

Generation and performance of pattern banks for FTK track reconstruction

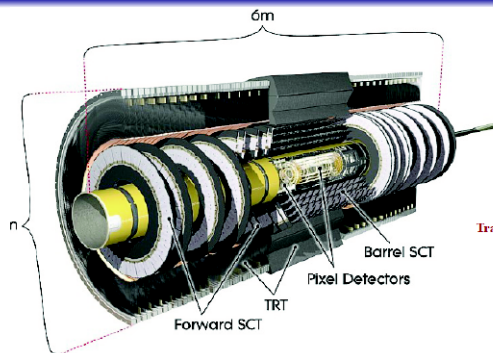
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Motivations

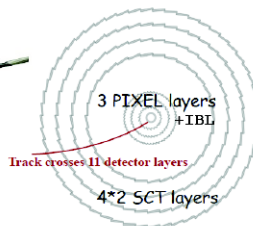
- High LHC luminosity \Rightarrow enormous background
 - The most interesting processes are rare and hidden under this extremely large background
 - Immense real-time data reduction is needed
 - Online track reconstruction is very important to separate the interesting events from the background
- \Rightarrow Solution: multi-level trigger system
- \Rightarrow In ATLAS: FTK

System description and segmentation



Example:

R-phi view of Barrel region:



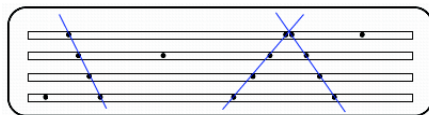
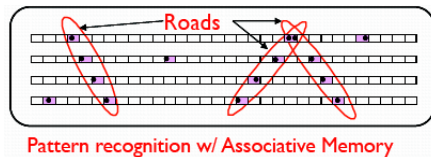
Segmentation

- 4 η -rings: 2 rings in the barrel, 2 rings in the endcap ($[1, 2.5]$ and $[-2.5, -1]$)
 - Each ring is divided into 16 ϕ -slices
- ⇒ 64 towers
- Endcap: $[0..15]$ and $[48..63]$
 - Barrel: $[15..47]$

FTK algorithm

Two sequential steps:

- 1 Pattern recognition, carried out by a dedicated device called Associative Memory (AM). Find coarse-resolution track candidates called “roads”.
- 2 Fit the full-resolution hits inside the road to determine the track parameters. Only the tracks passing the χ^2 cut are kept.



Parameters to define the pattern-bank performance

Pattern bank

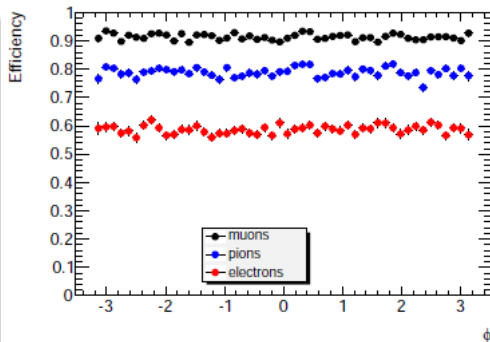
Each track generates a hit pattern. The collection of all these patterns defines both the space of the tracks we are looking for and how they appear in the detector: this collection is the pattern bank

Bank efficiency

$$Eff = \frac{NT(F)}{NT} \quad (1)$$

where:

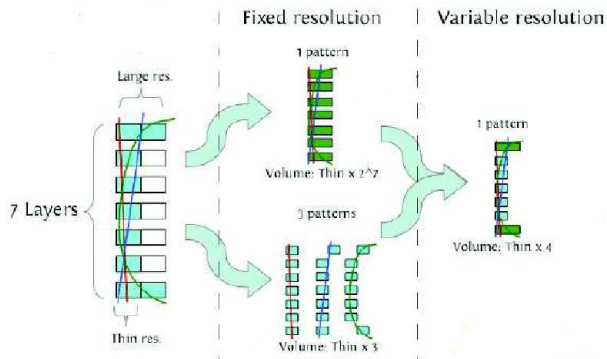
- NT = number of truth tracks
- $NT(F)$ = number of truth tracks matching FTK roads



