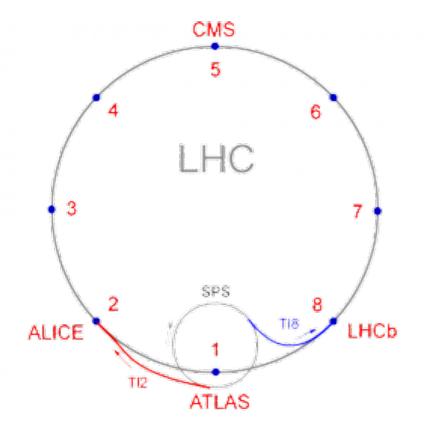
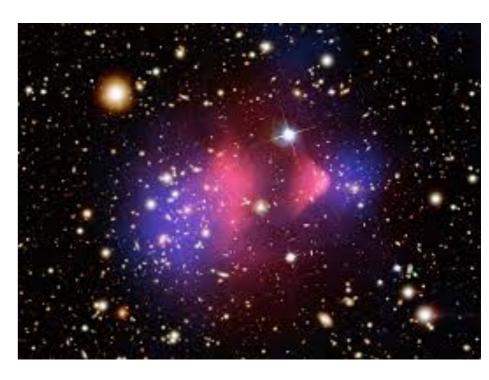


# Thoughts on MVA for Dark Matter Searches at the LHC



Jamie Gainer UH Manoa April 5, 2017





# Multivariate Analyses

An analysis using more than one variable

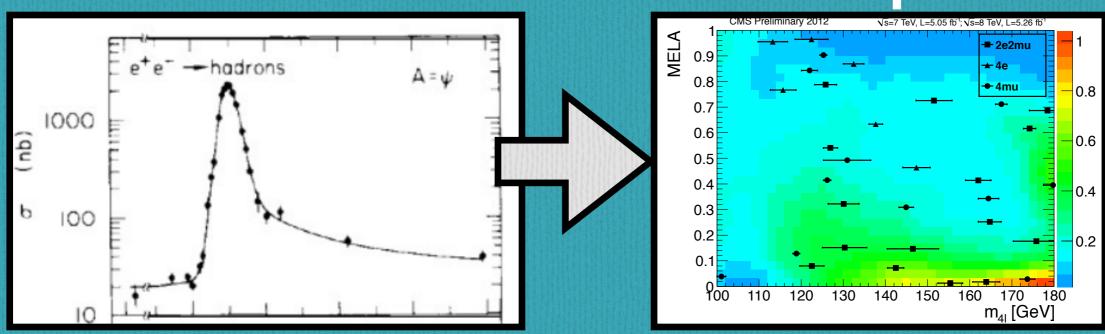


 Often refers to sophisticated/ powerful methods (neural nets, boosted decision trees, the matrix element method) which are contrasted with, e.g., binning data in a single variable

### **Revolution in Experiment!!!**



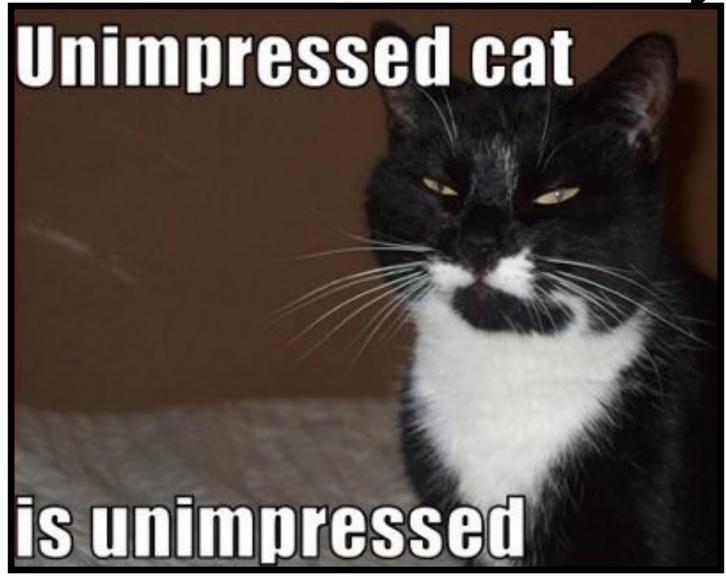
### Multivariate Methods are Now Ubiquitous



