

Qjets

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NPKI 2/28/12

Based on work with S. Ellis, A. Hornig, T. Roy, and M. Schwartz: arXiv:1201.1914, accepted by PRL and work in progress with above plus D. Kahawala, A. Thalapillil, and L.-T. Wang

Outline

- ❖ Quick Review of Jet Algorithms
- ❖ Introduction to Qjets
- ❖ Implementation
- ❖ Example: Jet Pruning
- ❖ Future Directions & Conclusions

Takeaway

- ❖ Many jet substructure analyses employ trees
- ❖ But, more than one tree can plausibly be associated with a jet
 - ❖ Typically, we use k_T or C/A to choose the “best” tree
- ❖ However, if we force ourselves to only consider a single tree for each jet, we make ourselves more susceptible to arbitrary choices of the jet algorithm
- ❖ By looking at many trees for each jet, we can decrease random fluctuations and create a more powerful analysis

Review of Jets & Jet Substructure

Types of Algorithms

- ❖ There are two main classes of jet algorithm

- ❖ Sequential recombinations

- ❖ Combine four-momenta one by one

Focus on these



- ❖ Cone algorithms

- ❖ Stamp out jets as with a cookie cutter

Sequential Recombination

- ❖ Define a distance measure between every pair of four-momenta in an event (jet-jet distances)

$$d_{ij}$$

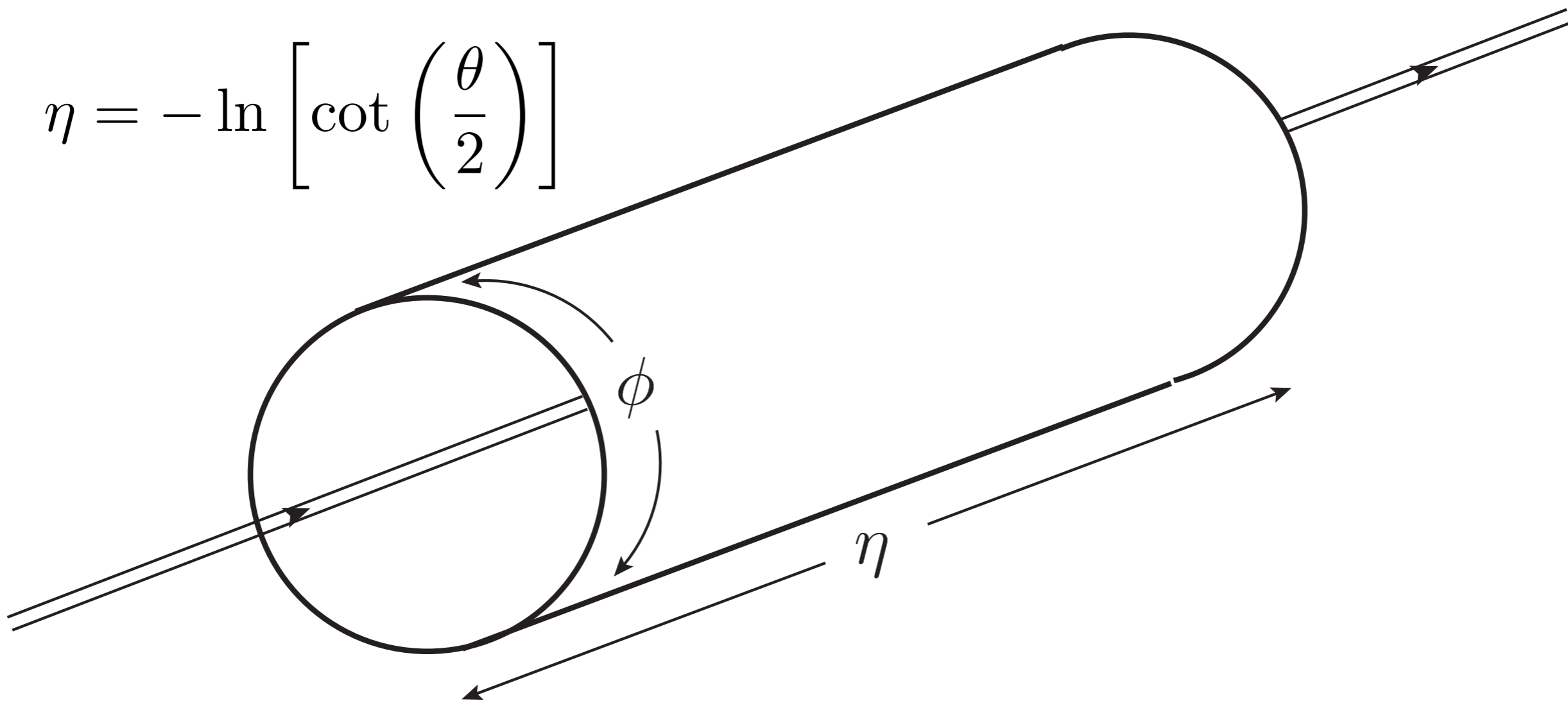
- ❖ Define a distance measure for each four-momenta individually (jet-beam distances)

$$d_{iB}$$

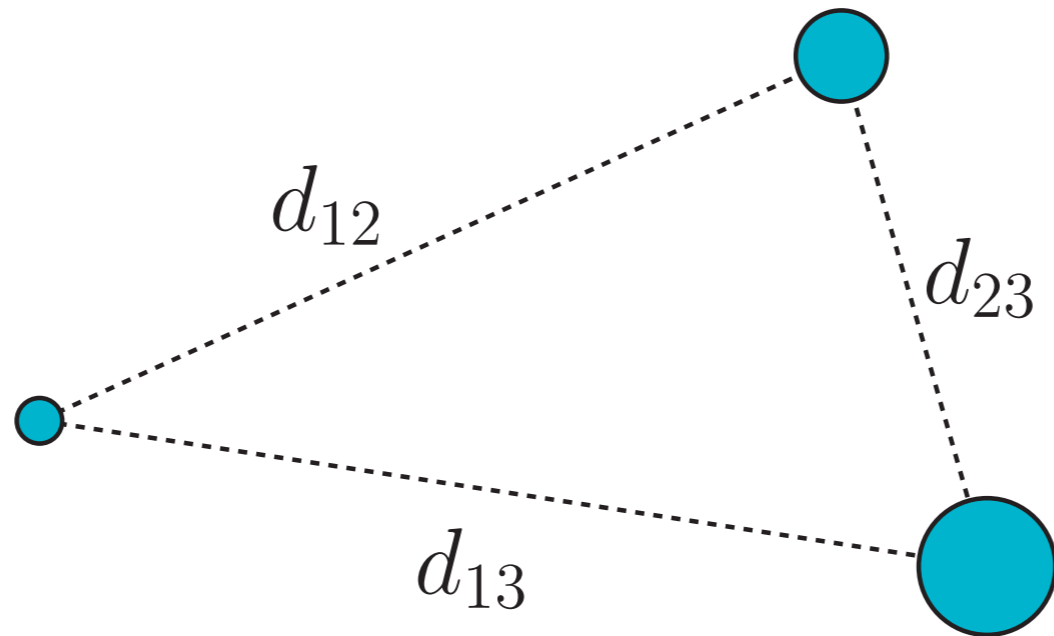
- ❖ If smallest distance at any stage in clustering is jet-jet, add together corresponding four-momenta
 - ❖ Otherwise take jet with smallest jet-beam distance and set it aside
- ❖ Repeat till all jets are set aside
- ❖ In this way, jets are constructed by pairwise recombinations - get a tree-like sequence at the end.

Coordinate System

$$\eta = -\ln \left[\cot \left(\frac{\theta}{2} \right) \right]$$



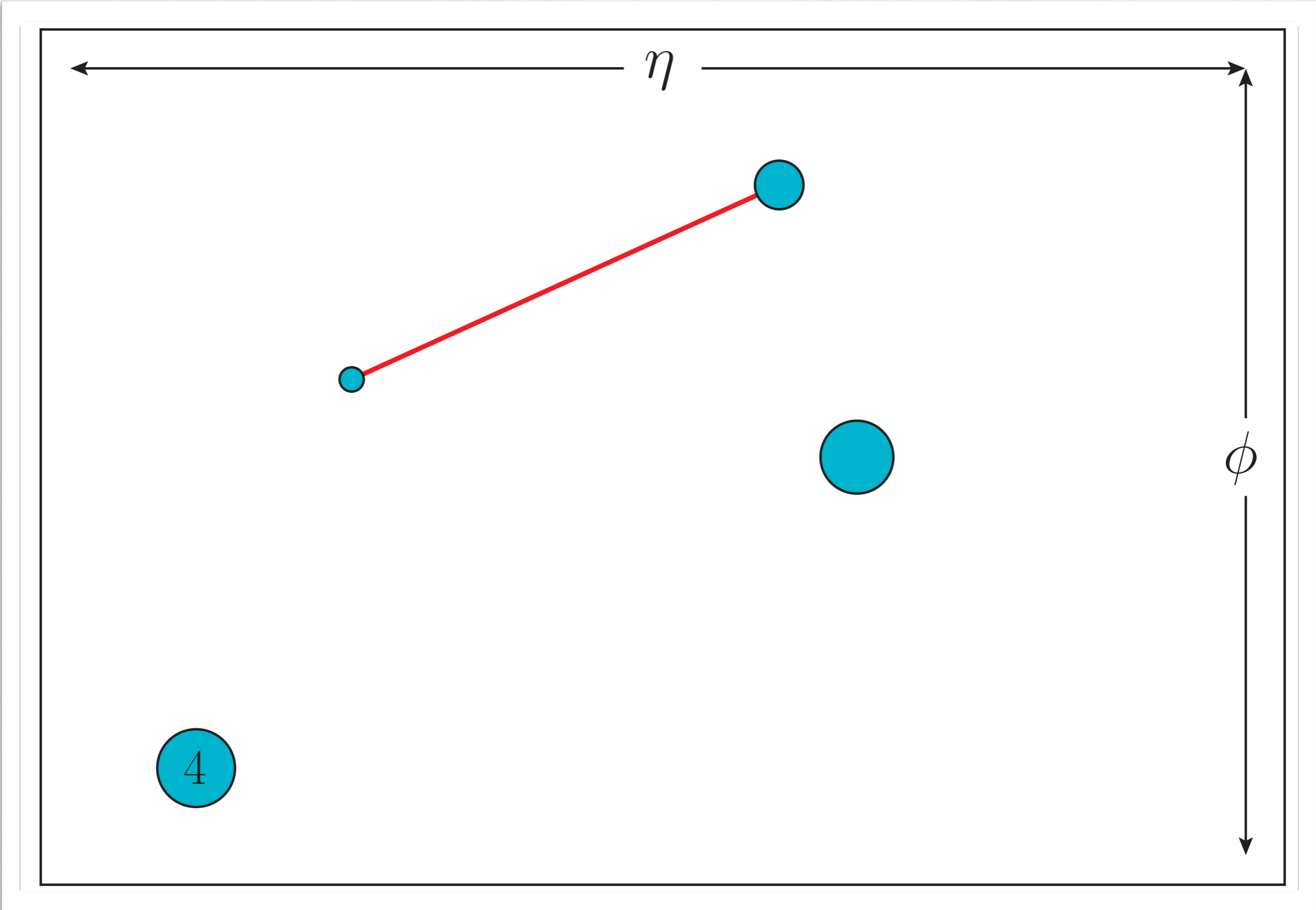
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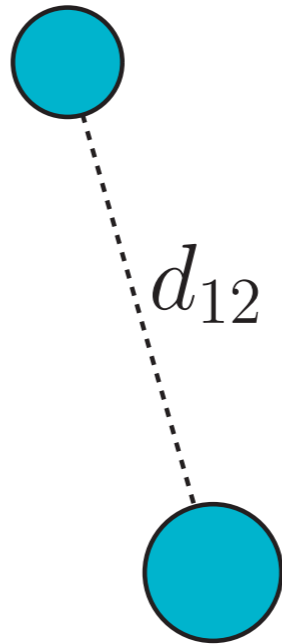
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$$d_{12} < d_{13} < d_{23} < d_{(1,2,3)B} < d_{i4}$$

