

# Using Selection Criteria to Optimize Analysis in High Energy Physics

Comparing Methods to Find New Particles

Chris Davis\*, Dr. David Toback, Daniel Cruz

Texas A&M University

Dr. Joel Walker, Jacob Hill

Sam Houston State University



# Outline

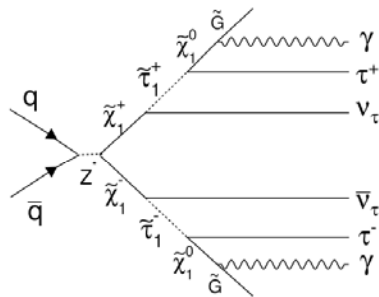
- ▶ Overview
  - Motivation for using selection criteria to find new particles
- ▶ Using Selection Criteria (Cuts)
- ▶ Comparing Different Approaches
- ▶ Results

# Motivation

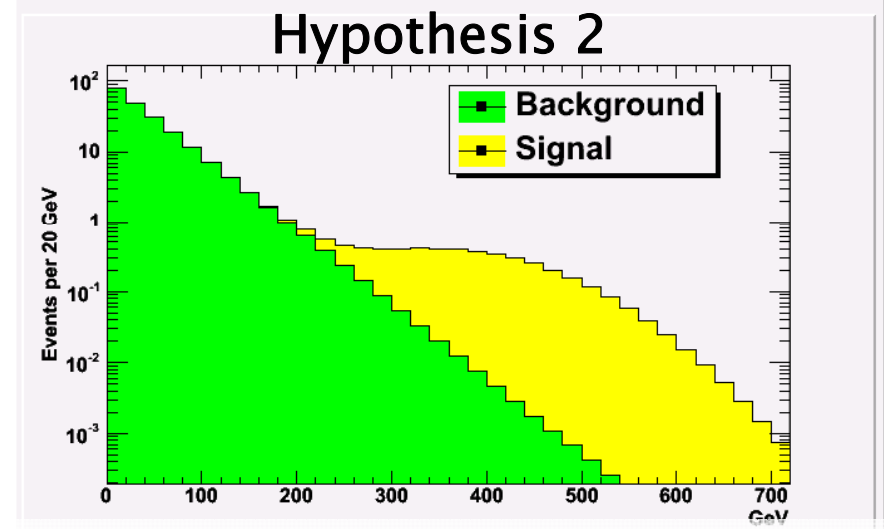
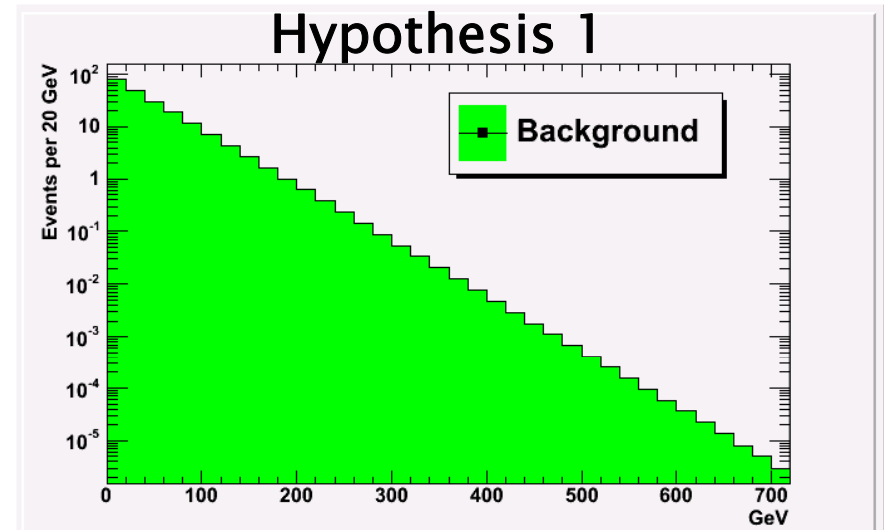
- ▶ We want to be more sensitive to new particles in High Energy Physics
- ▶ Huge amount of collisions at colliders such as LHC means lots of data to look through
- ▶ Many ways to look for new particles
  - However, most are dominated by Standard Model particle “backgrounds”
- ▶ In some places, new particle “signal” dominates the background
- ▶ Using selection criteria allows us to be the most sensitive to new particles in these regions

# The Data We Used

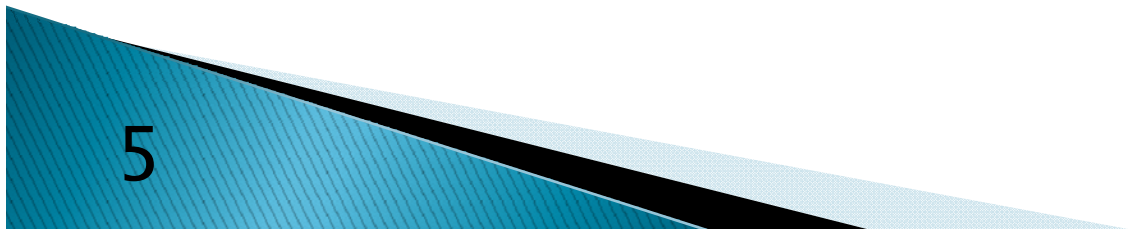
- ▶ Data is from a diphoton search for supersymmetry at Fermilab<sup>1</sup>