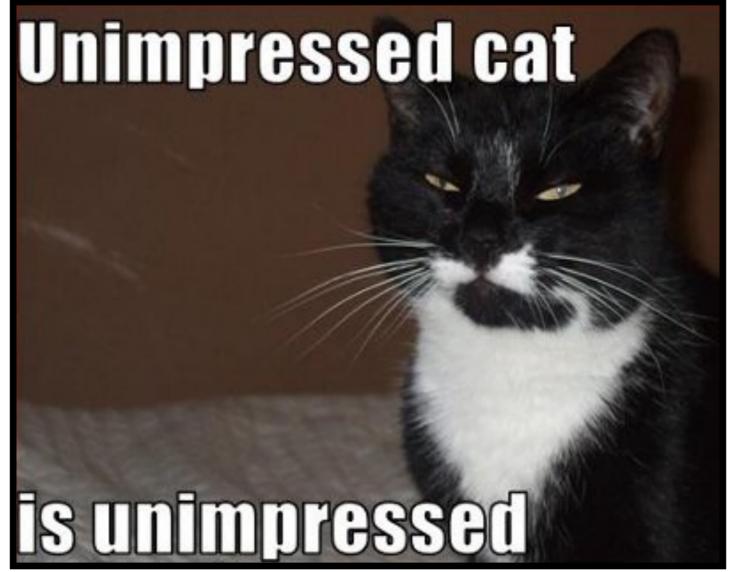
MVA information helpful even in discovering a narrow resonance in leptons!

# Less So in Theory...

Less So in Theory...



# Less So in Theory...



- I'm overstating the case. Plenty of work by theorists, especially on the Matrix Element Method.
- But not proportionate to the use of MVAs (especially neural nets and BDTs) in experiment

# Understanding the Physics that Drives the Sensitivity

- The real problem is that it is hard to understand the physics reasons for increases in sensitivity that come from neural nets (black boxes?)
- Obviously the experimentalists don't want us to run neural nets for them, especially since some of the power comes from sensitivity to detailed aspects of the detector response
- But can we use theory tools to understand the additional sensitivity from MVAs?
- Yes, in the Matrix Element Method



Pauli

### The Matrix Element Method

- Using the likelihood ~ probability
- Essentially the differential cross section normalized to the total cross section
- Once we take into account that what we measure in detectors isn't the four momentum of a parton (transfer functions)
- And we integrate over particles we don't see at all (like dark matter!)
- According to our friends Neyman and Pearson, the likelihood is an "optimal test statistic"

### The Matrix Element Method

- And it works great in Higgs physics, especially Higgs to fourleptons
- But also in top physics, B physics, some SUSY...
- The big thing is that since your MVA output is essentially a differential cross section, the underlying physics can be understood
  - Heavy Higgs to four-leptons: sensitivity from different helicity amplitudes in signal and background, i.e., sensitivity driven by angular momentum considerations, spin of Higgs, etc.
  - SM Higgs to four-leptons, sensitivity driven by different propagators in signal ( $ZZ^*$ ) and background ( $Z\gamma^*$ ) leading to different distribution of  $M_{Z2}$