4



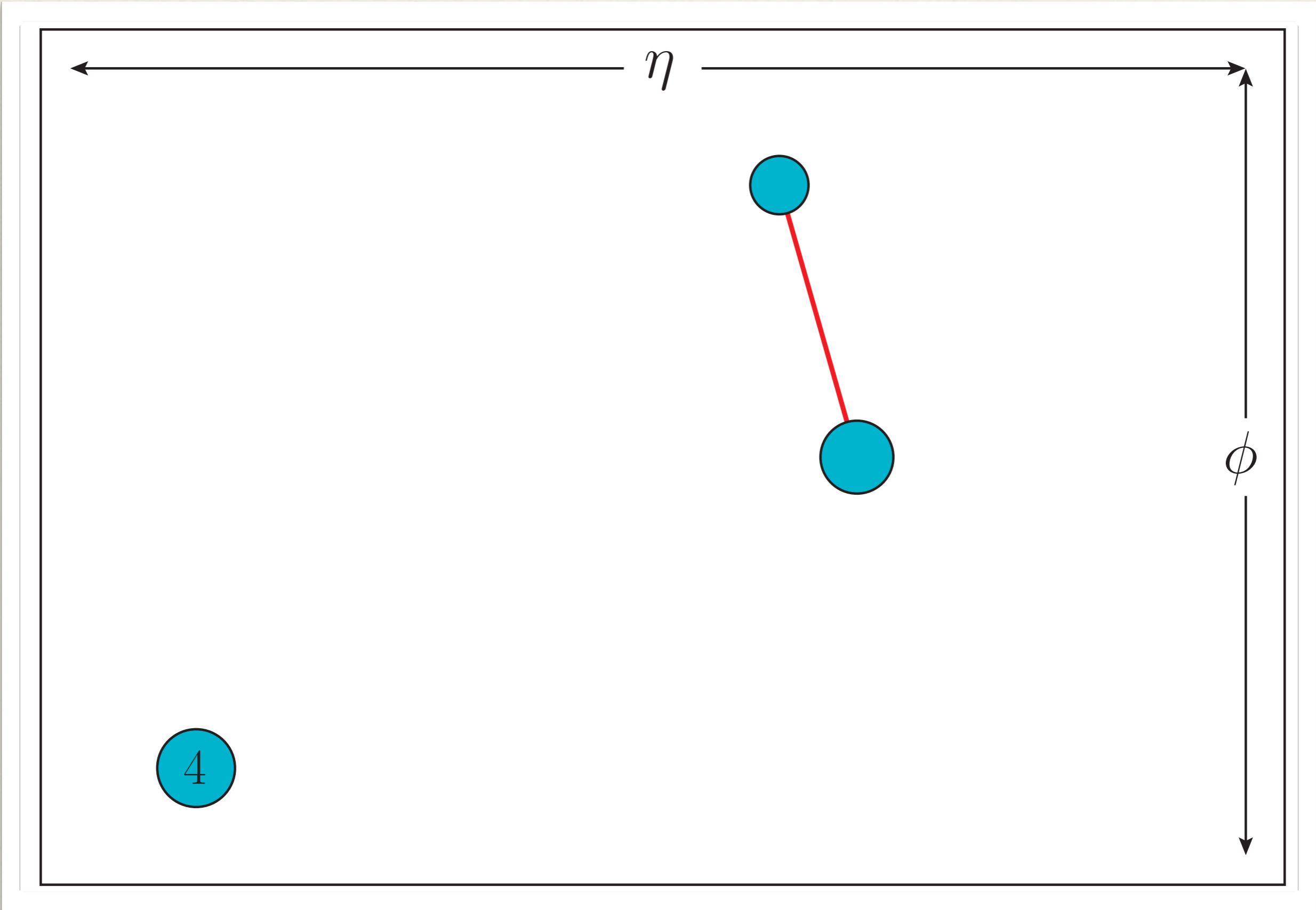
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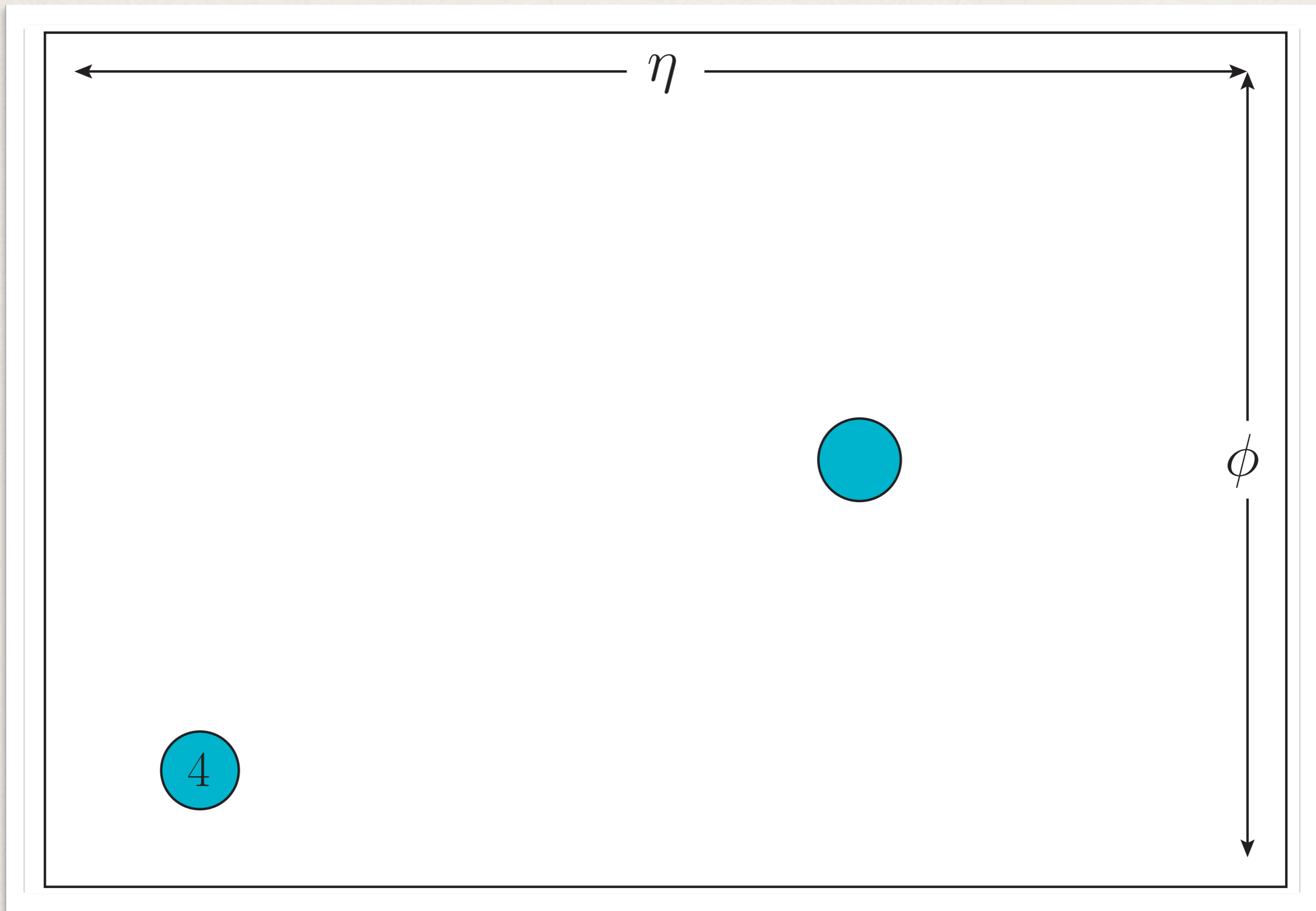


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$$d_{12} < d_{(1,2)B} < d_{i4}$$

4





Done!

Standard Recombination Algorithms

- ❖ k_T algorithm

$$d_{ij} = \min(p_{Ti}^2, p_{Tj}^2) \left(\frac{\Delta R}{R_0} \right)^2, \quad d_{iB} = p_{Ti}^2$$

- ❖ C/A algorithm

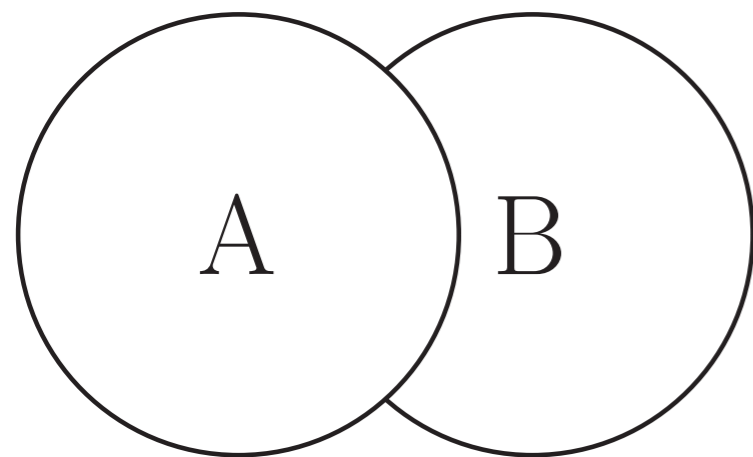
$$d_{ij} = \left(\frac{\Delta R}{R_0} \right)^2, \quad d_{iB} = 1$$

- ❖ anti- k_T algorithm

$$d_{ij} = \min(p_{Ti}^{-2}, p_{Tj}^{-2}) \left(\frac{\Delta R}{R_0} \right)^2, \quad d_{iB} = p_{Ti}^{-2}$$

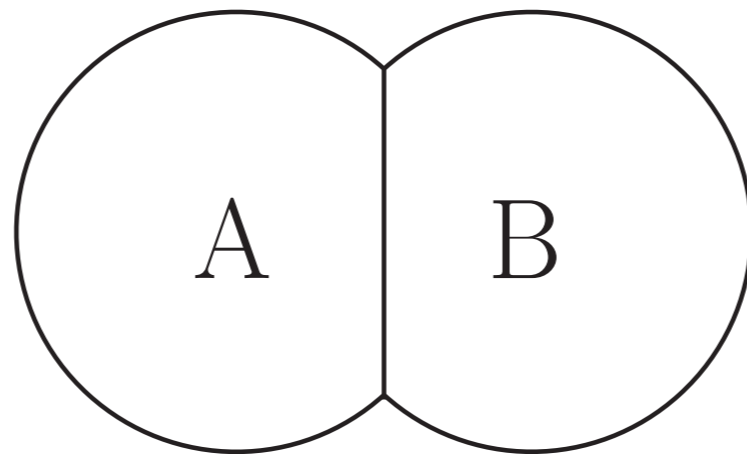
Approximate Jet Behavior:

