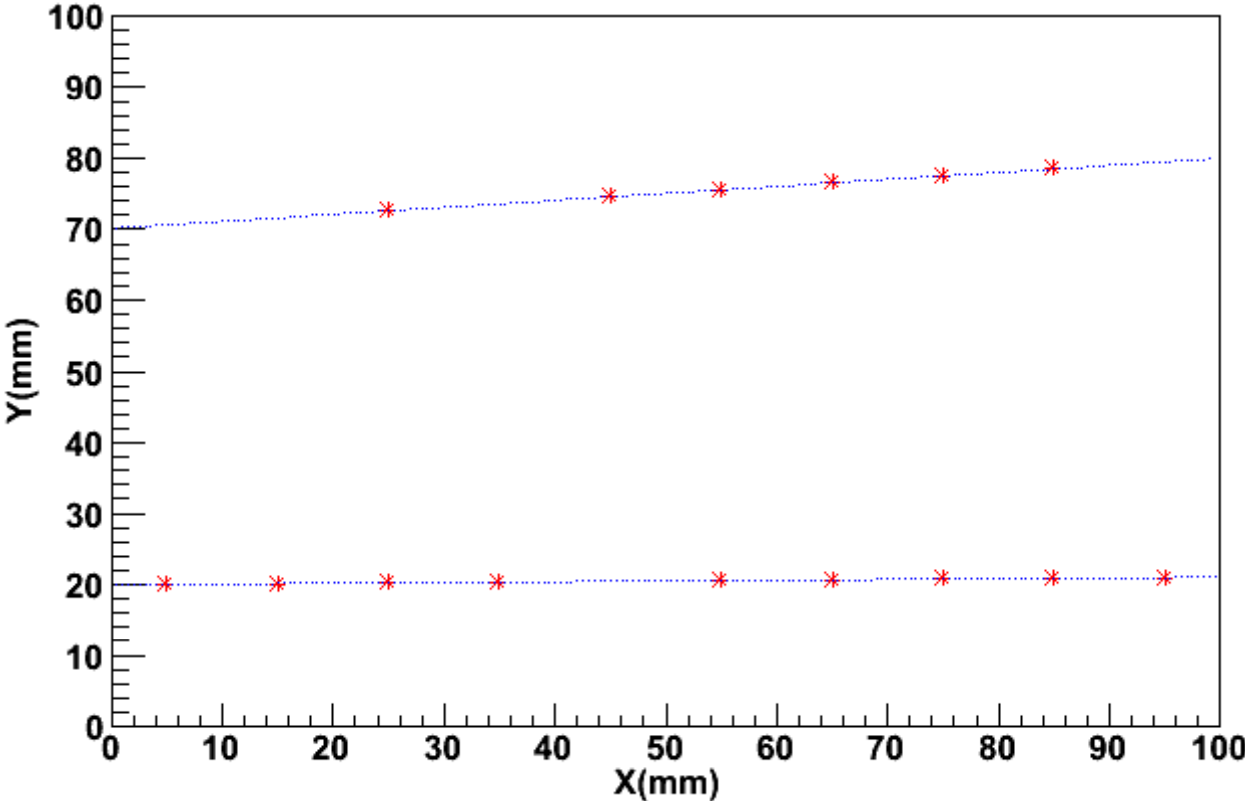
Test 2-8

Add 65 noise hits



Test 2-8

Reconstructed Tracks

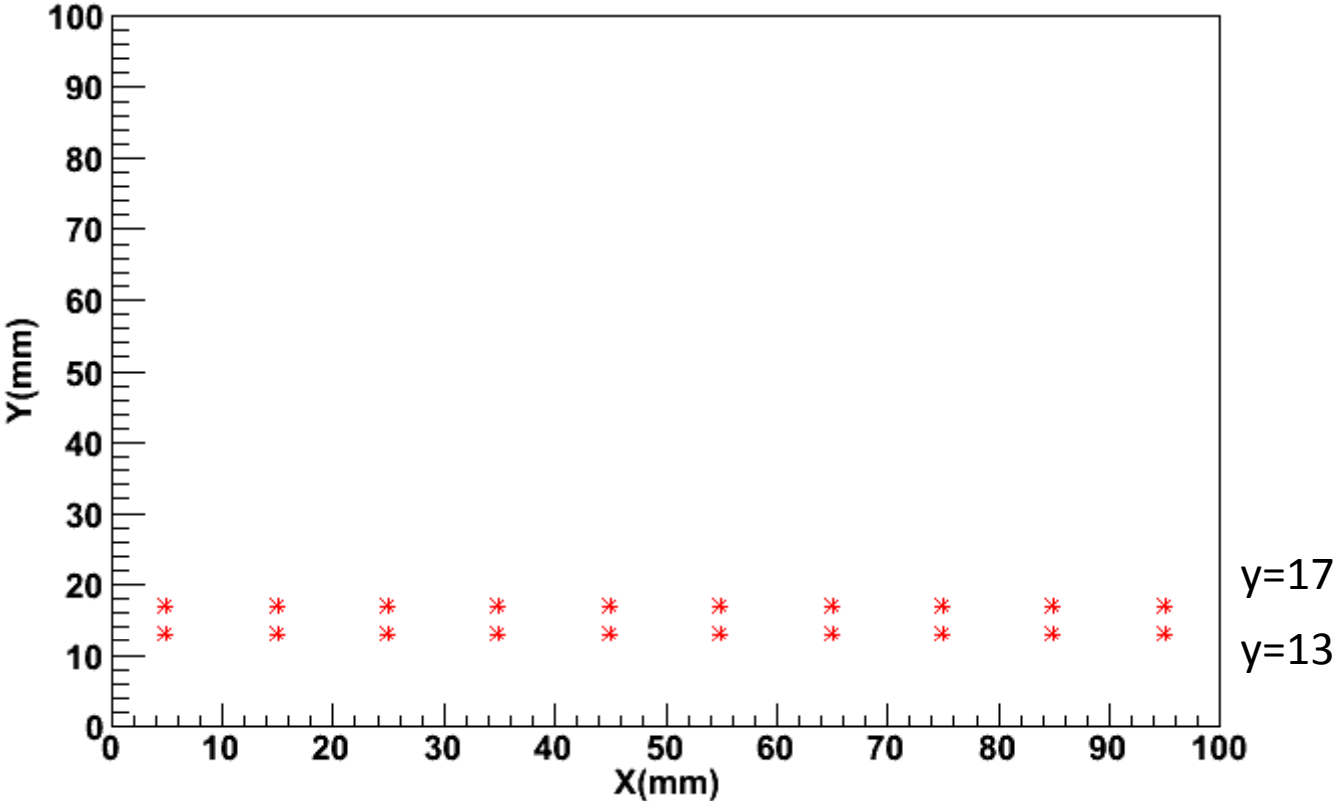


Conclusions from test 2

- Multi-track with noise hits can be recognized in different levels.
- With noise hits increasing, the number of found tracks decreases.