Parameters to define the pattern-bank performance

- Critical parameter: road width
 - Too narrow \Rightarrow too large needed size of the AM \Rightarrow too large cost
 - Too wide \Rightarrow big number of fake roads \Rightarrow excessive work for the track fitter
 - Solution: “Don’t Care” (DC) bits
 - Variable resolution patterns
 - \rightarrow Each pattern for any layer can have an optimal width
 - \rightarrow Define the granularity layer by layer and pattern by pattern
- \Rightarrow Limits the number of patterns within the hardware limits
- \Rightarrow High rejection of fake roads

"Don't care" bits



- High compression factor in case of similar patterns
- Only one pattern location

Goal

① Performance study

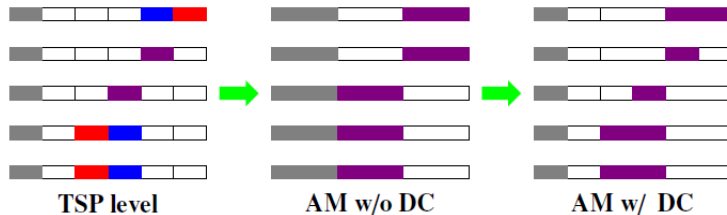
- High efficiency with limited number of patterns
- Limited data bandwidth to obtain high fake suppression

⇒ Optimal use of DC bits

② Improving pattern bank generation

- Reducing time of pattern generation
- Improving the quality of pattern generation sw

Multiple DC bits study



- 1 Generate a bank with maximum resolution patterns: the TSP bank (the violet SSs are those shared by the red and blue TSP patterns)
- 2 Cluster the TSP patterns in the widest patterns for the AM chip
- 3 Apply the DC-bits to change the shape of the AM patterns

HW constraints per pile-up events

Constraints for each board:

- 8M patterns
 - 8K roads
 - 40K fits
- ① 46 pile-up events: 16 boards working on 32 towers
⇒ constraints for each tower:
- $\#AM < 4M(4 * MPattern)$
 - $Roads < 4 * 10^3$
 - $Fits < 20 * 10^3$
- ② 72 pile-up events: 128 boards working on 64 towers
⇒ constraints for each tower:
- $\#AM < 16M(16 * MPattern)$
 - $Roads < 16 * 10^3$
 - $Fits < 80 * 10^3$

Configurations

- TSP bank configuration: $15 \times 16 \times 36$
 - 15×36 = number of pixels clustered in the same Super Strip
 - 16 = number of strips clustered in the same Super Strip
 - Dataset with 72 pile-up events
 - Constraints:
 - $\#AM < 16M(16 * MPattern)$
 - $Roads < 16 * 10^3$
 - $Fits < 80 * 10^3$
 - AM bank configurations: $(N_{DC}(pixel_x), N_{DC}(pixel_y)) - N_{DC}(SCT)$
 - 1 $(1,0)-1 \rightarrow 30 \times 32 \times 36$
 - 2 $(1,1)-1 \rightarrow 30 \times 32 \times 72$
 - 3 $(1,2)-1 \rightarrow 30 \times 32 \times 144$
 - 4 $(1,1)-2 \rightarrow 30 \times 64 \times 72$
- ⇒ Grouping with DC bit makes the SS granularity decreases

Endcap

0-7 towers:

DC bit	#TSP ·10 ⁶	#AM ·10 ⁶	Efficiency(%) R=64	Roads/evt ·10 ³	Fits/evt ·10 ³	Tracks/evt
(1,0)-1	120	34	87	5.8	33	109
(1,1)-1	120	18	91.2	7.1	56	106
(1,1)-1	112	16.8	91.2	6.9	55	...
(1,1)-1	100	15	91	6.2	50	...
(1,2)-1	...	8	92	5	90	...
(1,1)-2	...	8	93	9	154	...