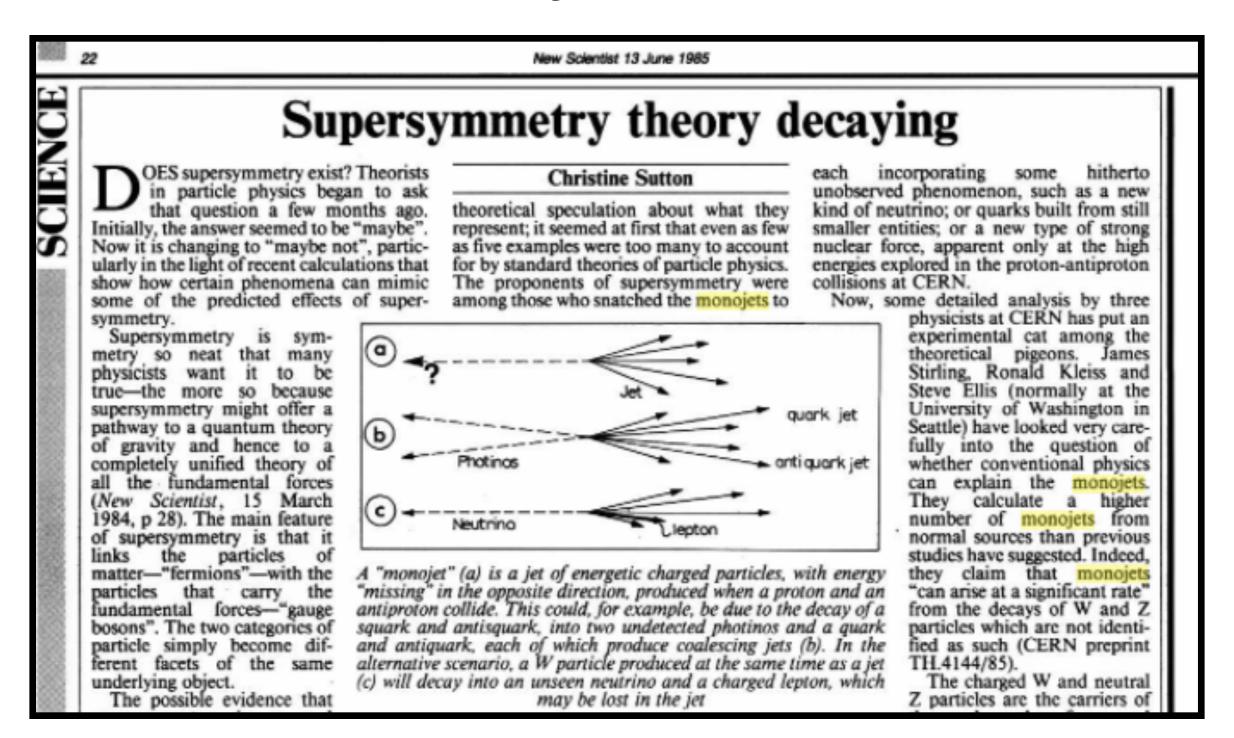
### The Matrix Element Method

- But it's really hard to calculate likelihoods, especially for
  - many particle final states,
  - many jets— transfer functions matter!
  - many missing particles— hard integrals
  - reducible backgrounds-- need to understand the detector

### MVA for Dark Matter Searches

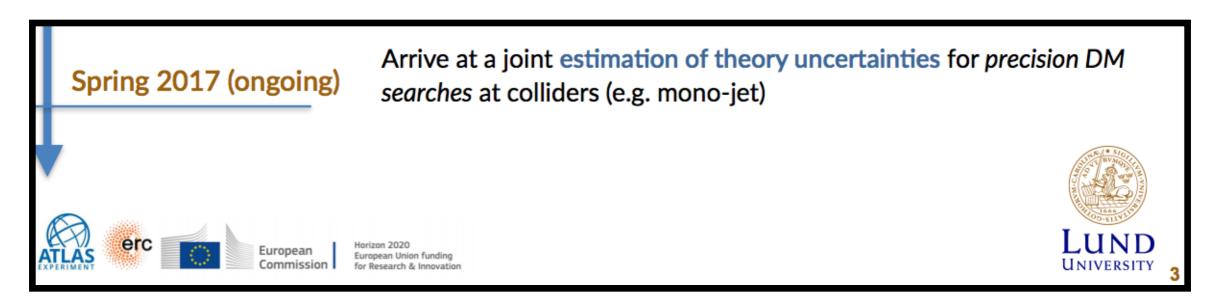
- I'd love to be wrong, but the Matrix Element Method does not seem great for dark matter searches at the LHC
- Always have invisible particles—integrals. (Even worse if we also have neutrinos or more than two DM particles in the fianl state)
- How to deal with jets: for four-leptons can focus on leptons.
  QCD effects enter analysis only through pT and eta of the four-lepton system
- What's the analogue of this for, e.g., monojets?
- Vast advances, but still significant dependence on detector modeling (which goes into the MC expressions that are reweighted).
   For example, when do you miss the lepton in W + jets?
- Many dark matter searches— but I'm going to focus on maybe the most dramatic example: monojets