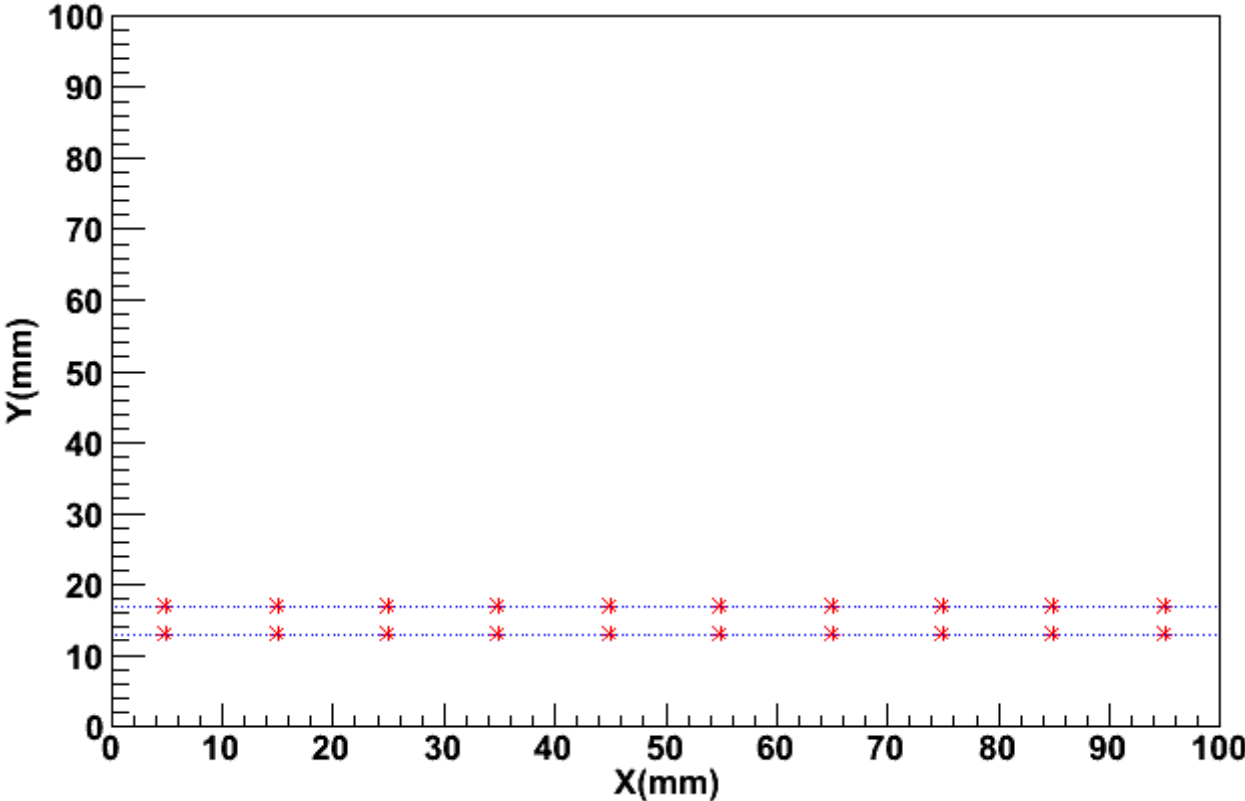
Test 3-1

Found Hits

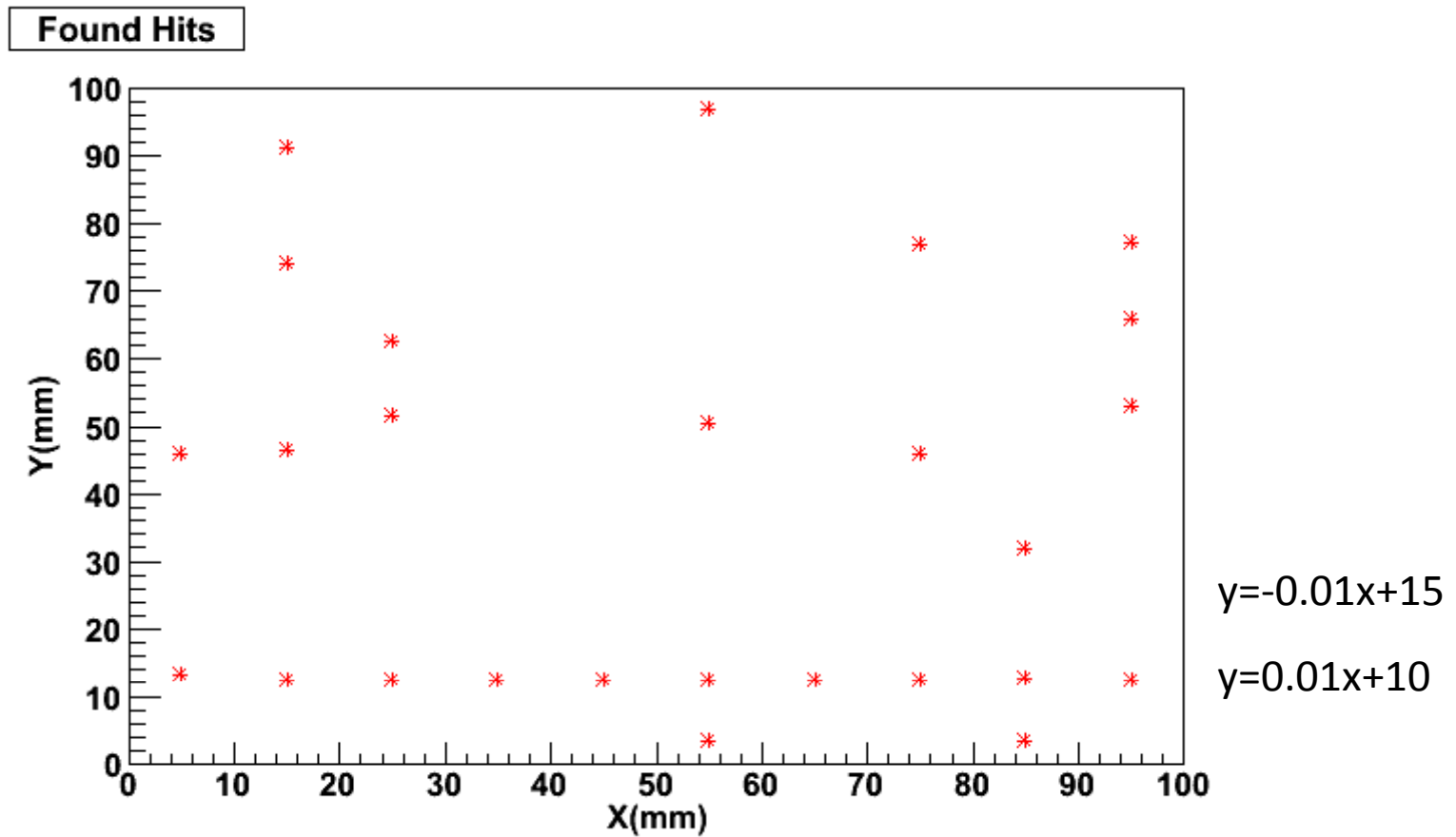


Test 3-1

Reconstructed Tracks



