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ATLAS

# Digestible Dark Dijets and other Delicacies

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# Project Overview

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- **Performance work: In-situ jet calibration**
  - In particular combination of in-situ methods for large-R jets
  - Have done this since QT and been trying to get out of it for a year now
- **Upgrade work: gFEX**
  - An electronics board that will be part of the current upgrade of the ATLAS Level-1 Calorimeter Trigger
  - Lower granularity, but info from full calorimeter available
  - Will improve triggering on
    - heavy, boosted objects
    - Missing transverse energy
    - Interesting heavy ion collisions
  - I work on developing online control software for the board
- **Analysis: Dark dijet resonance search**
  - This is what I will talk about today!

# School Recommendations



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- **CERN School of Computing**
  - Two weeks in Madrid
  - Thorough introduction to many important concepts in (and outside of) HEP
- **International School of Triggering and Data Acquisition (ISOTDAQ)**
  - Ten day school in Vienna
  - Half lectures / half lab-work (e.g. FPGA-programming 😊 )
- **GEANT 4 course**
  - One week here at Lund + project
  - Learn to make a simple mock-up simulation of your favorite detector
- **Nordic Detector School**
  - One week in Copenhagen, one week in Helsinki, project work after
  - Fun, fun, fun
- **Distributed Computing (COMPUTE)**
  - 16 hours lectures/exercises over four weeks + cool project at the end
  - Good ECTS value for work hours



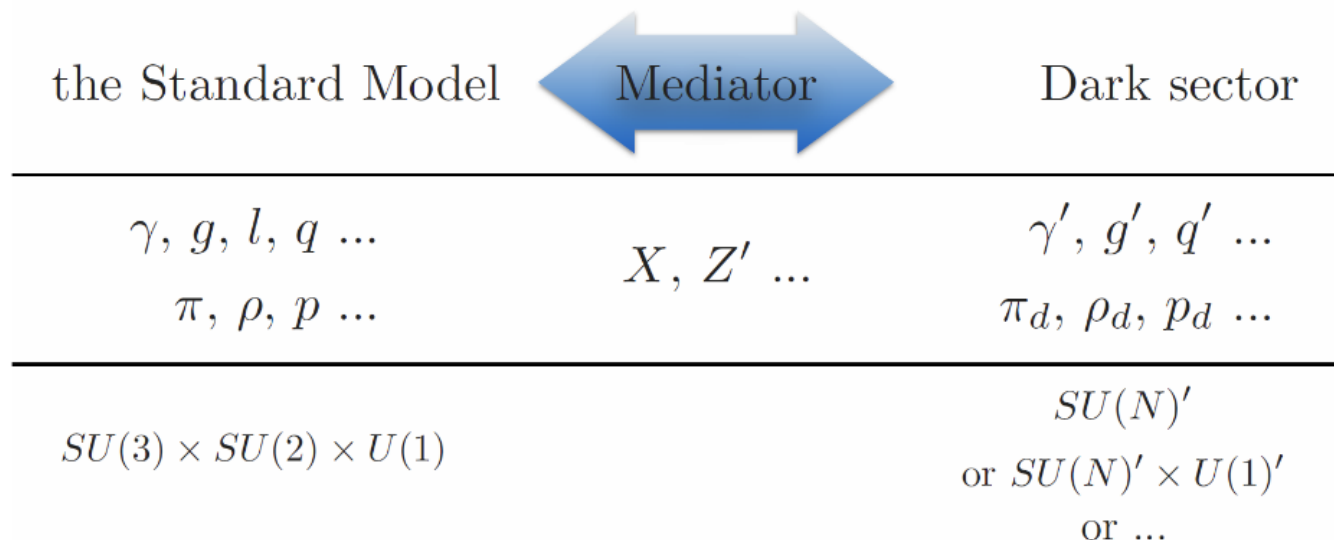
# Analysis: Dark Jet Resonance Search

# Motivation: Complex Dark Matter



- Direct and indirect **WIMP searches have come up empty** so far
- Some cosmological simulations point towards self-interacting dark matter (DM)
  - E.g. composite DM arising from hidden sector with confinement
- Combined with asymmetric DM production, could explain mass and number density of DM and Baryons
- Requires heavy mediator linking hidden sector to visible sector
  - We might be able to **produce and detect dark jets!**