$$p_{TA} > p_{TB}$$



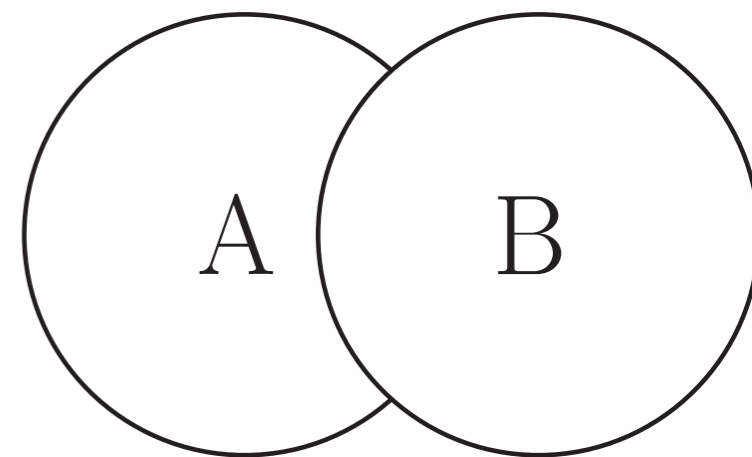
anti- k_T

Hard to Soft



C/A

Near to Far



k_T

Soft to Hard

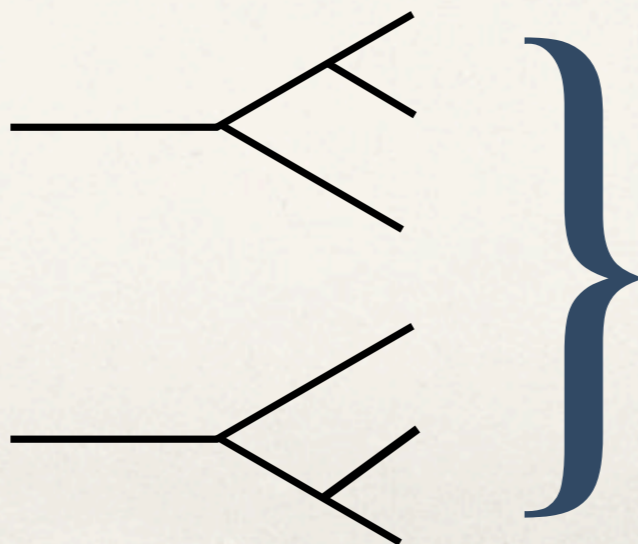
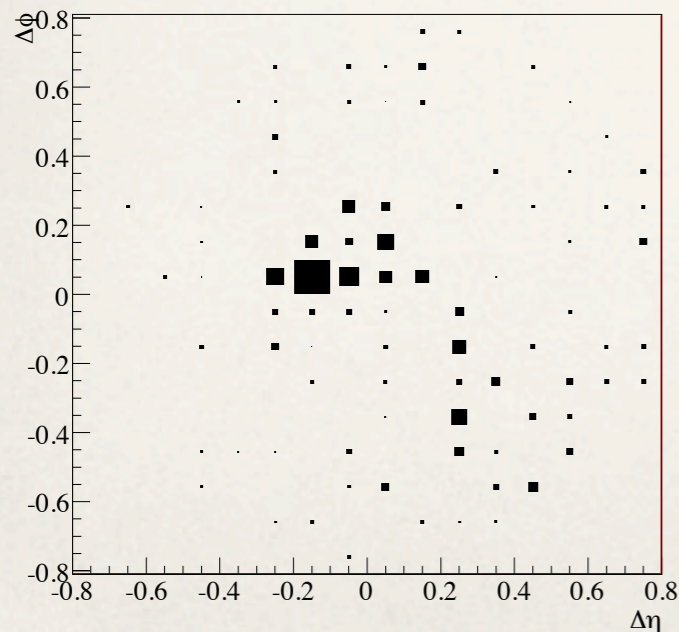
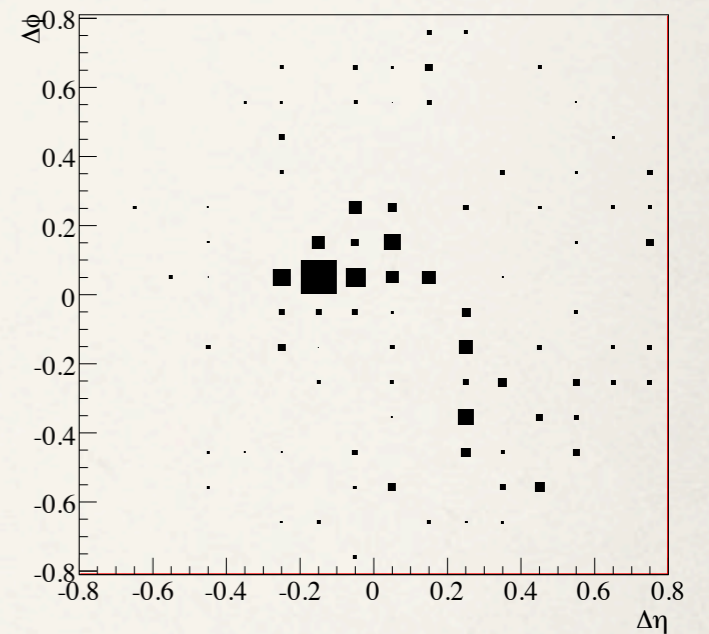
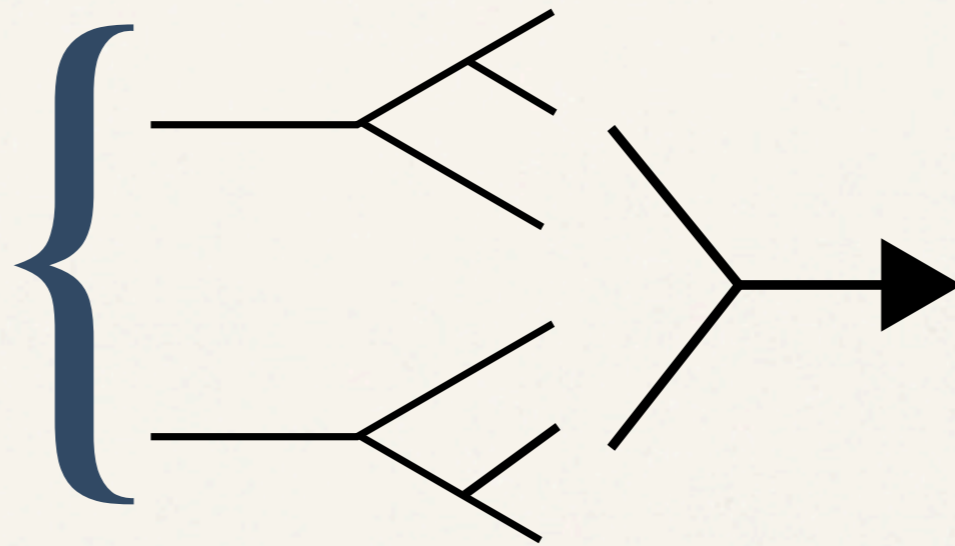
Qjets

Two Basic Approaches to Substructure

1. Consider only the two-dimensional distribution of energy in a jet
 - * Examples: Trimming & Filtering, N-Subjettiness, Jet substructure w/o trees
2. Try to associate a tree structure with a jet
 - * Allows one to use heuristic pictures of parton shower & decay chains.
 - * Examples: Pruning, energy sharing variables, mass drop
 - * However, the current procedure for constructing a tree is not ideal.

Mapping Jets to Trees

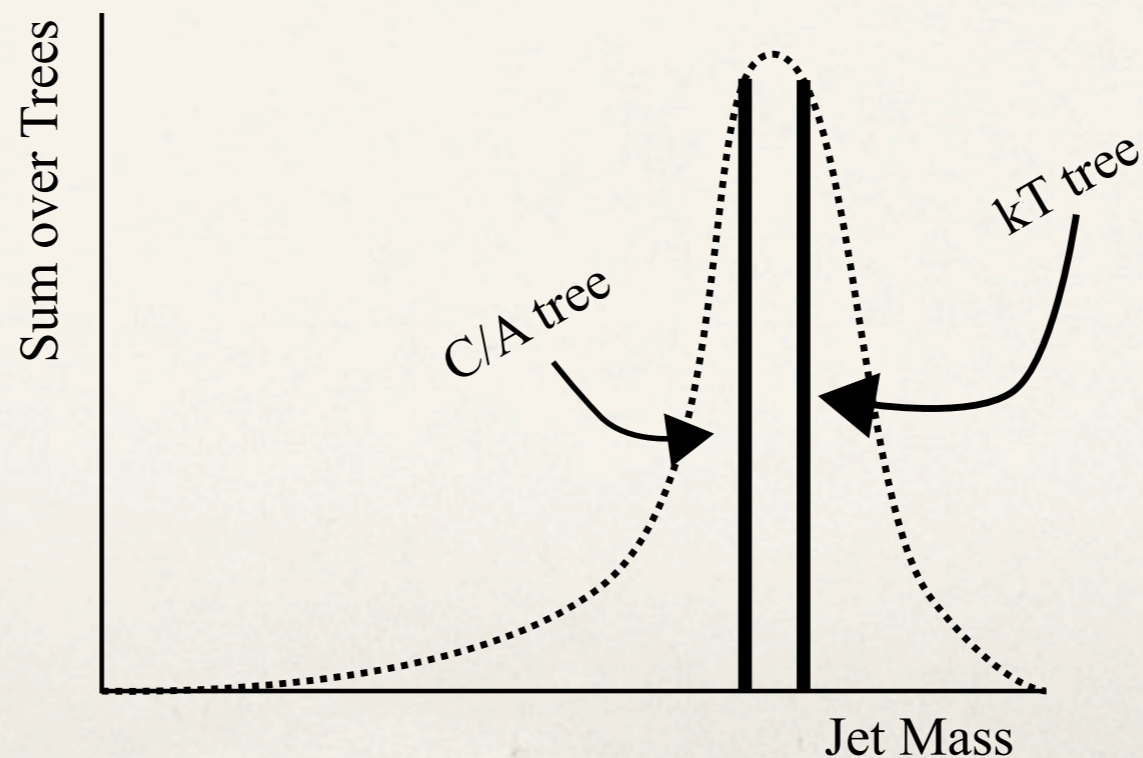
The energy distribution for a particular tree is unambiguous



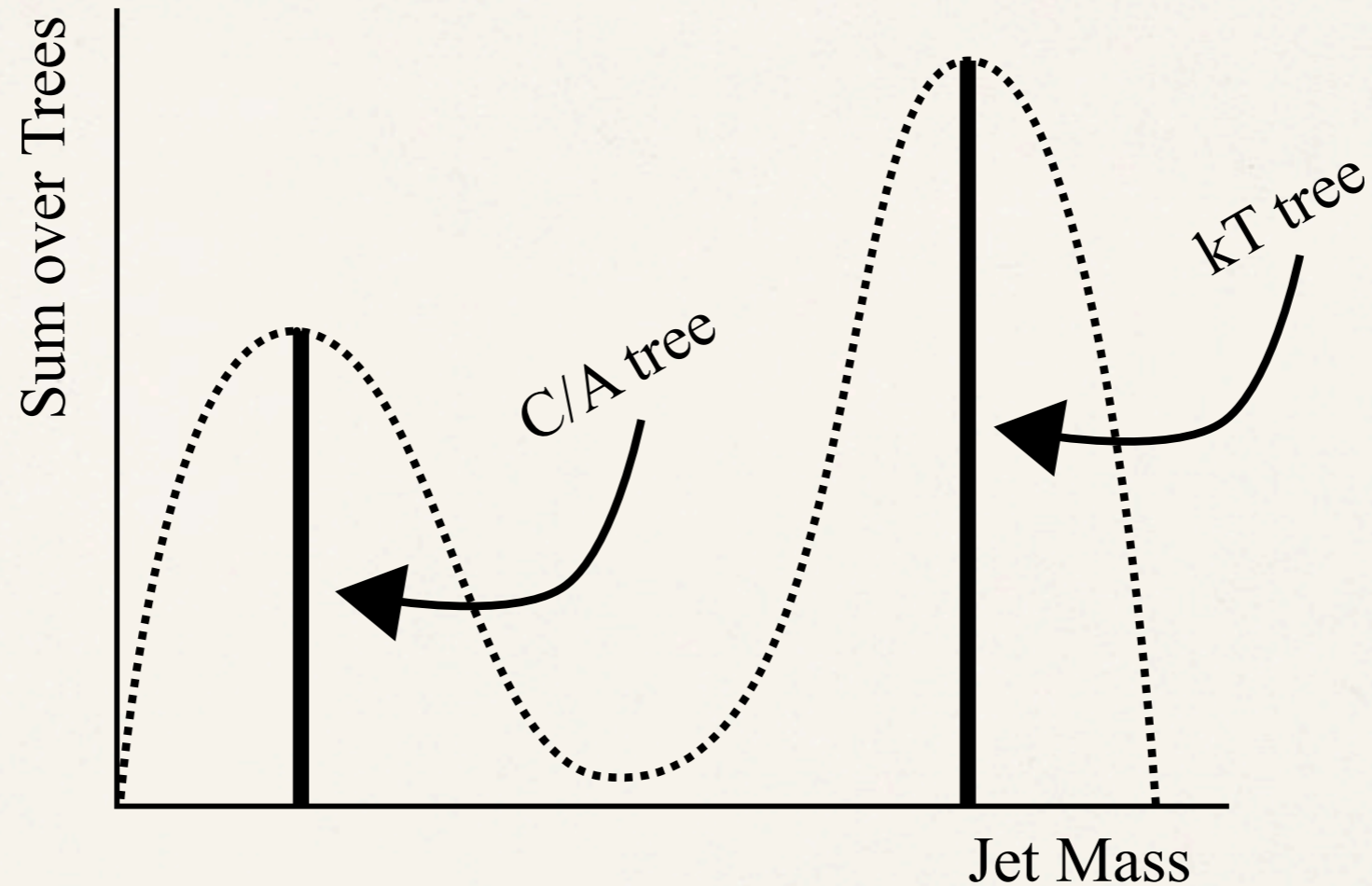
But, more than one tree can correspond to the same energy distribution

Unnecessary Choices

- ❖ How do we assign a particular tree to an energy distribution?
- ❖ Standard answer: Use a well motivated algorithm like C/A or kT
- ❖ Ideally, since both are well motivated algorithms they'll give the same answer:



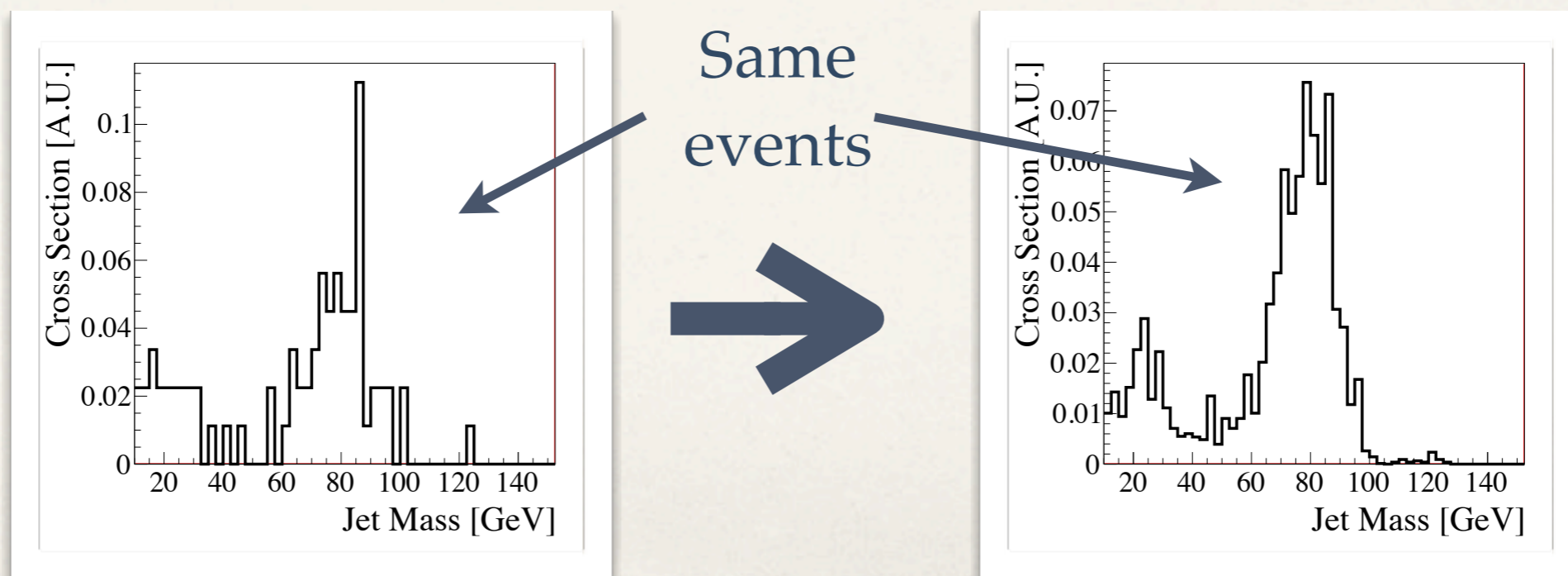
- ❖ However, sometimes the answers are very different.