## Monojets Then



 New Scientist article from June 13, 1985 describing the excitement over an "excess" of monojets in UA1 as well as Ellis, Kleiss, and Stirling's determination that SM backgrounds could explain the excess

### And Now



#### Motivation and goal for this focused effort:

- Ambitious program of LHC mono-X searches up to HL-LHC needs:
  - Understand impact of electroweak corrections
  - Reaching 1% uncertainties on theoretical modelling
  - Understand correlations between uncertainty sources

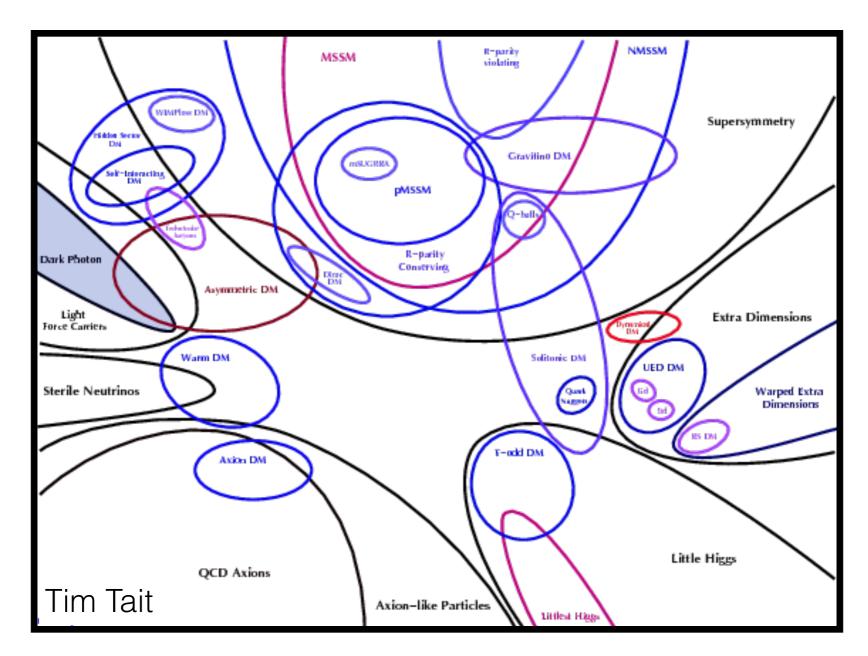
from Caterina Doglioni's talk on Monday

 See also Stefano Pozzorini's talk went into the details of the theoretical calculations needed to make this a reality

### And Now

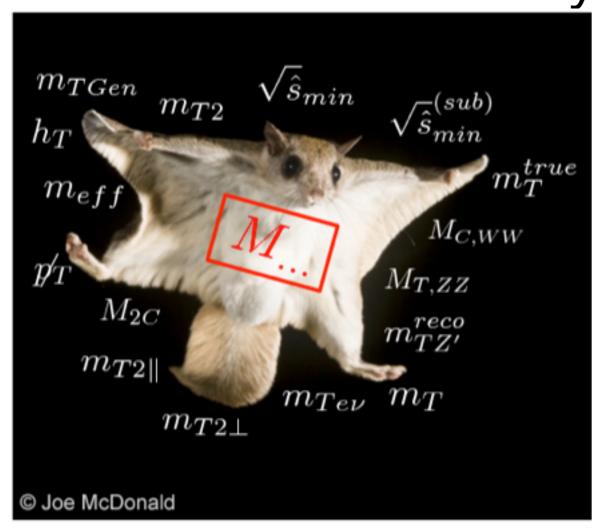
- I think a major theme of the monojet talks here is that we are entering the precision regime
- As confidence in experimental and theoretical modeling of backgrounds and signals grows, MVAs will play an increasing role
- How can we model-building/ testing theorists take this development into account in a useful way?
- As Doraval pointed out yesterday, reducing systematic errors in signal models can be challenging
- The challenge is heightened by the fact that there are...