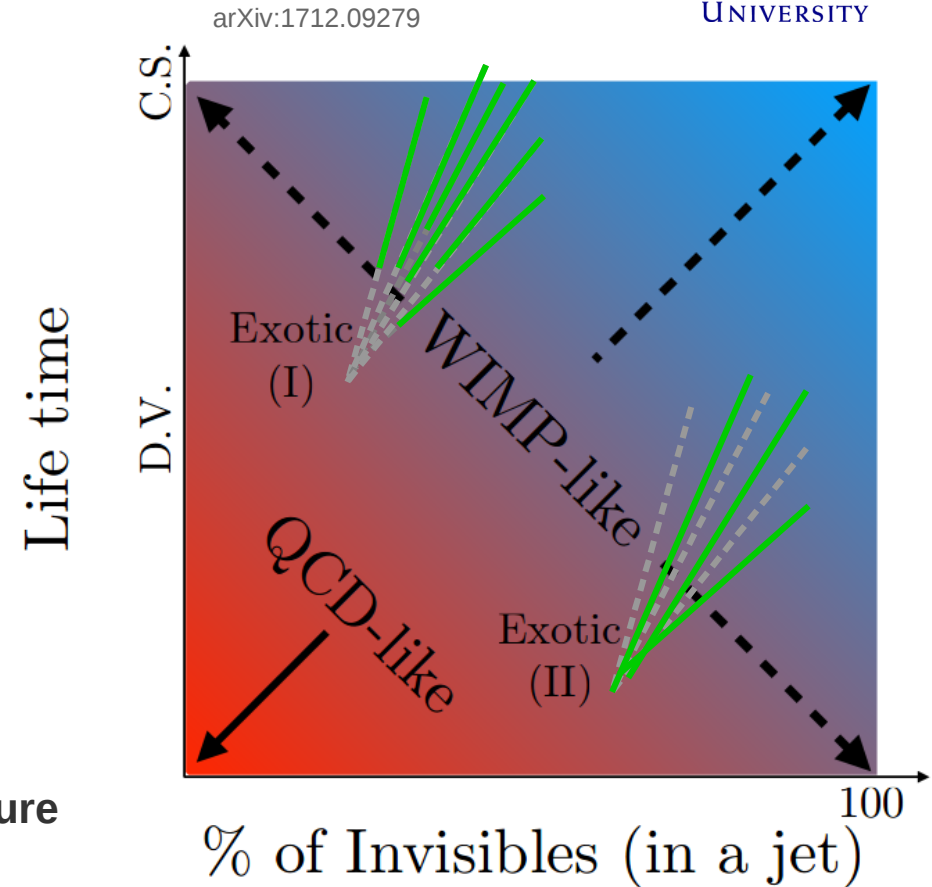
} References in  
Backup



# Motivation: Bench Mark Scenarios



- Different hidden sector models can lead to very different detector signatures
- Composition of **visible** and **invisible** partons in the jet dependent on parameter choice:
  - Exotic I: Displaced vertices, emerging jets
  - Exotic II: A fraction of dark mesons in a jet are stable and fraction decays promptly to Standard model particles
- **We target Exotic II like models:**
  - Idea: Look for dijet resonance using **substructure**
  - Based on [arXiv:1712.09279](https://arxiv.org/abs/1712.09279)
- Four models implemented in Pythia Hidden Valley process
  - Parameters chosen to be representative of different types of DM
  - Details in backup