- ▶ A typical, simple search involves counting
  1. Number of Background events expected
  2. Number of Signal events expected
  3. How many events are observed in the experiment
- ▶ Add up observed events to determine which hypothesis is more consistent with data

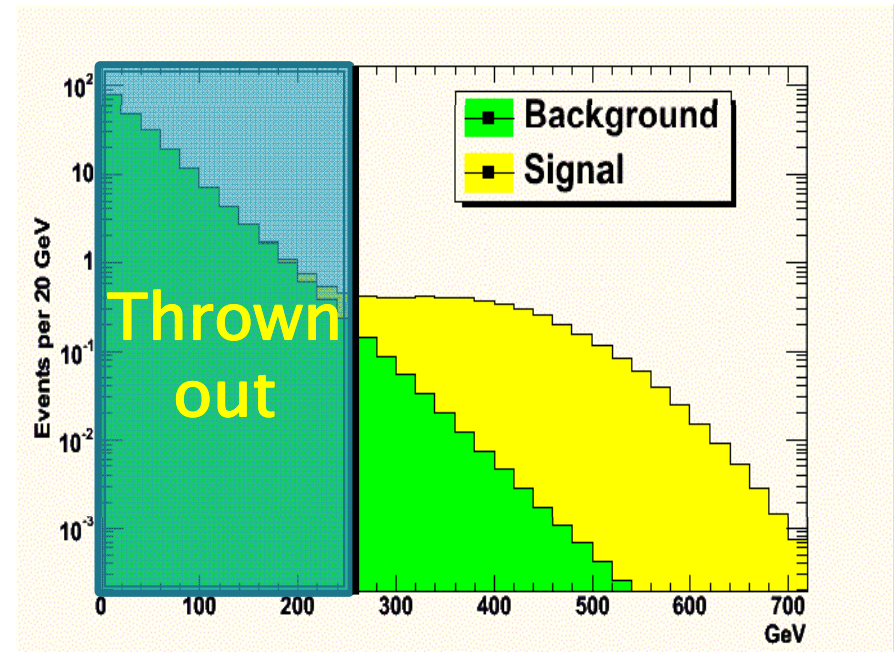


# Using Selection Criteria (Cuts)



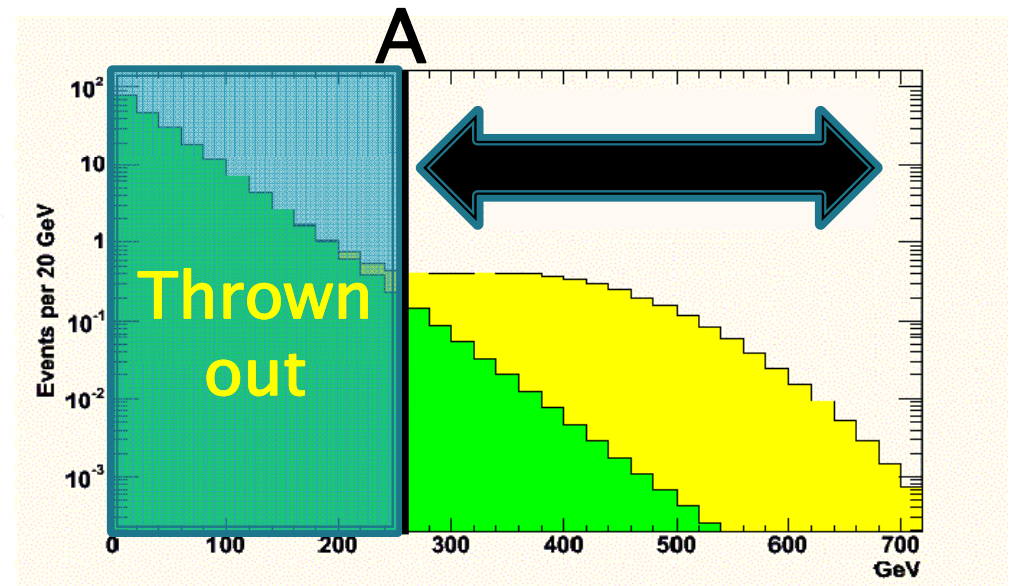
# Selection Criteria

- ▶ Selection criteria are used to optimize searches
  - Select only events that pass certain criteria
  - New particles easily pass them
  - Few Background events also pass