# Many Models...

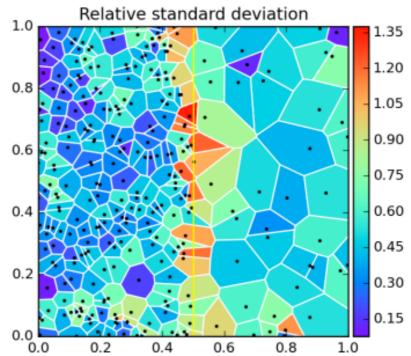


- Benchmarks essential
- But many possibilities, signatures
- We want to explore as many possibilities as possible, as well as possible

### and Many Variables...



(Barr, Khoo, Konar, Kong, Lester, Matchev, Park, 2011)



(Debnath, JG, Kilic, Kim, Matchev, and Yang, 2016)

Also old standbys like jet pT, missing ET, HT, sphericity, delta phi, etc.

- Many choices of variables (not all independent)
- Need to determine best variables to add: may depend on signal model
- Useful to be able to understand/ project experimental choices, and/or suggest good variables for new/ challenging models

# My Proposal: Quantile Binning

- My proposal is to consider the signal cross section in quantile bins for the background in the variable(s) of interest
- Quantile bins: construct bins so that each bin has the same number of events, or same cross section
- So we pick a variable, make bins in that variable so that each bin has the same background cross section, and then see what the signal cross section in each bin is
- (sort of an unboosted decision tree: we don't vary our bin limits to optimize sensitivity)

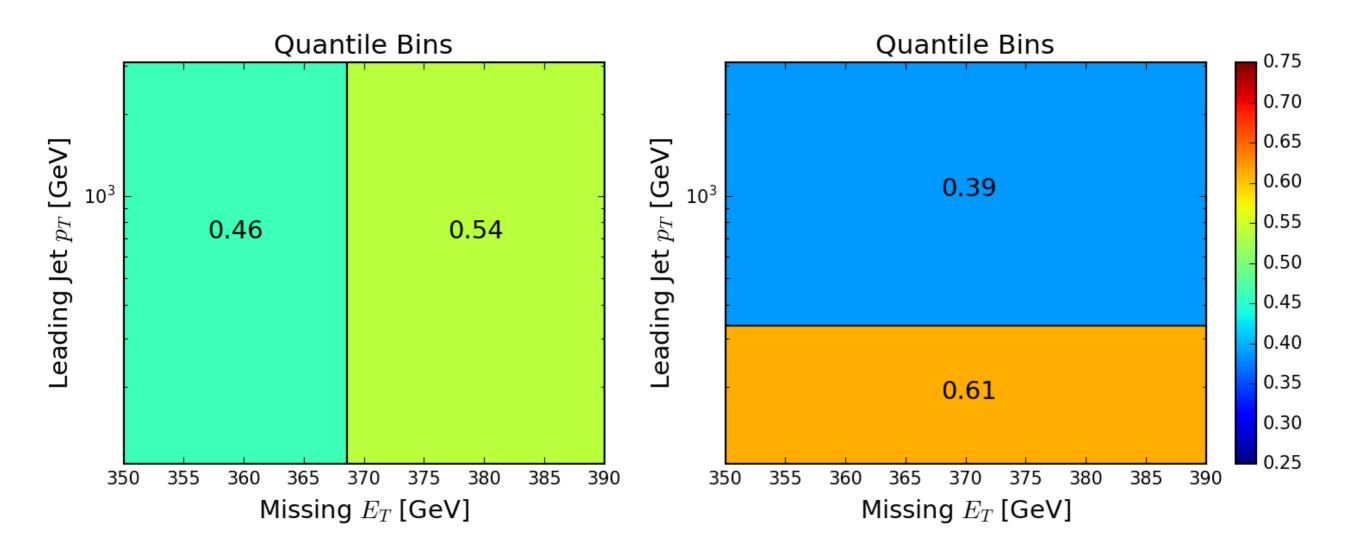
#### • Pros:

- easily investigate sensitivity of physics variables
- some systematics will (somewhat!) cancel, e.g. effects (detector or theory) common to signal and background
- Con: not as sensitive as using full distributions.
  - But this is okay: point is to let theorists determine good variables for studying new model points and project future sensitivities in a not-totally-crazy way.
     Experiments can then do better with neural nets, etc.