Test 3-2

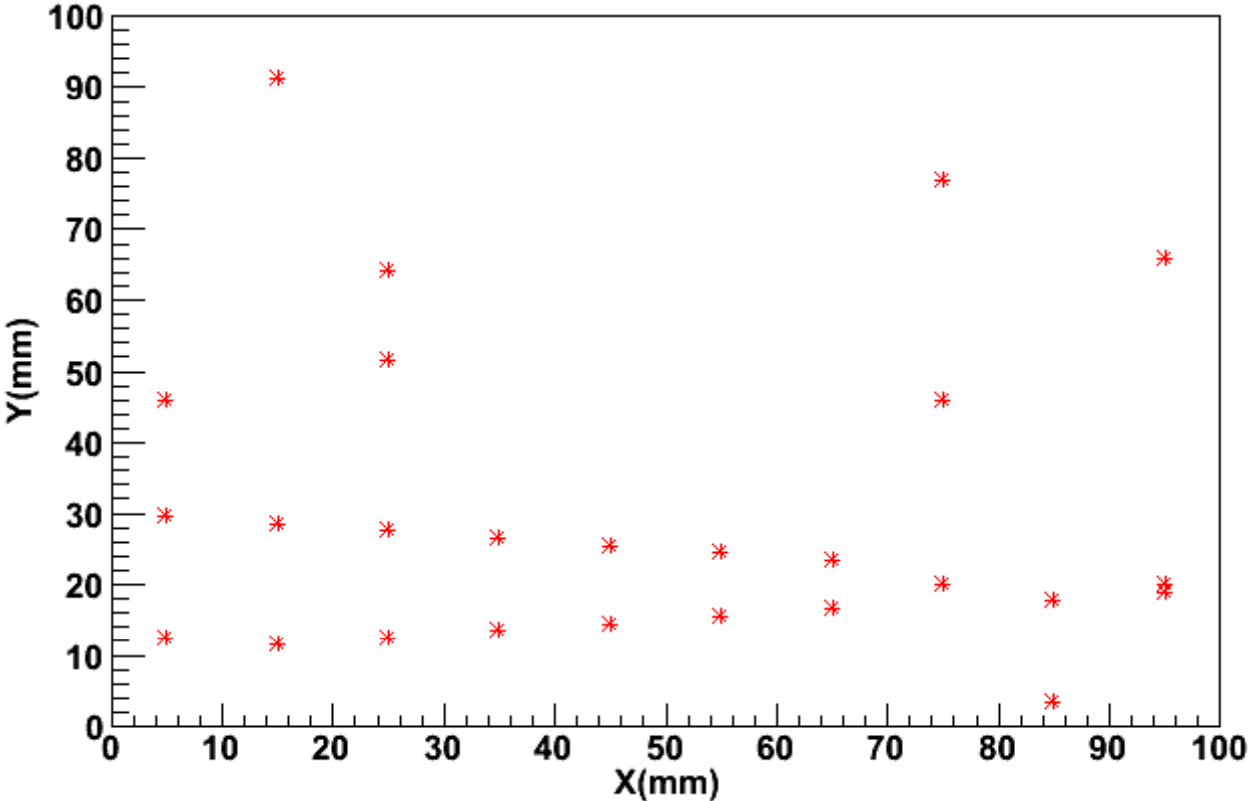


The min. separate is 4mm in this environment?

Hit reconstruction?