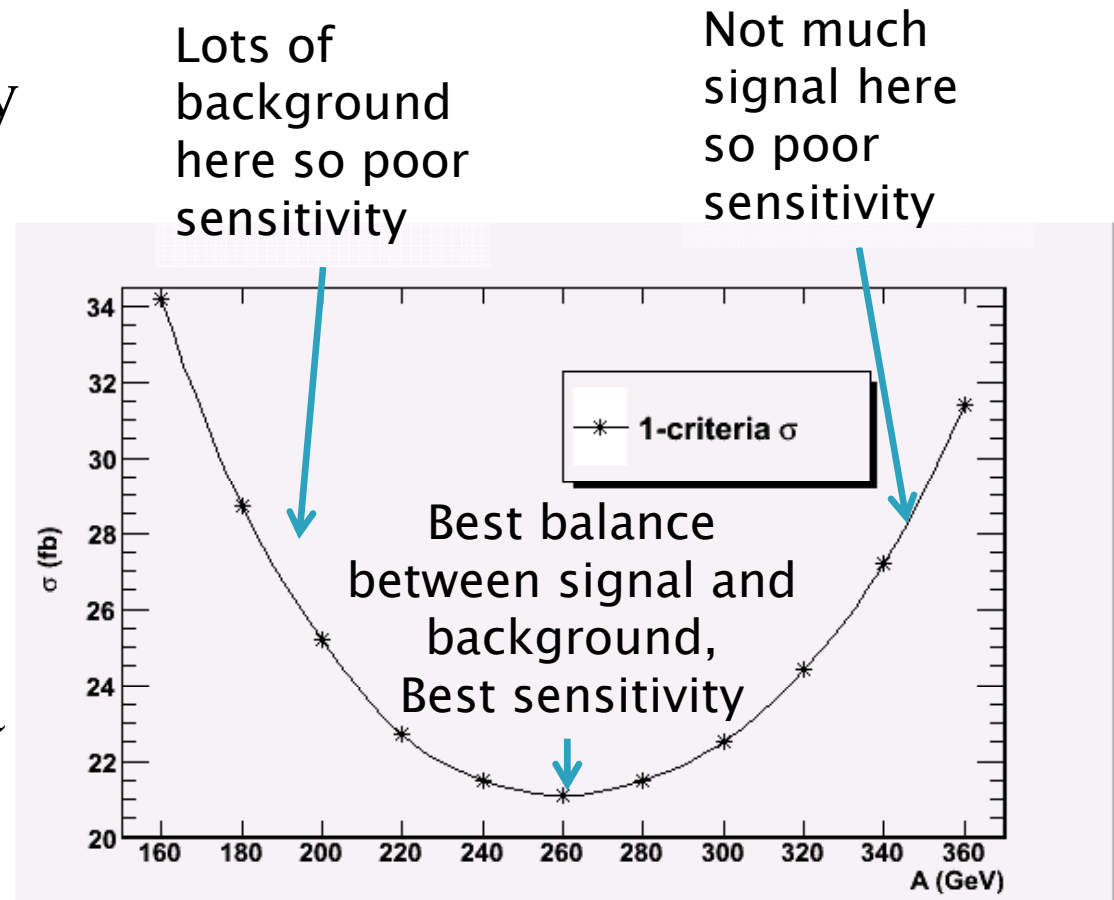
# Single Selection Criterion

- ▶ Creates a single set of data starting at A
  - Throw out all events that do not pass our criterion, count events from  $A \rightarrow \infty$ 
    - Lowering the value of A adds in more background, more signal
    - Raising value of A takes out background, but also signal
  - We look at data that is most sensitive to signal



# Single Criterion in Experiment

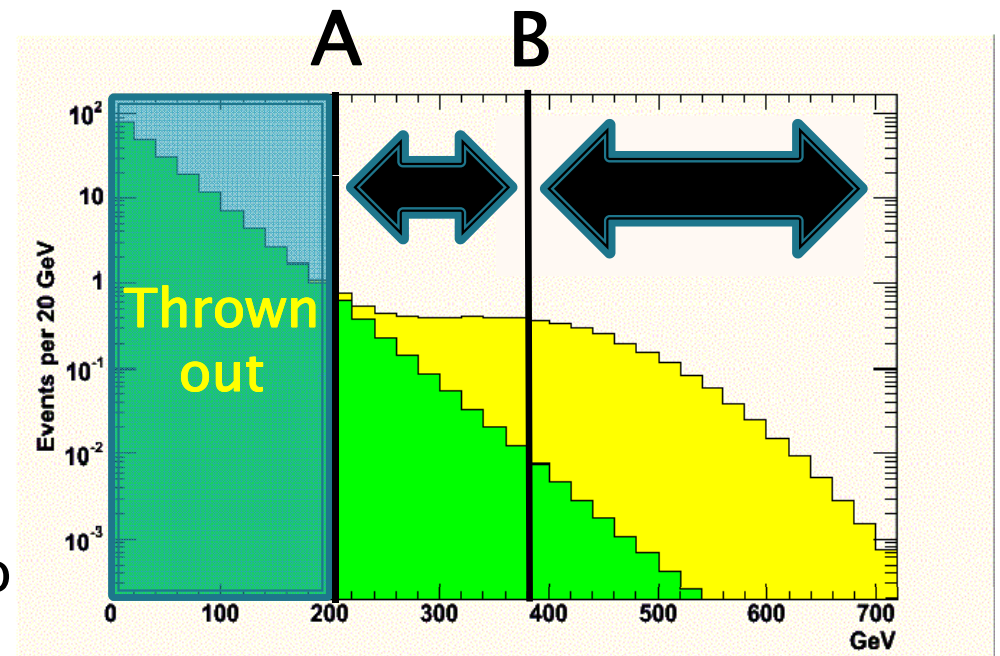
- ▶ Cross section,  $\sigma$ , is a measure of sensitivity
  - Lower  $\sigma$ , better sensitivity
  - Higher  $\sigma$ , worse sensitivity
- ▶ Vary  $A$  to optimize sensitivity
- ▶ Can we get better sensitivity by doing a more sophisticated analysis?