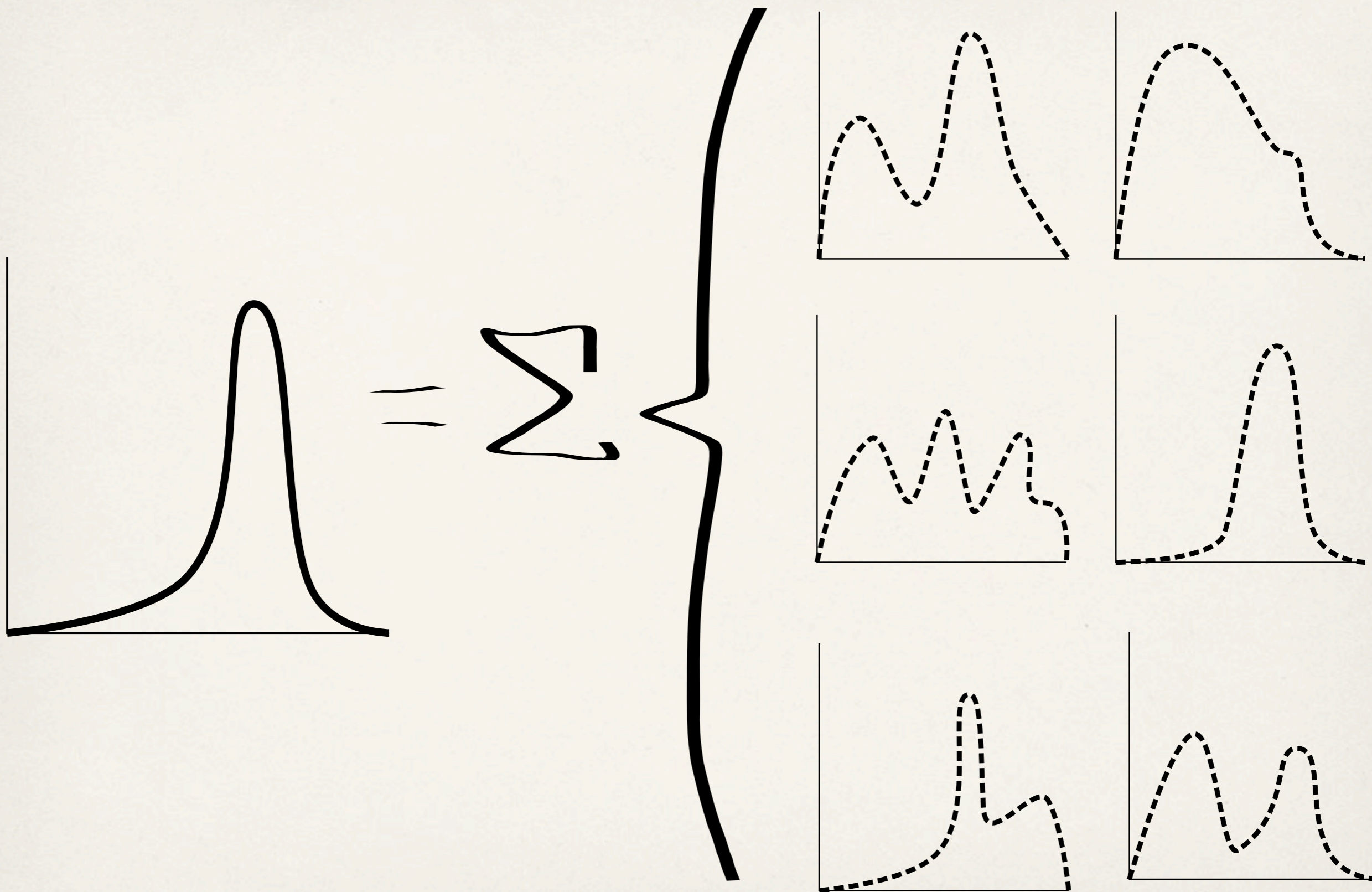


- ❖ Considering only the kT or C/A tree introduces an element of randomness into this process, resulting in unnecessary fluctuations in the final state observable.
- ❖ Intuitively it makes sense that defining an observable in a way which reflects the ambiguity of this clustering should yield better results.

Solution: Sum over Trees

- ❖ We propose that rather than assigning a single number to each event, instead each event should contribute a distribution obtained by summing the observable over many trees.
- ❖ When we sum these together, the result is much more stable than the histogram we would have had if we just considered one number per event.





Weights

- ❖ The only question is: when we add together the result obtained from different trees, how should we weight each tree's contribution?
- ❖ Surely they should not all count equally. If they did, then why would we use kT or C/A to find our trees in the first place?
- ❖ In theory, one could weight each tree by the product of splitting functions and Sudakovs one would obtain from a parton shower.
 - ❖ Work in progress.

Implementation

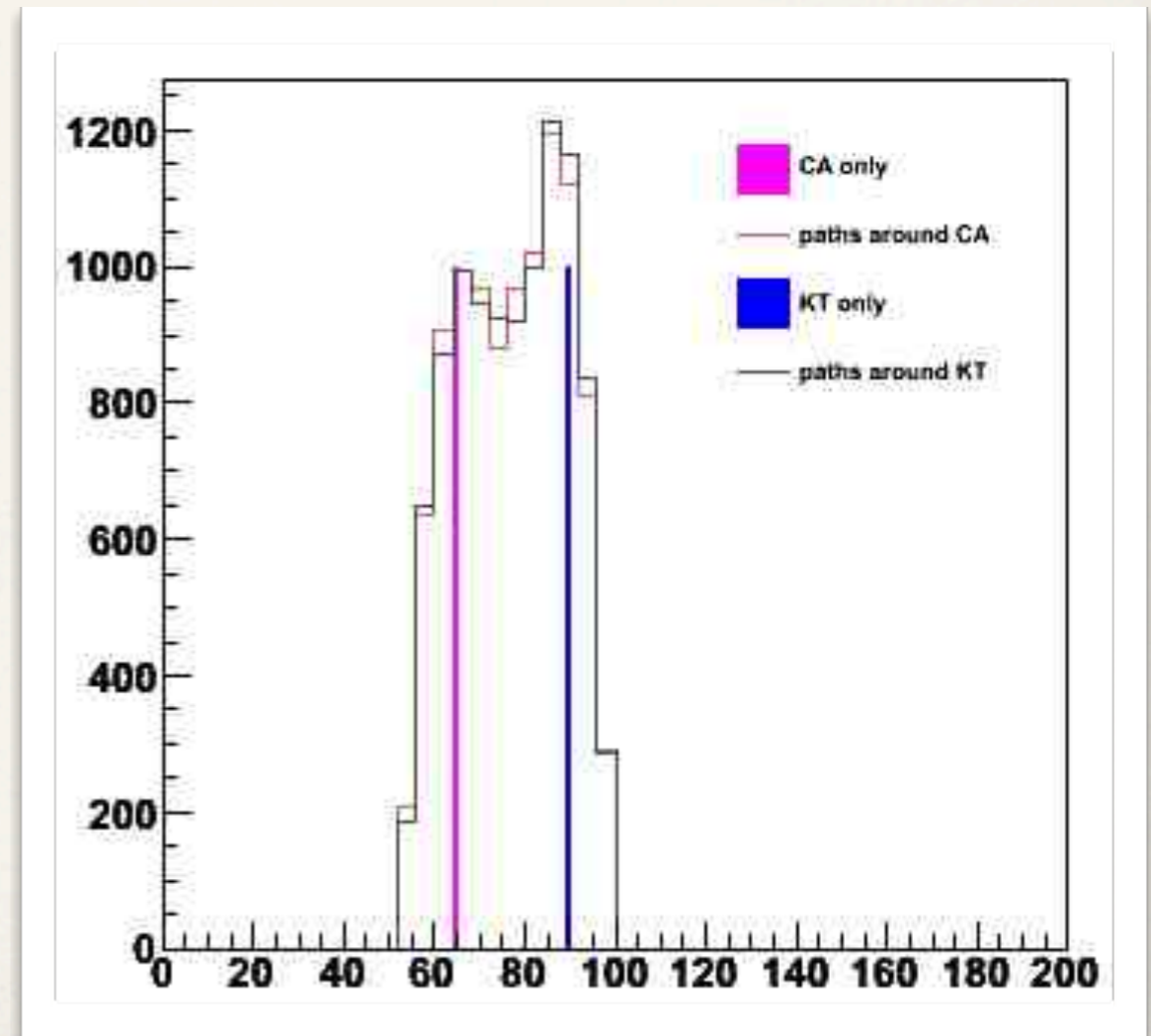
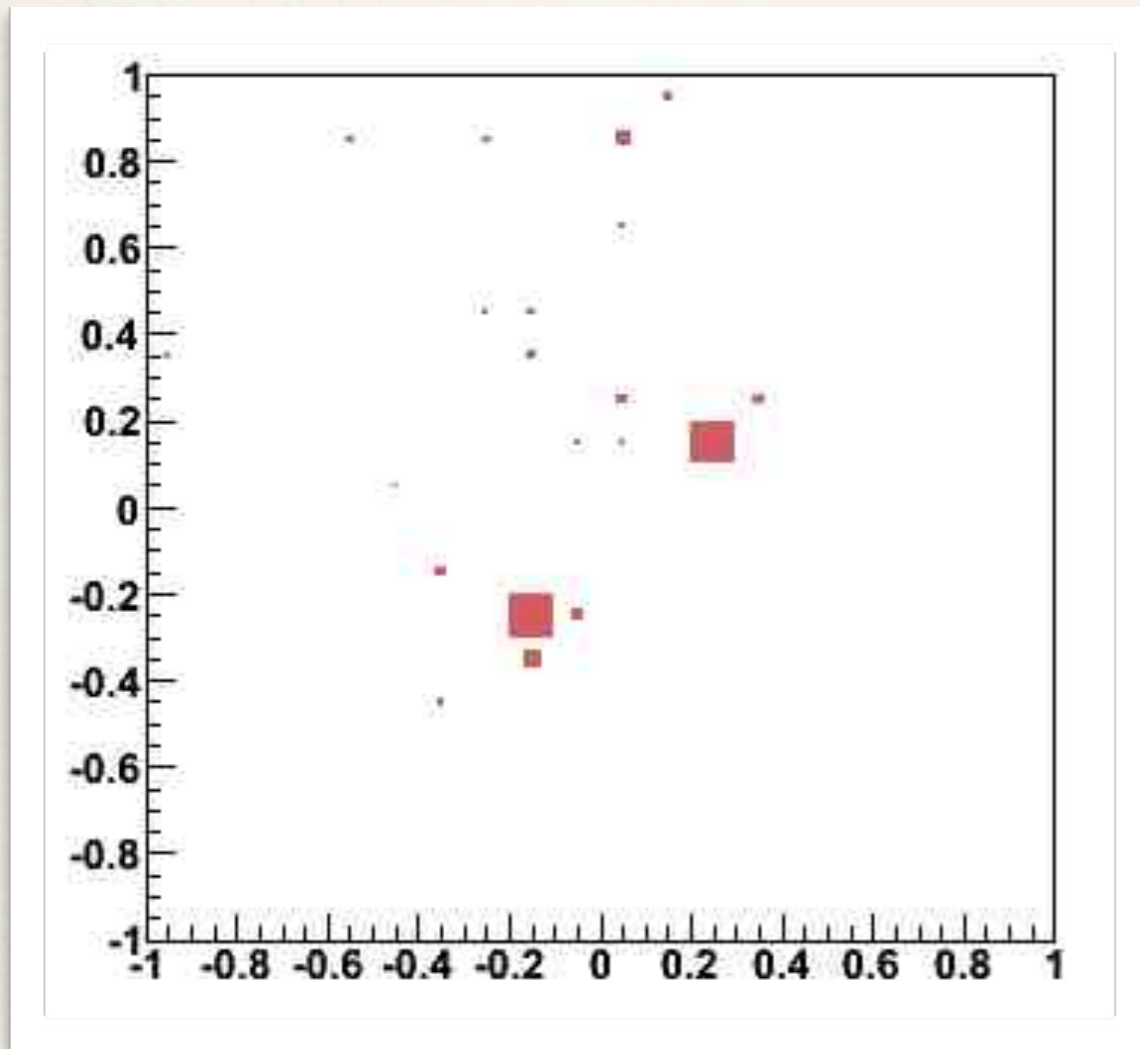
- ❖ Instead, we find a simpler Monte-Carlo procedure works quite well.
 - ❖ As in a sequential recombination algorithm, assign every pair of proto-jets a distance measure d_{ij} .
 - ❖ However, unlike a normal sequential algorithm (where the pair with the smallest measure is selected clustered), here we suggest that a given pair be randomly selected for merging with probability

$$\Omega_{ij} \equiv \frac{1}{\Omega} \exp \left(-\alpha \frac{d_{ij}}{d_{ij}^{\min}} \right), \quad \alpha = \text{rigidity parameter}$$