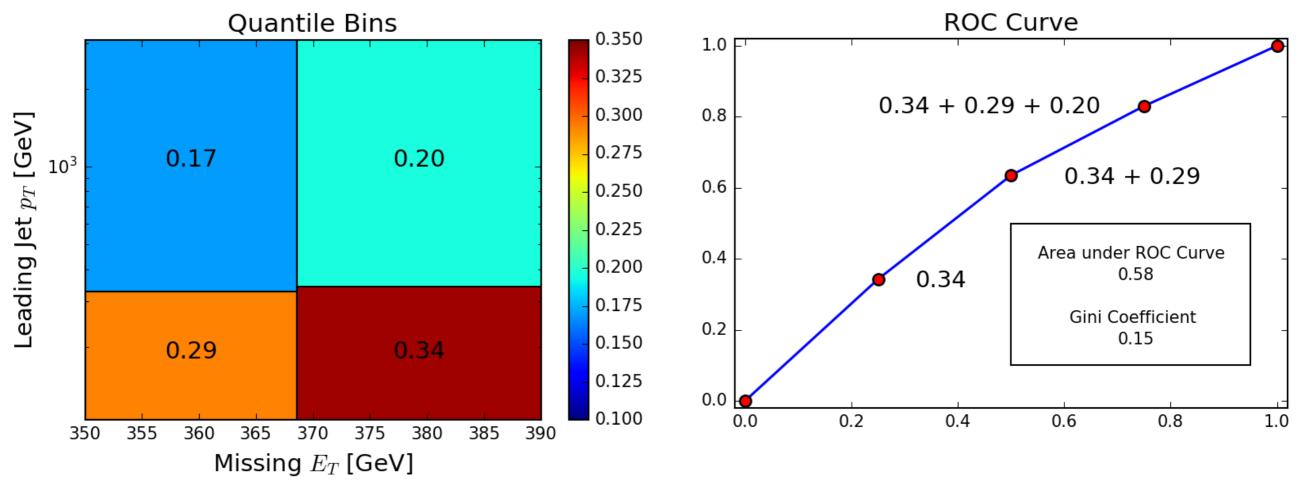
- See Osamu Jinnouchi's talk from Monday
- ATLAS most recent monojet search Phys. Rev. D94 (2016) 032005 (1604.07773) 3.2 fb<sup>-1</sup>
- CMS most recent monojet search PAS EXO-16-037 (1703.01651).
   12.9 fb<sup>-1</sup>
- Using the CMS search as the basis for my example
- My signal model will be a 600 GeV stop decaying to a 570 GeV neutralino + charm jet (no charm tagging, though)
- That analysis uses jet mass/ substructure to also do a mono-V search, but I am restricting my attention to the mono-jet search