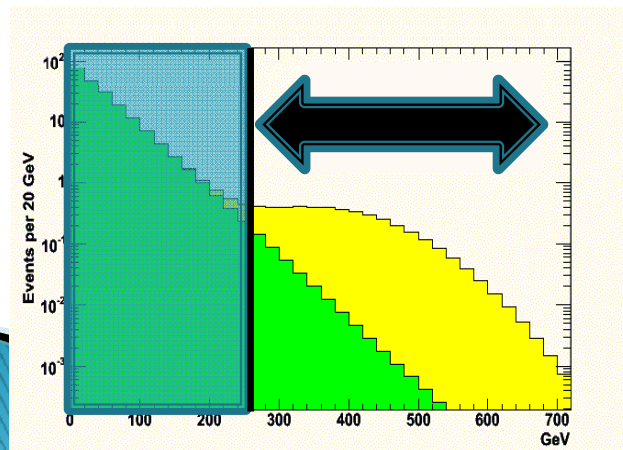
# Two Selection Criteria

- ▶ Data is placed into two sets
  - Count events from  $A \rightarrow B$  and  $B \rightarrow \infty$
- ▶ This is a more sophisticated analysis
  - Does being more sophisticated translate to being more sensitive?
  - Systematic errors can be introduced, we'll deal with the simplest case without them in this talk

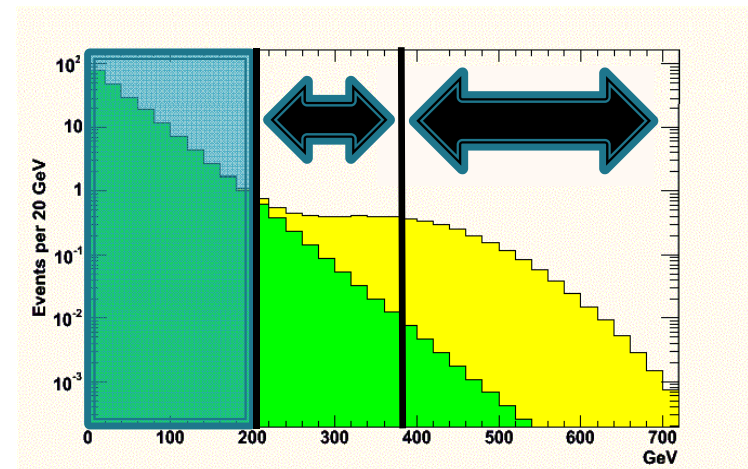


# Main Question

- ▶ Is it better to do one or two separate sets of independent criteria?
  - If we use two selection criteria, can we become more sensitive to new particles?.....Yes, will show!
  - Is using two selection criteria always more sensitive than using a single selection criterion?.....Surprisingly no, will show!