- ❖ Thus, paths which deviate from the CA or kT behavior are less likely to occur
- ❖ Repeat many (~ 100) times, till the distribution stabilizes

- ❖ The result is that you get many trees
- ❖ The probability of finding a given tree decreases as it becomes less k_T or C/A like

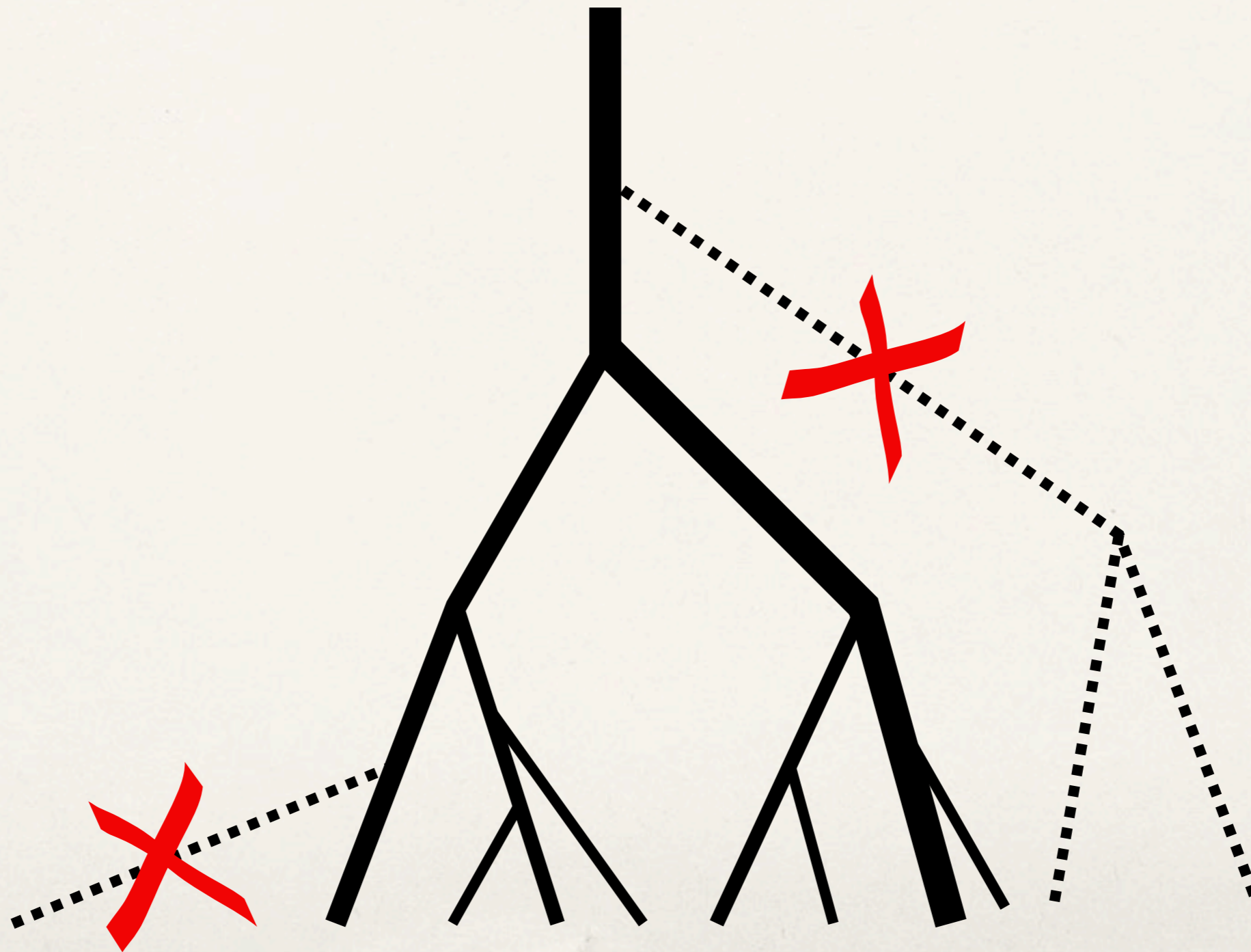
Example: Boosted W-Jets with Pruning



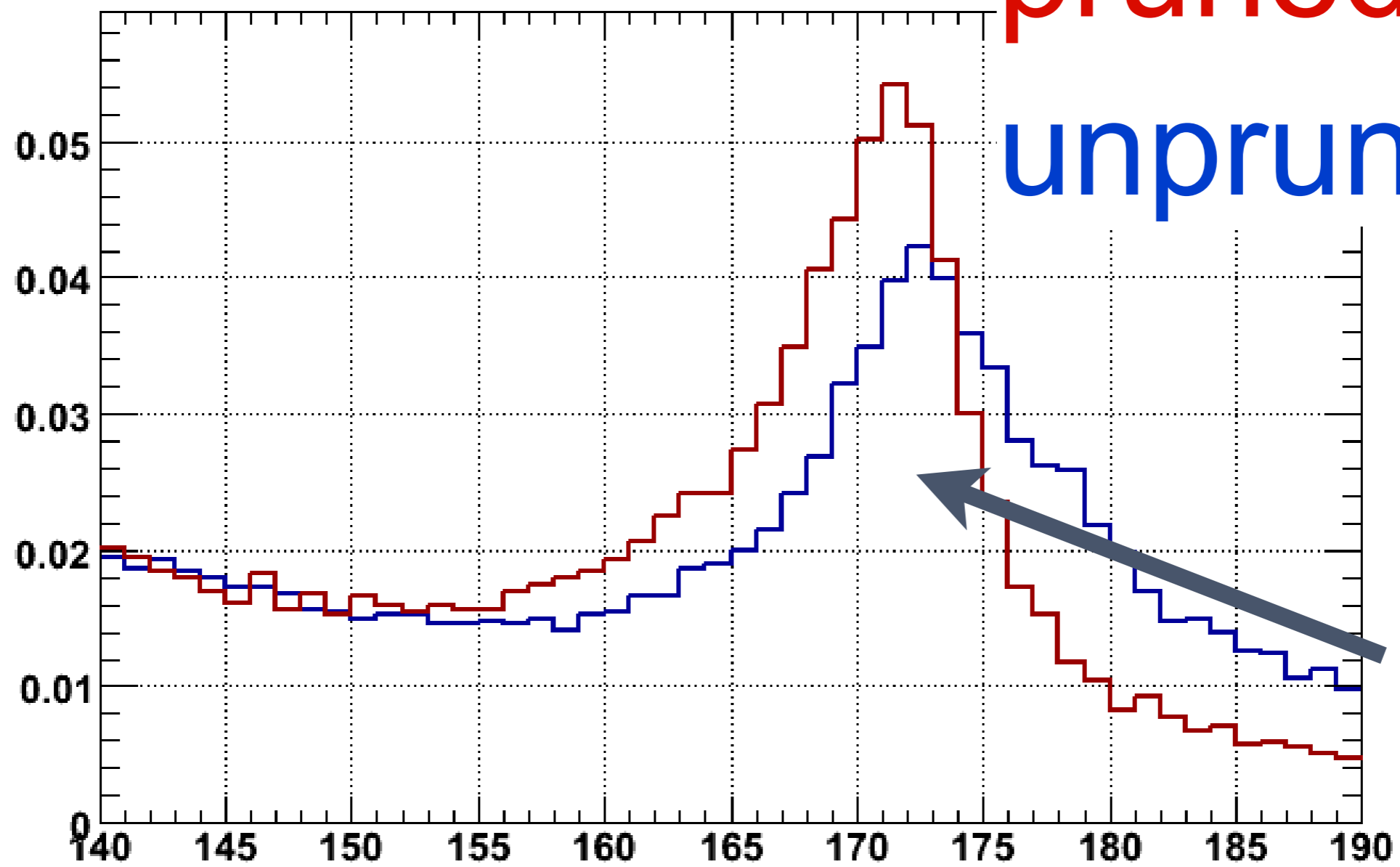
Pruning

- ❖ Pruning was introduced to look for boosted heavy objects (e.g., tops, higgses, W 's, etc) by cleaning up their mass.
- ❖ Intuition: QCD has soft / collinear singularities. Wide-angle emissions should come from hard decays.
 - ❖ Remove all parts of the jet which are *both* soft and wide angle.
- ❖ Two main advantages:
 - ❖ Boosted objects see their mass reconstruction improved
 - ❖ Massive QCD jets (a large background) see their mass substantially decreased -> lower backgrounds

A Pruned Tree



jet mass for jets with $p_T > 200$



pruned

unpruned

Top jets

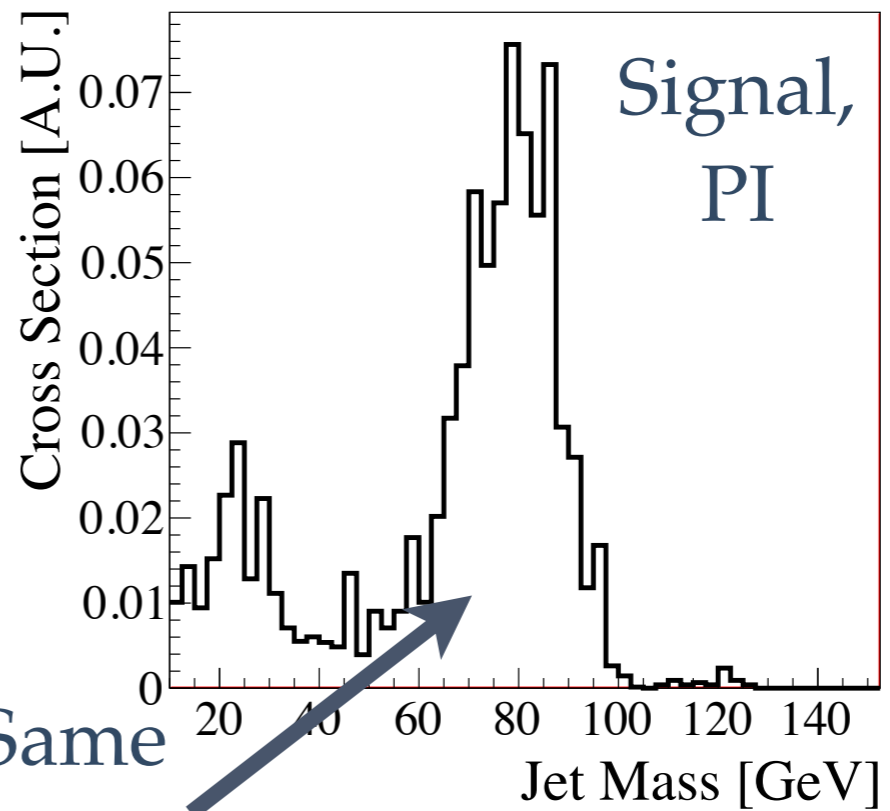
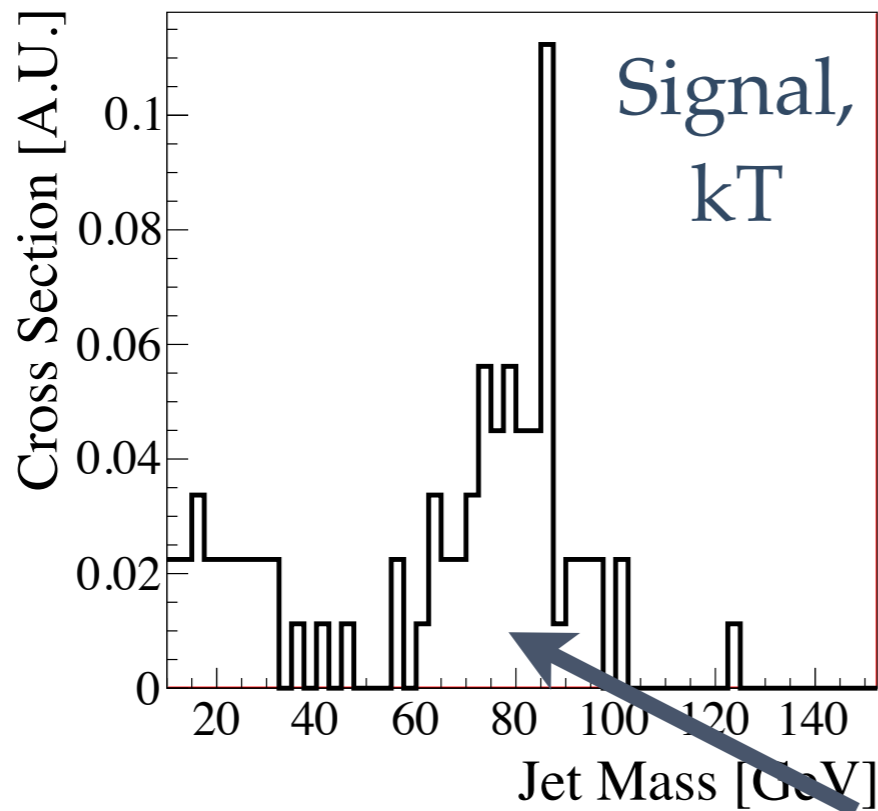
- ✦ Let's see what happens when we modify pruning so that it runs over trees generated via the Qjet procedure.