Test 3-3

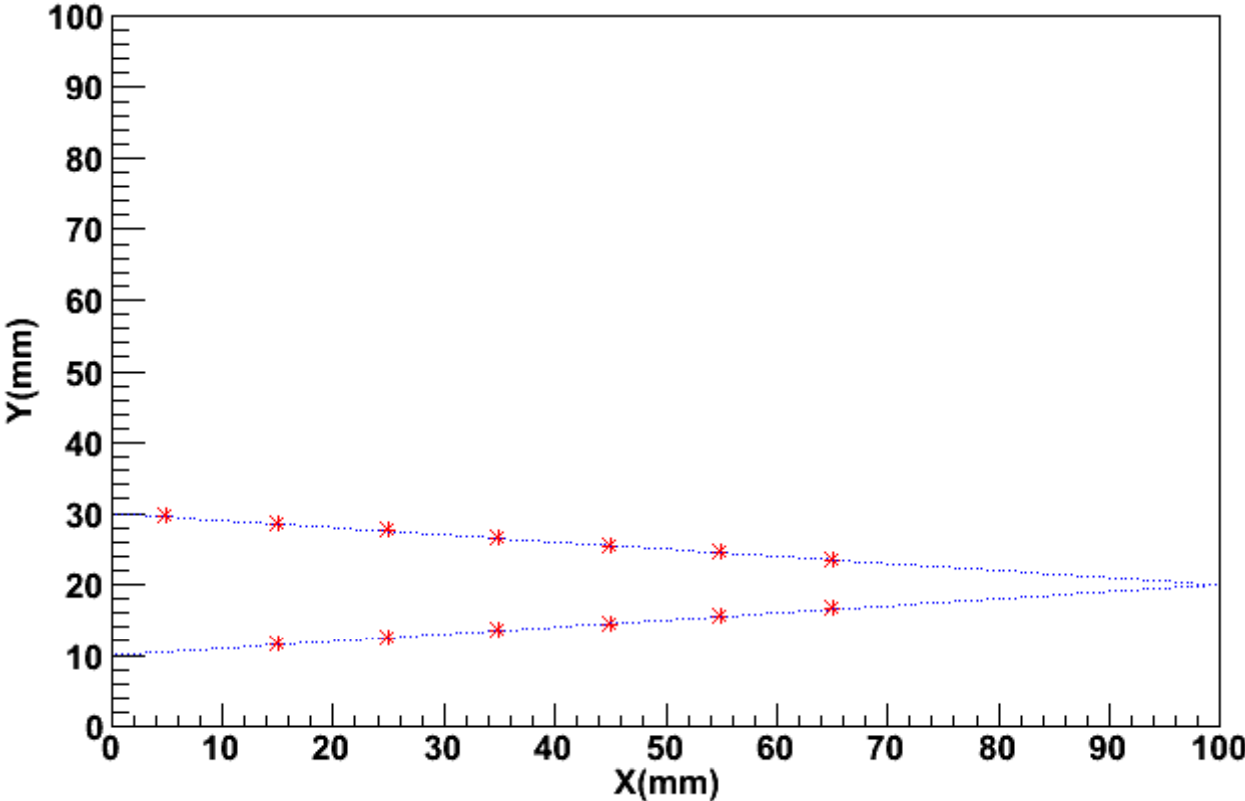
Found Hits



$y = -0.1x + 30$
 $y = 0.1x + 10$

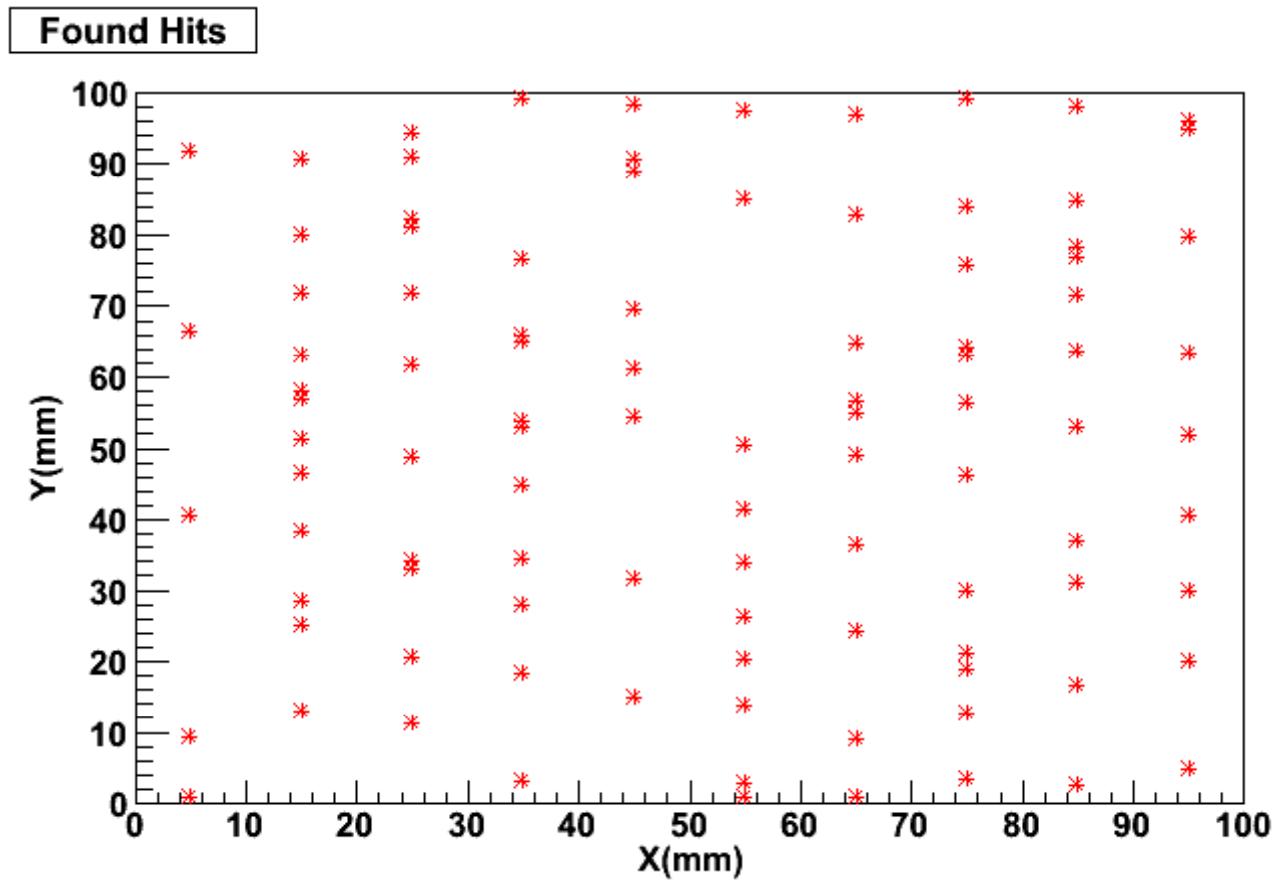
Test 3-3

Reconstructed Tracks



Test 4

200 noise hits



No track is found.

Conclusions from test 4

- In case the data is totally noise hit, track find code can get right result.
- But this progress will take more time for checking every noise hit.

Thank you!