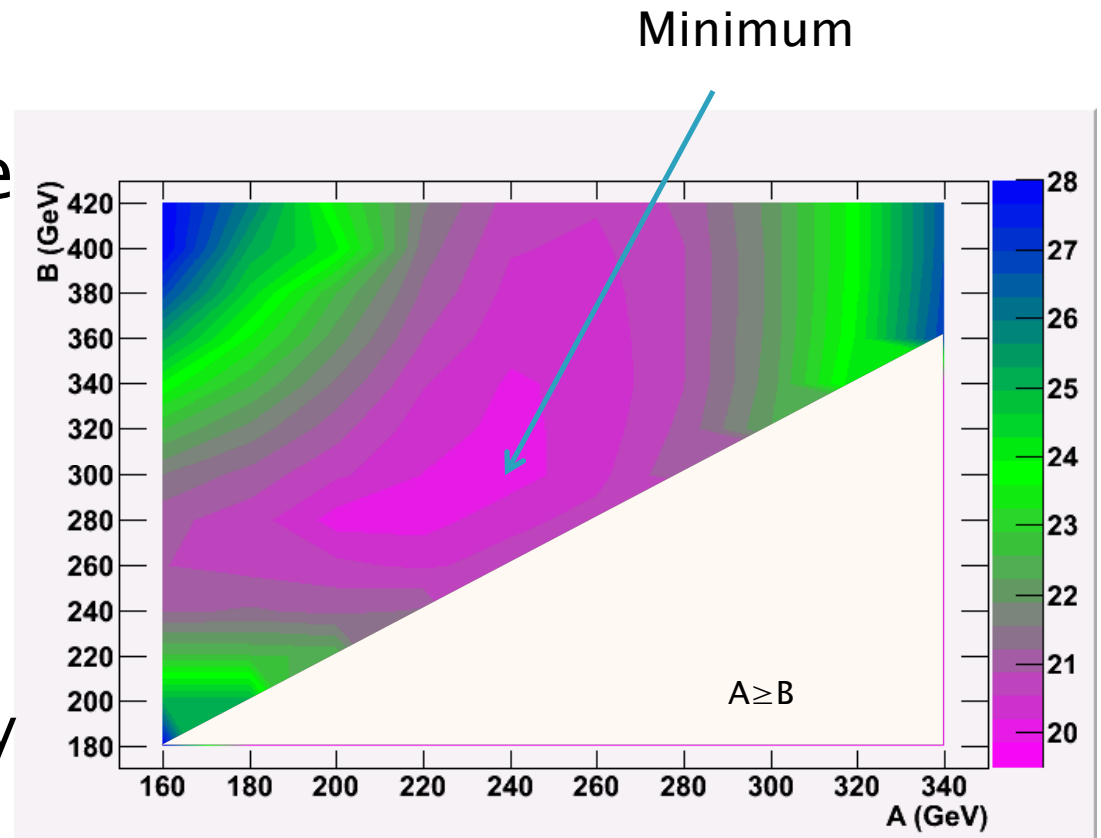


OR



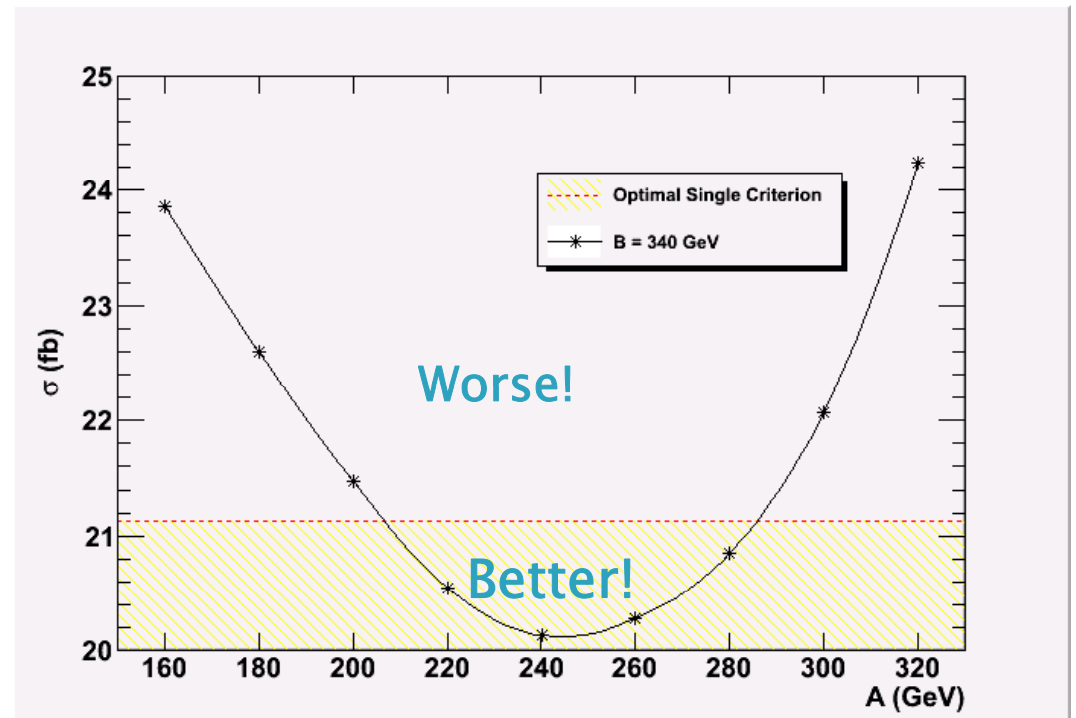
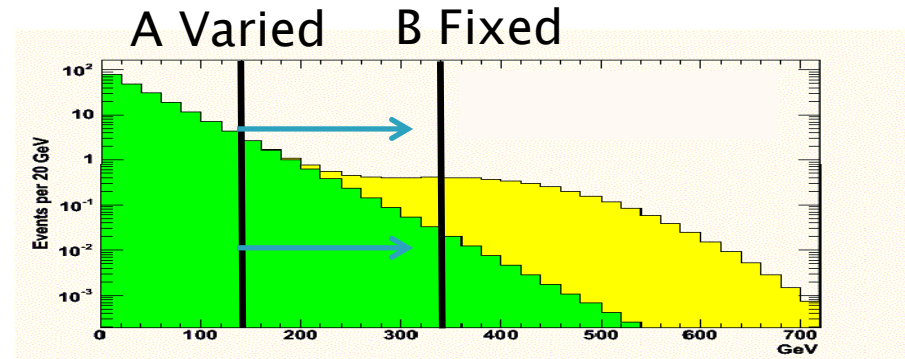
# Two Selection Criteria in Experiment