Signal Discovery & Exclusion

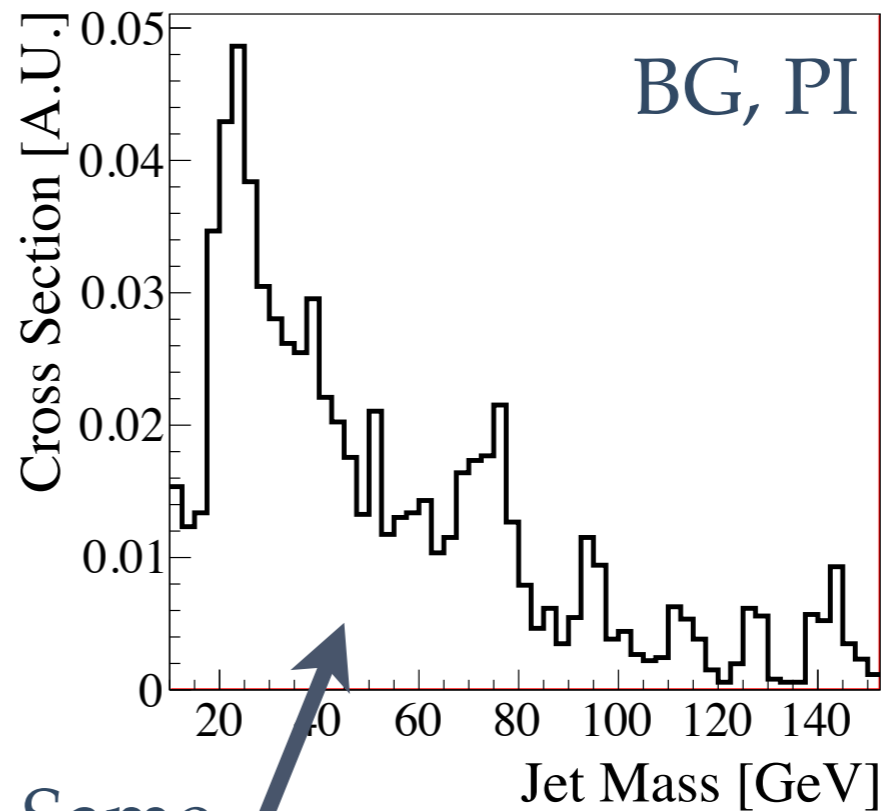
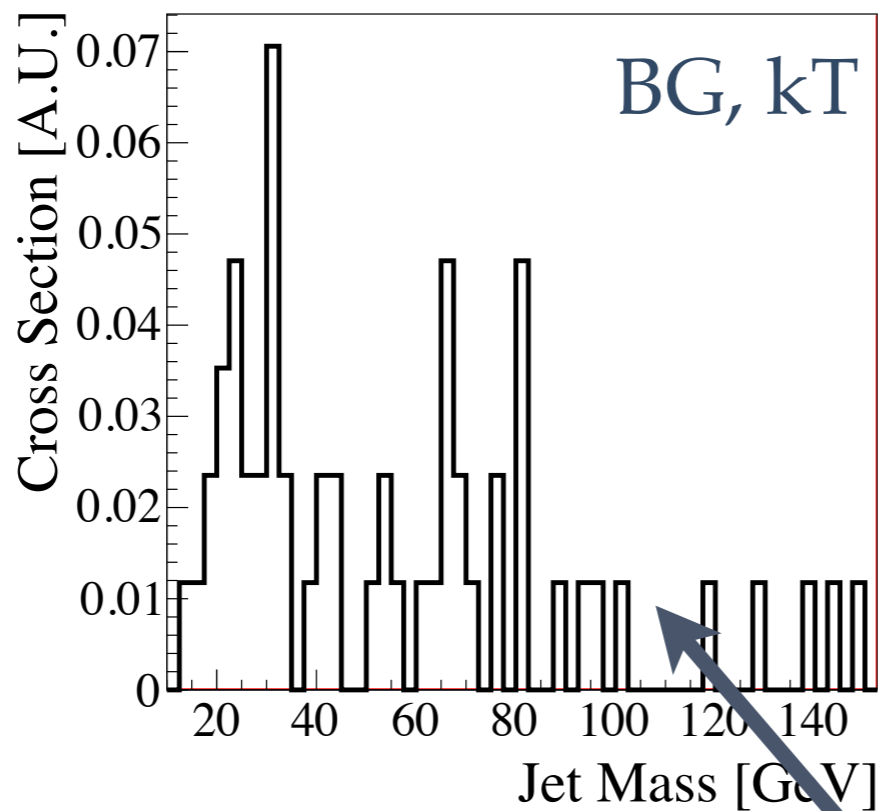
- * Signal = boosted W-jets, $p_T > 500$
- * BG = light QCD jets, $p_T > 500$
- * Measure the signal size in a bin (here 70-90 GeV) and compare it to the size of the BG fluctuations (Poisson stats included)
- * Need only $\sim 70\%$ the luminosity to have the same significance

$$S/\delta B \propto \sqrt{N}$$

Algorithm	$S/\delta(B)$	Relative Lumi Required
kT	4.9	1.00
Flat ($\alpha = 0$)	6.0	0.83
Qjets ($\alpha = 10^{-1}$)	6.3	0.69



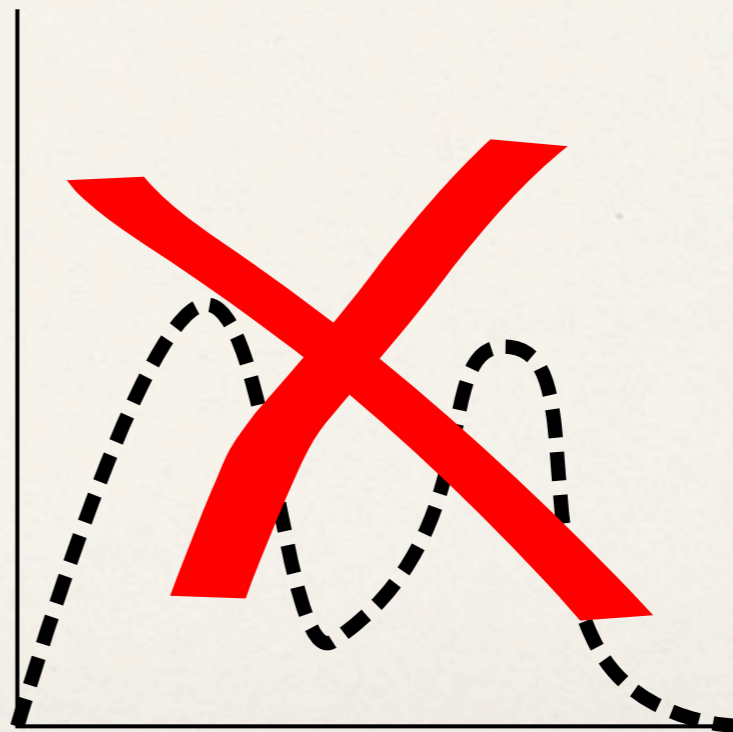
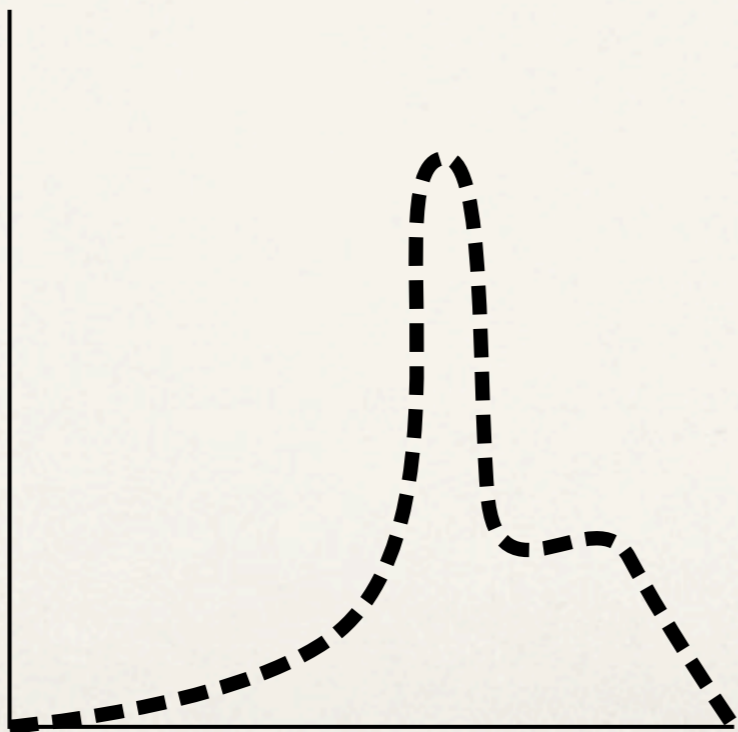
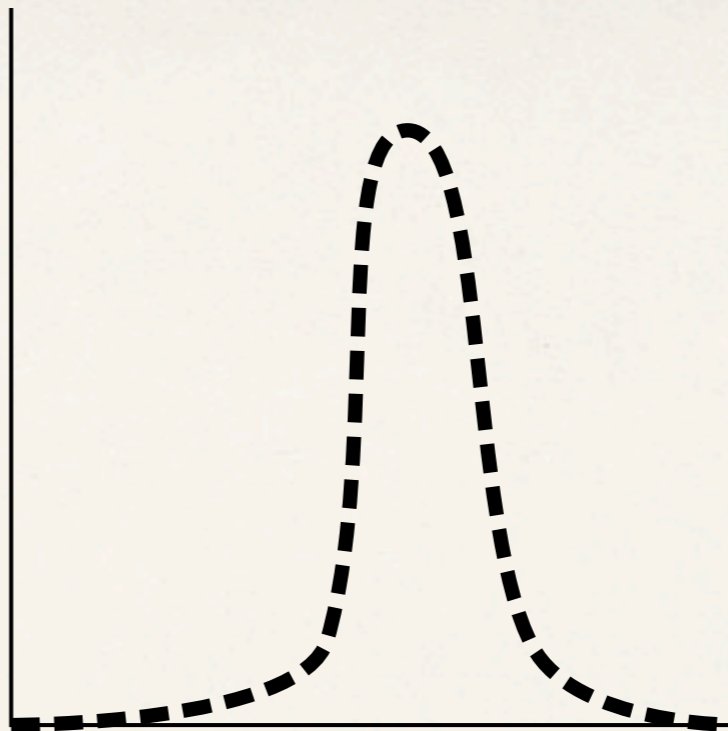
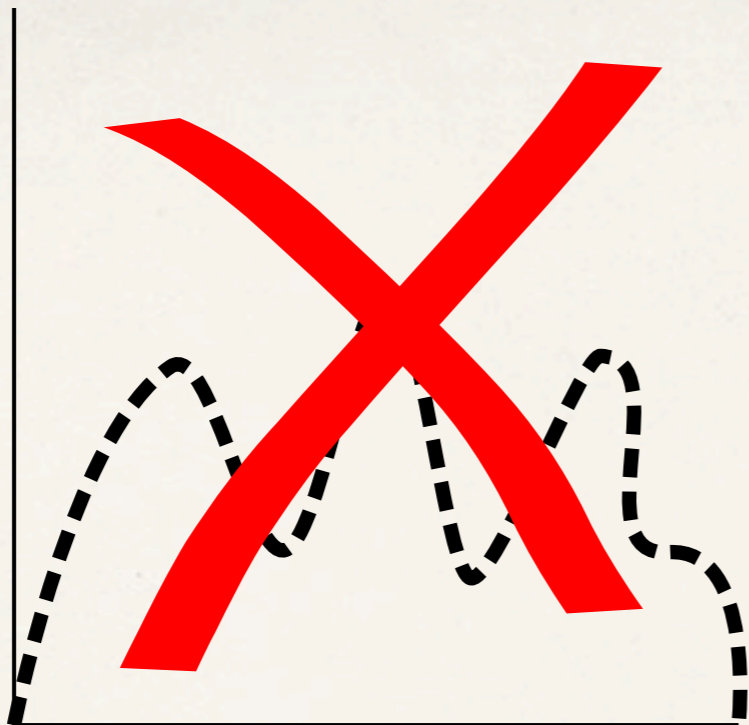
Same events



Same events

Signal vs. Background Discriminant

- ❖ When there's a "right answer" for a jet's mass, most of the trees tend to center around that value.
 - ❖ There's a "right answer" for the pruned mass of a boosted particle's jet, but not for a background QCD jet
- ❖ The width of a mass distribution serves as a good signal to background discriminant!