Table: Results in 0-7 towers - endcap. #AM patterns, #Roads, #Fits and #Tracks are evaluated in tower 0.

- The TSP bank size is defined on the cut of the less frequent patterns
- The efficiency is evaluated on the single muon dataset
- #AM patterns, #Roads and #Fits are evaluated on the dataset with 72 pile-up events
- The #Roads provides a measure of the fake roads

Endcap

10-15 towers:

DC bit	#TSP $\cdot 10^6$	#AM $\cdot 10^6$	Efficiency(%) R=64	Roads/evt $\cdot 10^3$	Fits/evt $\cdot 10^3$	Tracks/evt
(1,1)-1	120	7259	58	102
(1,1)-1	100	15	91	6256	50	93
(1,1)-1	112	16.8	91.22	6871	55	99

Table: Results in 10-15 towers - endcap. #AM (#TSP) patterns, #Roads, #Fits and #Tracks are evaluated in tower 10.

Barrel

32-39 towers:

DC bit	#TSP $\cdot 10^6$	#AM $\cdot 10^6$	Efficiency(%) R=64	Roads/evt $\cdot 10^3$	Fits/evt $\cdot 10^3$	Tracks/evt
(1,0)-1	120	34	...	3836	23	51
(1,1)-1	120	21	94.71	4916	42	52

Table: Results in 32-39 towers - barrel. #AM patterns, #Roads, #Fits and #Tracks are evaluated in tower 32.

Barrel

26-31 towers:

DC bit	#TSP ·10 ⁶	#AM ·10 ⁶	Efficiency(%) R=64	Roads/evt ·10 ³	Fits/evt ·10 ³	Tracks/evt
(1,0)-1	120	34	90	3.8	23	...
(1,1)-1	120	21	94.75	3.9	33	42
(1,1)-1	100	18	94.07	3.4	28	38
(1,1)-1	90	16.8	93.35	3.2	26	36
(1,2)-1	...	8	95	4	60	...
(1,1)-2	...	8	96	6	98	...

Table: Results in 26-31 towers - barrel. #AM (#TSP) patterns, #Roads, #Fits and #Tracks are evaluated in tower 26.

Work in progress

- Exploring new TSP bank configurations
- TSP bank configuration: $11 \times 12 \times 18$
 - 11×18 = number of pixels
 - 12 = number of strips
- We have generated the TSP bank
- We are trying some DC-bits bank configurations:
 - $(1,2)$ -1 → $22 \times 24 \times 72$
 - $(2,2)$ -2 → $44 \times 48 \times 72$
- We will have the efficiency numbers soon

Improvement

“State of the art”

Generating the TSP bank:

- C-written sw 5 years old
 - Still large SVT legacy
- The old sw is efficient but it's almost impossible to maintain or to improve it
 - It has many functionalities (sectors and patterns from real, pattern from constants)
 - A specific target should improve its performance
- The old sw can't be integrated with ATLAS production

Improvement

- GOAL: Generating the TSP pattern bank as quickly and efficiently as possible
- ⇒ Collaboration with the Japanese group to update the existing tools in order to make the existing generation system:
- modern
 - efficient
 - quick
 - maintainable

Steps

- ① Porting the old source code from C to C++
 - Formats optimization
 - ASCII → ROOT
 - Performance check
 - Final tests
- ② Revisiting the existing algorithm and its optimization
 - Don't change the original algorithm and check the performance
- ③ Searching for alternative methods for pattern generation in order to
 - Improve the patterns' efficiency
 - Improve the patterns' quality

Results

Step 1: almost done!

- Pattern production algorithm ported

Performance (time for each track): old MakeBank vs new PattBankGen

- MakeBank
 - cpu: $0.035ms/track$
 - real: $0.036ms/track$
- PattBankGen
 - cpu: $0.037 + / - 0.001ms/track$
 - real: $0.123 + / - 0.018ms/track$