- ▶ Optimal criteria give lower  $\sigma$  than the optimal single criterion
  - $\approx 5\%$  less in this particular experiment
    - More sensitive!
  - Varying A and B to optimize sensitivity



# Can it be Worse?

- ▶ Look at two criteria in one dimension to compare with an optimal single criterion
  - Arbitrarily fix B and vary A
  - There is a region where two criteria are better
  - However, also regions where two criteria are worse!

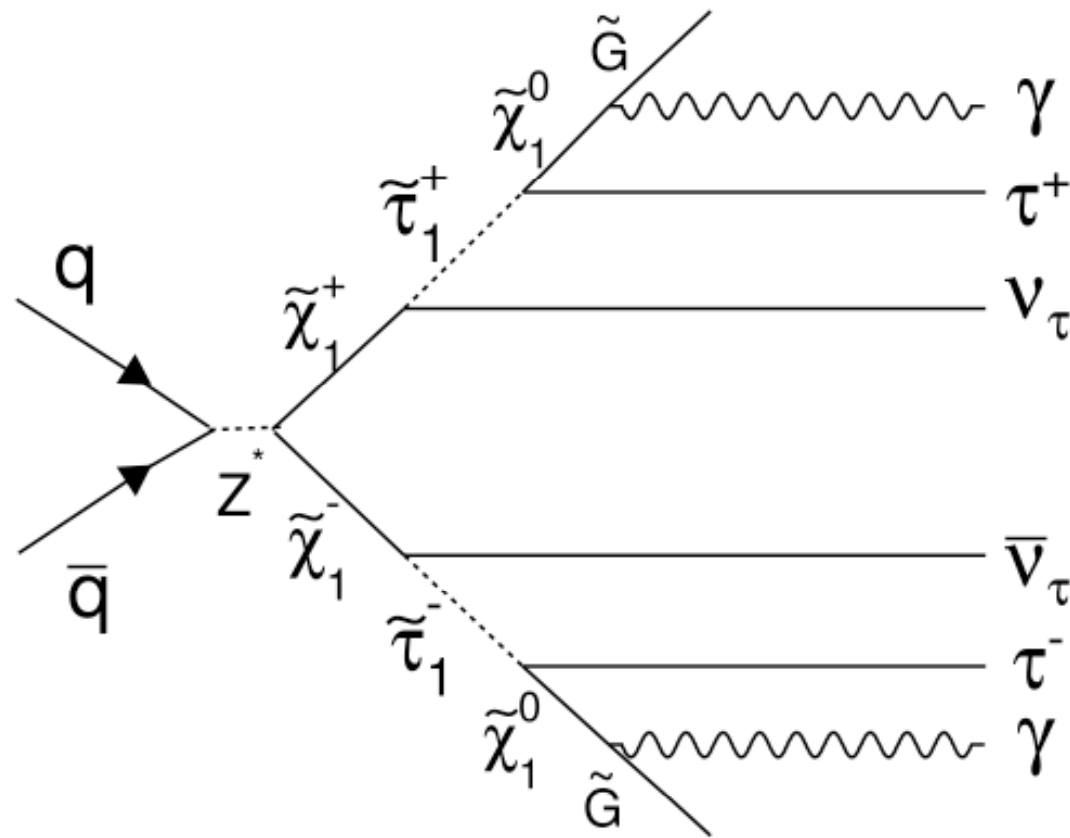


# Conclusions

- ▶ Our sensitivity to new particles is improved when we use selection criteria
- ▶ We have determined that
  - Two criteria **CAN** be better than a single, optimized criterion
    - Need to look for a minimum!
  - Two criteria **CAN ALSO** give a worse result if used incorrectly

# BACKUP SLIDES

# Signal Events



# Limit Calculator

- ▶ Example One-Cut input

- 160
- 1
- -1 2.59 .0790 .1218 4.251 .3188
- 0

- ▶ Example Two-Cut Input

- 360
- 2
- -1 2.59 .0399 .1218 4.218 .3188
- -1 2.59 .0391 .1218 .0326 .3188
- 0