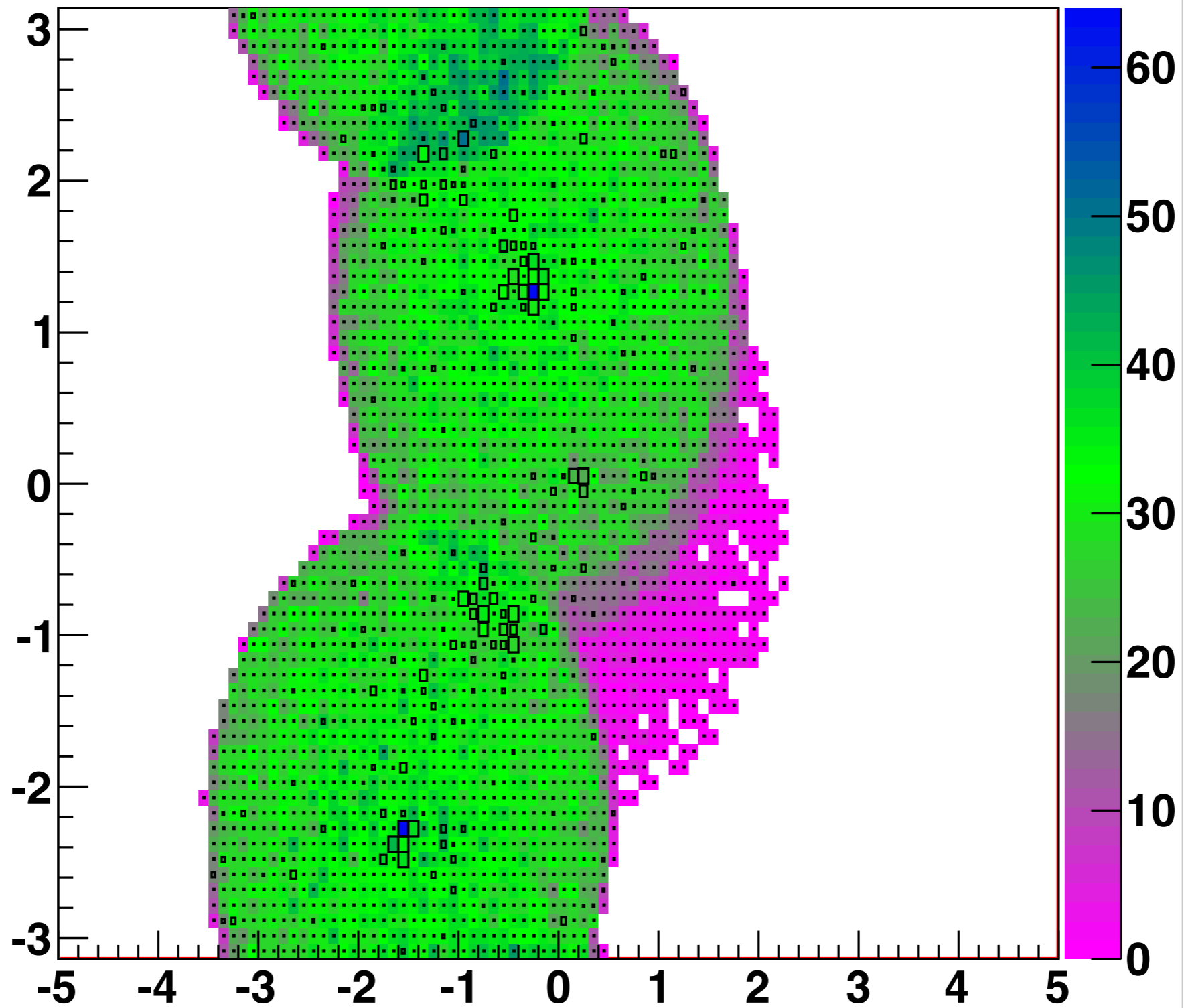


Generalize to a Jet Algorithm

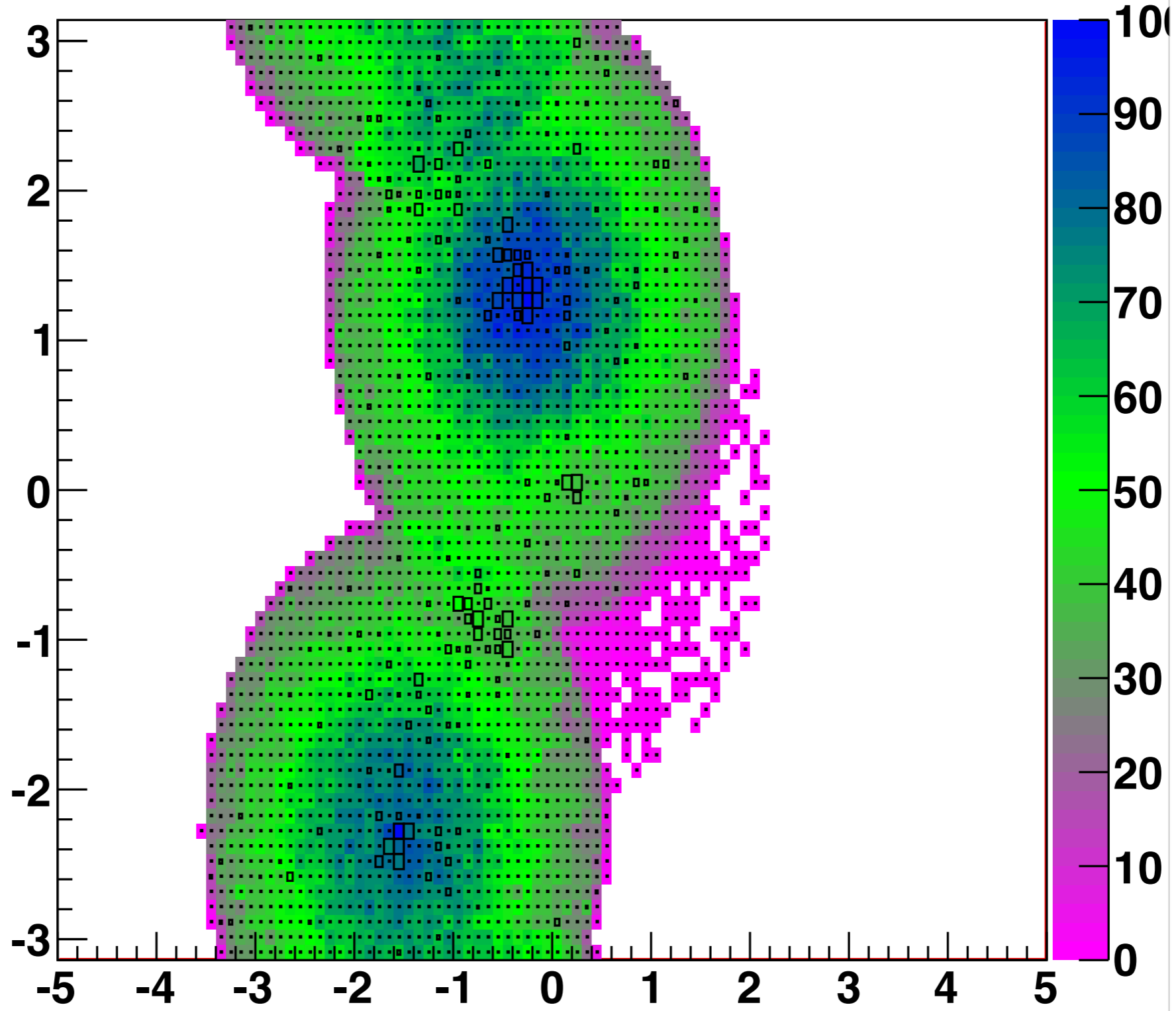
Qanti-kT

- ❖ **Work in progress**
- ❖ Take anti-kT and perturb around it as with Qjets
- ❖ Final state is now different
 - ❖ Different jet four-momenta
 - ❖ Different jet multiplicities

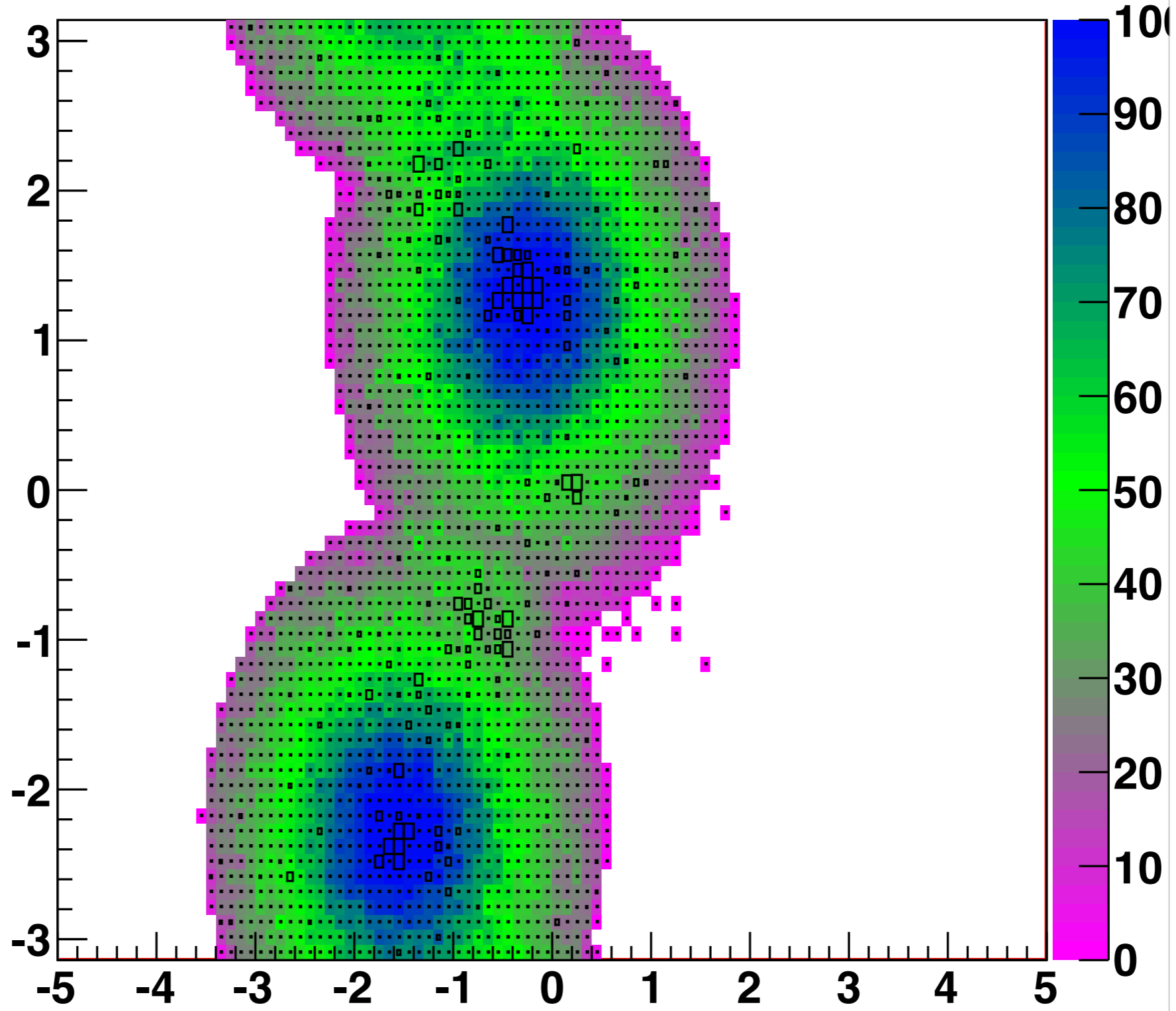
eta, phi vs frequency, pT, 1TeV scalar, alpha= 0.001 akt m12= 794.047