$E_{\rm T}^{\rm miss}$ [GeV]	$Z(\nu\bar{\nu})$ +jets	$W(\ell \nu)$ +jets	Top quark	Dibosons	Other	Total bkg.	Observed
200-230	$71300 \pm 2200$	$54600 \pm 2300$	$2140 \pm 320$	$1320 \pm 220$	$2470 \pm 310$	$132100 \pm 4000$	140642
230-260	$39500 \pm 1300$	$27500 \pm 1200$	$1060 \pm 160$	$790 \pm 130$	$1090 \pm 130$	$69900 \pm 2200$	73114
260-290	$21900 \pm 670$	$13600 \pm 550$	$440 \pm 65$	$364 \pm 61$	$498 \pm 65$	$36800 \pm 1100$	38321
290-320	$12900 \pm 400$	$7300 \pm 290$	$210 \pm 31$	$235 \pm 40$	$216 \pm 30$	$20780 \pm 630$	21417
320-350	$8000 \pm 280$	$4000 \pm 170$	$107 \pm 16$	$145 \pm 24$	$124\pm18$	$12340 \pm 400$	12525
350-390	$6100 \pm 220$	$2800 \pm 130$	$74 \pm 11$	$111 \pm 19$	$87 \pm 13$	$9160 \pm 320$	9515
390-430	$3500 \pm 160$	$1434 \pm 66$	$30.1 \pm 4.5$	$58.4 \pm 9.9$	$33.4 \pm 5.3$	$5100 \pm 200$	5174
430-470	$2100 \pm 98$	$816 \pm 37$	$16.6 \pm 2.5$	$42.4 \pm 7.1$	$16.3 \pm 2.7$	$3000 \pm 120$	2947
470-510	$1300 \pm 66$	$450 \pm 20$	$7.4 \pm 1.1$	$24.6 \pm 4.1$	$9.6 \pm 1.6$	$1763 \pm 79$	1777
510-550	$735 \pm 39$	$266 \pm 13$	$5.2 \pm 0.8$	$18.5 \pm 3.1$	$7.0 \pm 1.3$	$1032 \pm 48$	1021
550-590	$513 \pm 31$	$152 \pm 8$	$2.4 \pm 0.4$	$13.5 \pm 2.3$	$1.1 \pm 0.3$	$683 \pm 37$	694
590-640	$419 \pm 23$	$120 \pm 6$	$1.5 \pm 0.2$	$10.6 \pm 1.8$	$2.1 \pm 0.4$	$554 \pm 28$	554
640-690	$246 \pm 16$	$62.8 \pm 3.8$	$1.3 \pm 0.2$	$11.4 \pm 1.9$	$1.0 \pm 0.2$	$322 \pm 19$	339
690-740	$139 \pm 11$	$34.2 \pm 2.4$	$0.6 \pm 0.1$	$4.2 \pm 0.7$	$0.20 \pm 0.07$	$178 \pm 13$	196
740-790	$97.2 \pm 7.2$	$22.7 \pm 1.7$	$0.22 \pm 0.03$	$1.4 \pm 0.2$	$0.63 \pm 0.12$	$122 \pm 8$	123
790-840	$59.8 \pm 5.8$	$12.9 \pm 1.2$	$0.13 \pm 0.02$	$1.5 \pm 0.3$	$0.05 \pm 0.02$	$74.5 \pm 6.6$	80
840-900	$64.3 \pm 6.4$	$12.3 \pm 1.1$	$0.24 \pm 0.04$	$0.92 \pm 0.1$	$0.03 \pm 0.01$	$77.8 \pm 7.2$	68
900-960	$31.5 \pm 4.3$	$6.0 \pm 0.7$	$0.21 \pm 0.03$	$0.74 \pm 0.1$	$0.01 \pm 0.01$	$38.4 \pm 4.8$	37
960-1020	$20.8 \pm 3.0$	$3.4 \pm 0.5$	_	$0.94 \pm 0.2$	$0.01\pm0.01$	$25.1 \pm 3.4$	23
1020-1090	$16.3 \pm 2.6$	$3.1 \pm 0.5$	$0.04 \pm 0.01$	$1.6 \pm 0.3$	$0.01 \pm 0.01$	$21.1 \pm 3.0$	12
1090-1160	$8.1 \pm 1.8$	$1.3 \pm 0.3$	_	_	_	$9.4 \pm 1.9$	7
>1160	$18.6 \pm 2.7$	$2.7 \pm 0.4$	_	$1.3 \pm 0.2$	_	$22.6 \pm 3.0$	26

We're looking at a single missing E<sub>T</sub> bin



 Within our bin, dividing into quantile bins in leading jet pT is more useful than a further subdivision in Missing E<sub>T</sub>



- Here we consider quantile bins in  $p_T$  and MET, slight increase in sensitivity over the two  $p_T$  quantile bin analysis
- Simple translation between signal fractions in quantile bins and ROC curves: common measure of the sensitivity of analyses

# Conclusions

- MVA are a huge part of experimental searches
  - We theorists need to respond to this development in a useful way
- Simple MVAs like quantile binning provide a simple and robust framework for evaluating variables and communicating with experiment
- May be especially useful in the context of dark matter searches