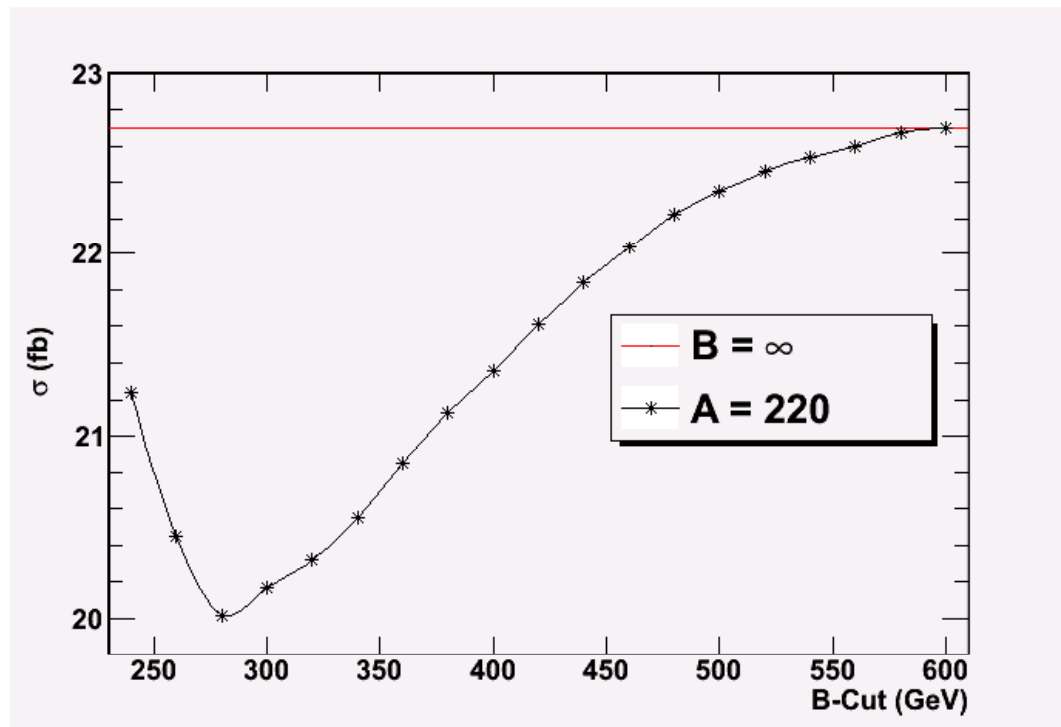
# Expected Cross Sections

- ▶  $N_{\text{events}} = \text{Luminosity} * \sigma_{\text{production}} * \text{Acceptance}$
- ▶ Find 95% confidence limits on  $\sigma_{\text{production}}$ 
  - Taking cuts allows us to optimize expected  $\sigma$
  - $\sigma_{\text{expected}}^{95} = \sum_{N_{\text{obs}}=0} \text{Poisson}(N_{\text{obs}}) * \sigma^{95}(N_{\text{obs}})$
- ▶ Used improved Limit Calculating program<sup>1</sup>