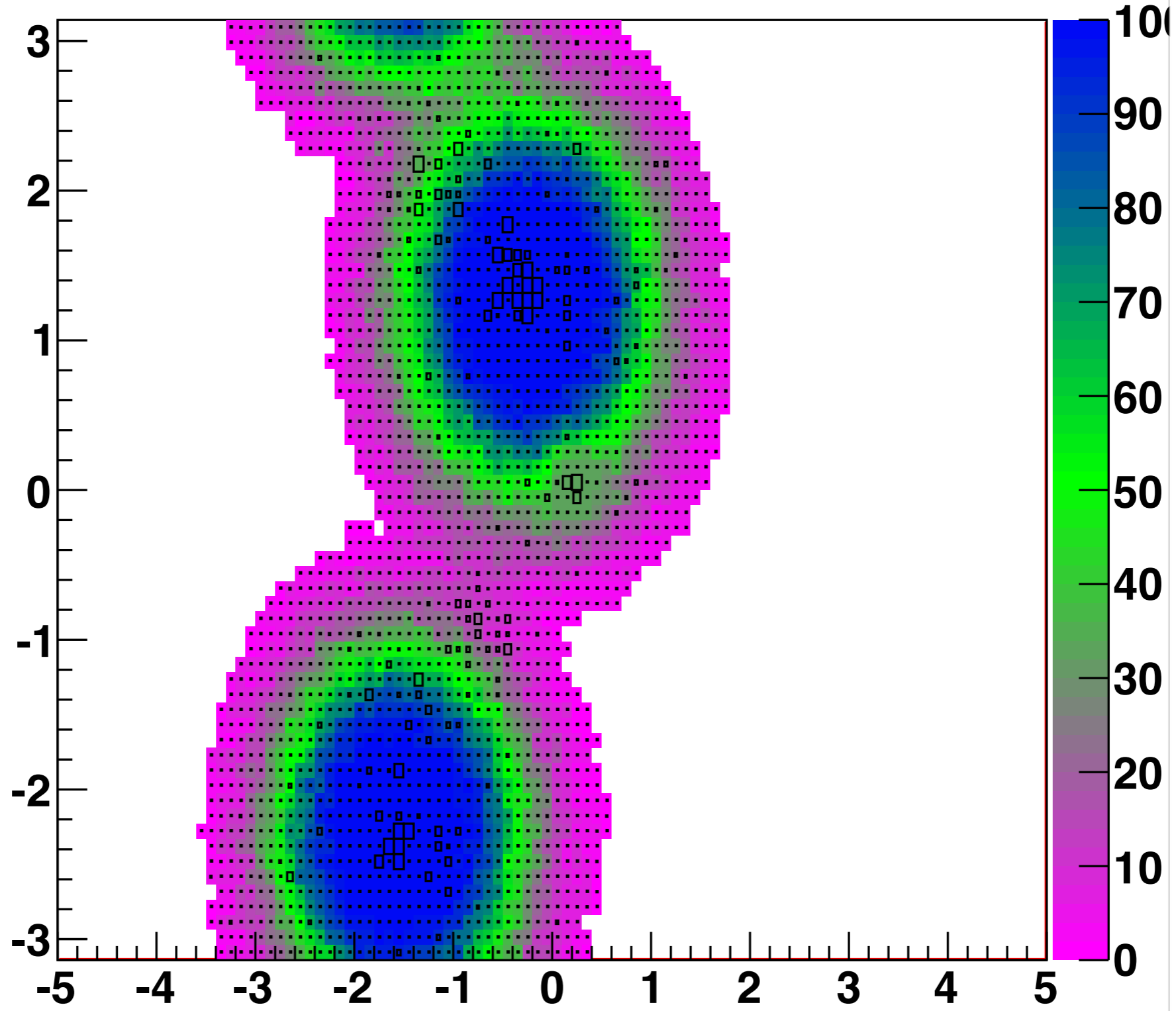
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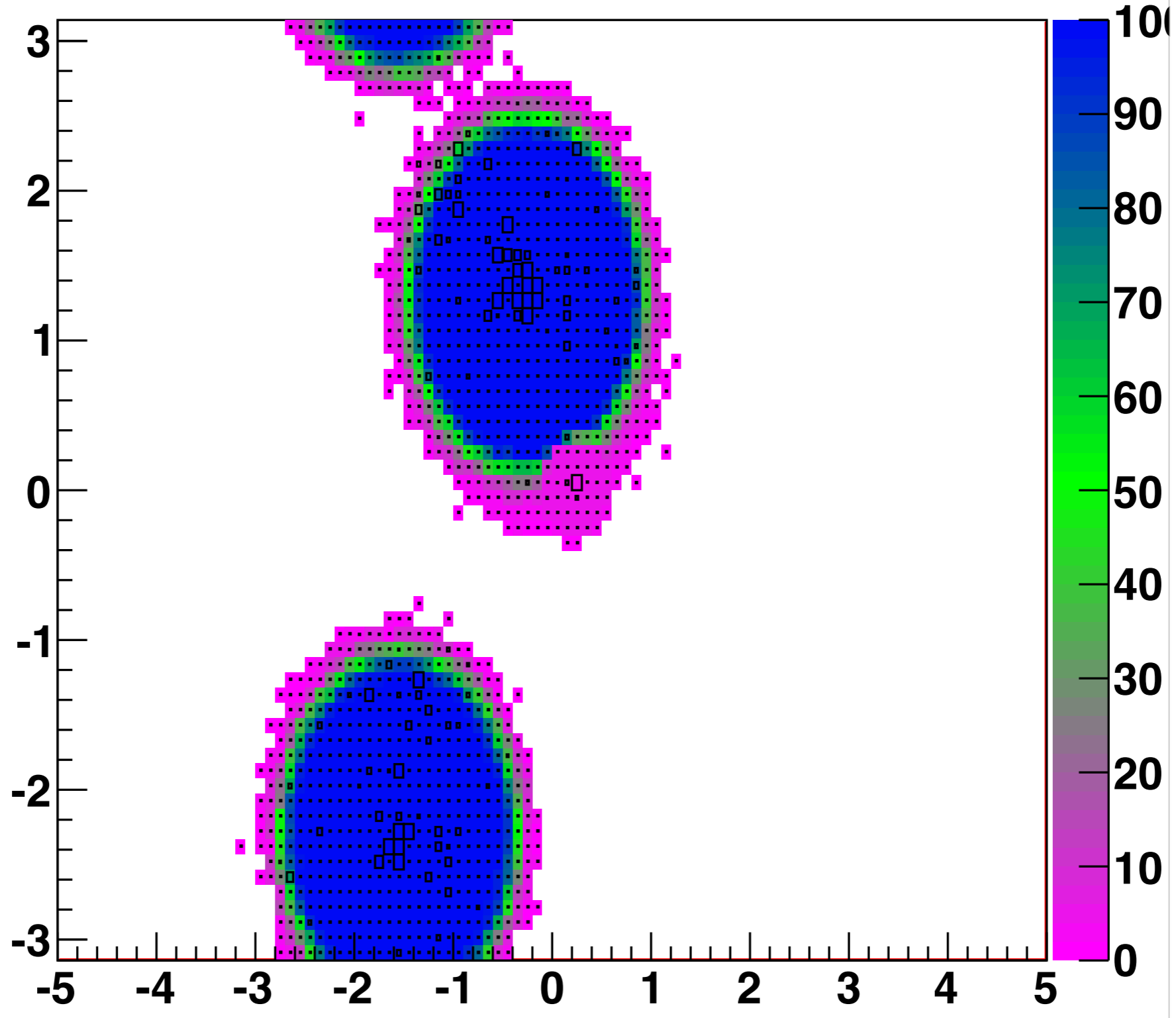
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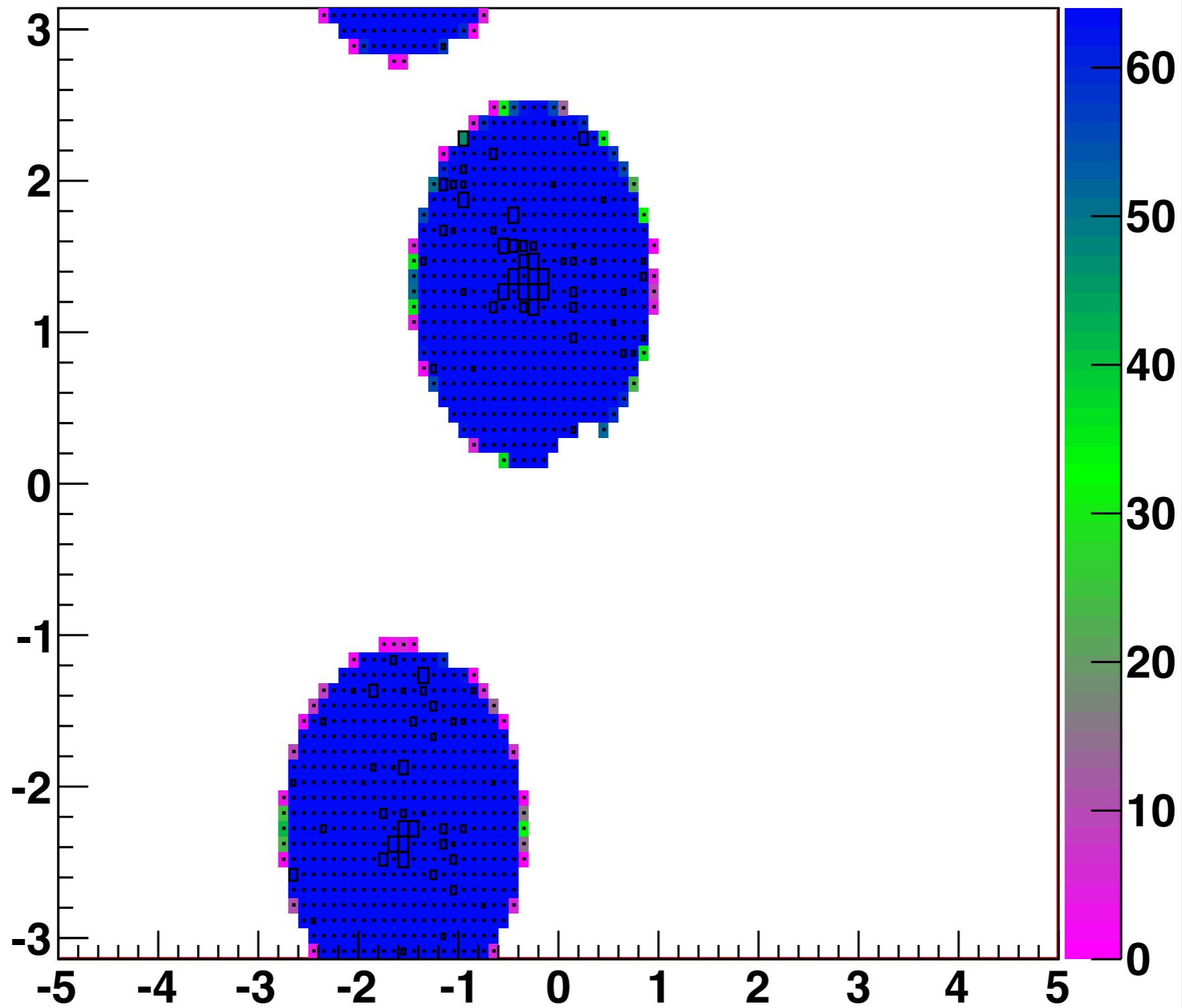
eta, phi vs frequency, pT, 1TeV scalar, alpha= 1 akt m12= 794.047



eta, phi vs frequency, pT, 1TeV scalar, alpha= 10 akt m12= 794.047



eta, phi vs frequency, pT, 1TeV scalar, alpha= 100 akt m12= 794.047



Conclusion

- ❖ When we use C/A or k_T to associate a tree with a jet this is really just our “best guess” for the showering history.
- ❖ Sometimes these two algorithms return very different answers for the event at hand.
 - ❖ By choosing, e.g. the k_T answer over the C/A one, we introduce randomness into the picture, and the statistics are degraded.
- ❖ We propose that all trees be considered, each with a set weight, and a distribution obtained for each event (rather than a single number).
 - ❖ The results obtained from this are much less susceptible to unwanted fluctuations: equivalent to a $\sim 2x$ increase in luminosity.