1. Developed by Dr. Joel Walker, Sam Houston State University

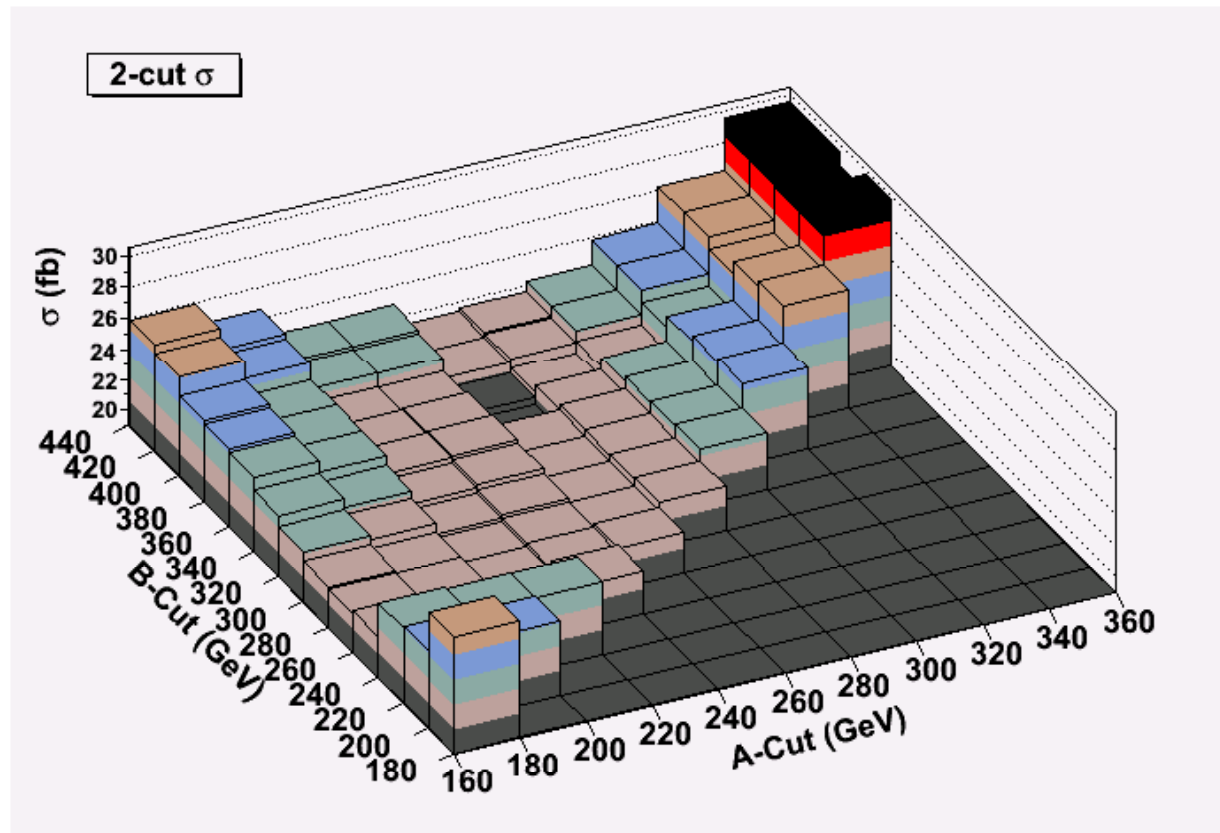
# Splitting Single Cut into Two

- ▶ If you take a single cut and place a B cut in it, you will always improve your sensitivity
  - Possibly not much better, but never worse



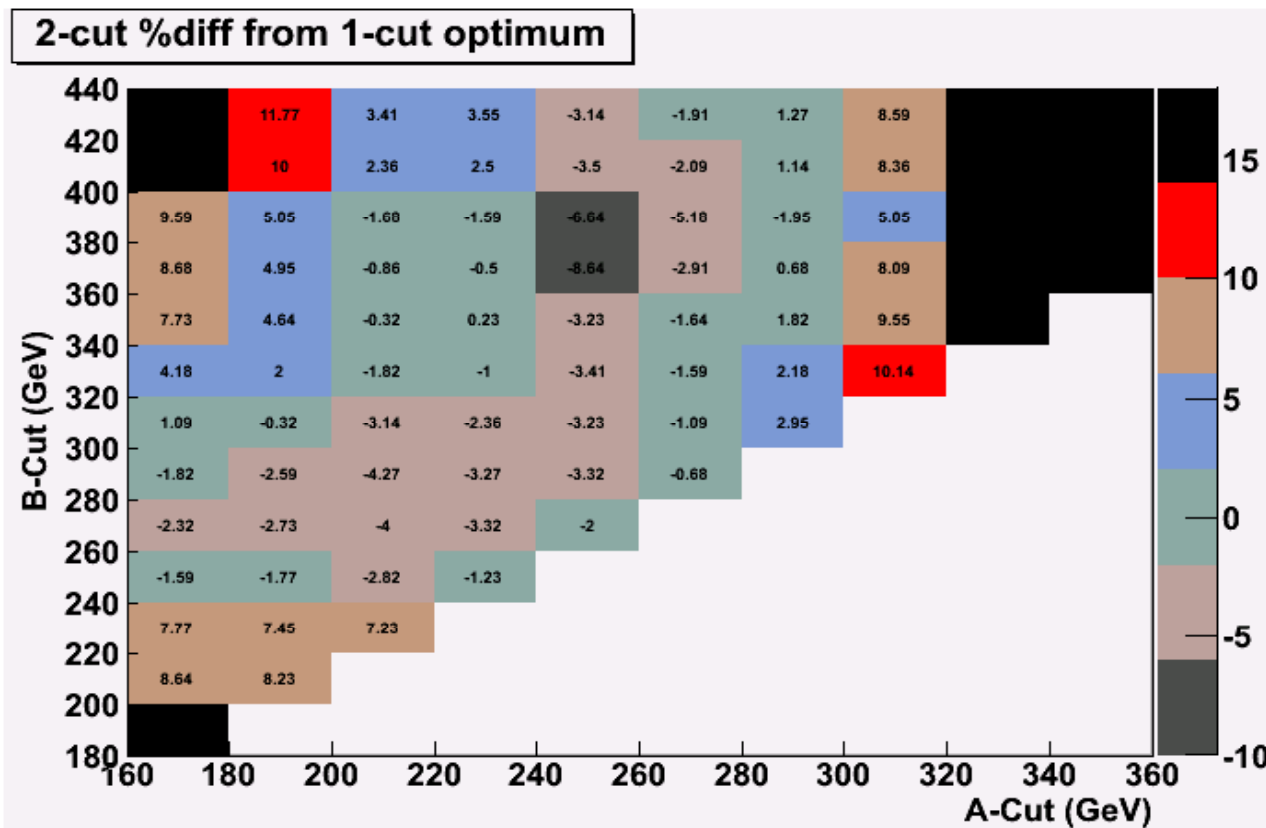
# Binned Value Two-Cut

- ▶ The optimal cuts give 20.1 fb at A:240 B:360



# Percentage Decrease

- ▶ Two cuts can be used to improve the optimal expected limit
  - Able to achieve slightly under 10% decrease (8.64%)



# Acceptance

- ▶  $A_{signal} = \frac{N_{events}^{passing\ requirements}}{N_{events}}$
- ▶ Related to signal by a scaling factor