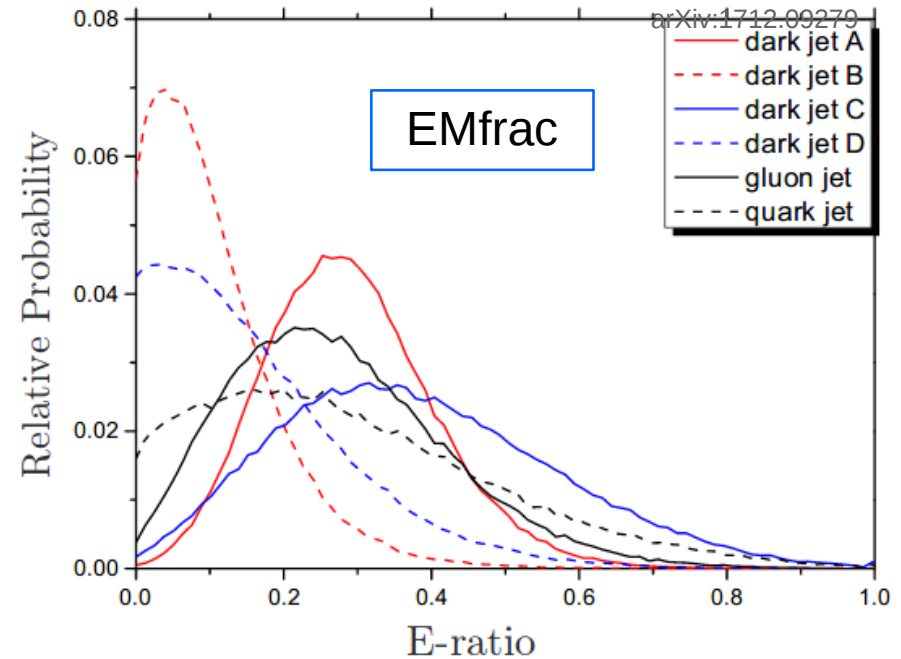
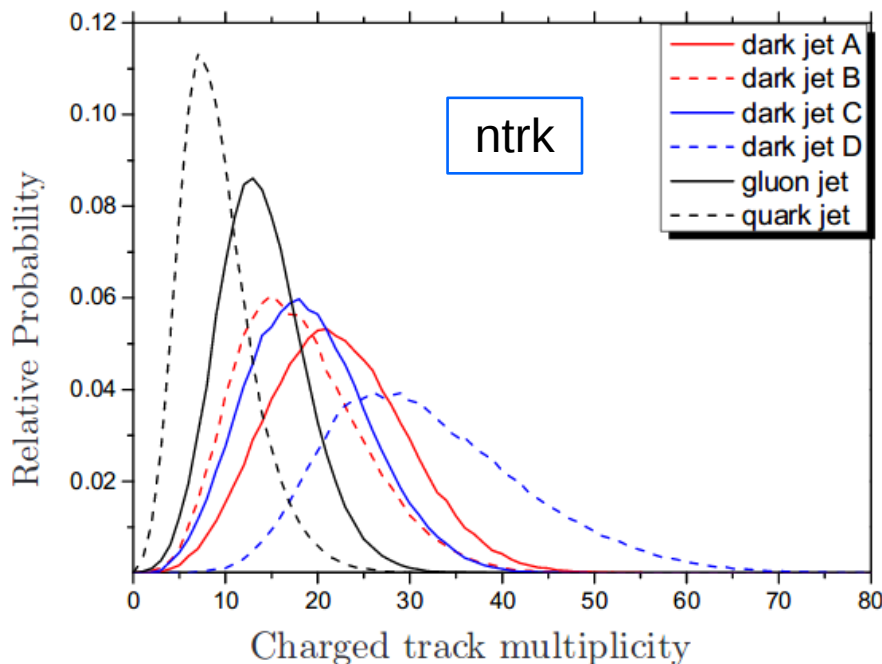


# Signal Selection Strategy



- Charged track multiplicity  $n_{trk}$  is strongest discriminating variable for a most models
  - Not true for all though – depends on coupling constant and decay time
- No upper limits exist on cross section, coupling, sensitivity
- Two signal regions:
  - **SR1:** High sensitivity, very pure, based on strict  $n_{trk}$  cut
  - **SR2:** More inclusive, less sensitive based on softer cuts on  $n_{trk}$  and  $Emfrac$

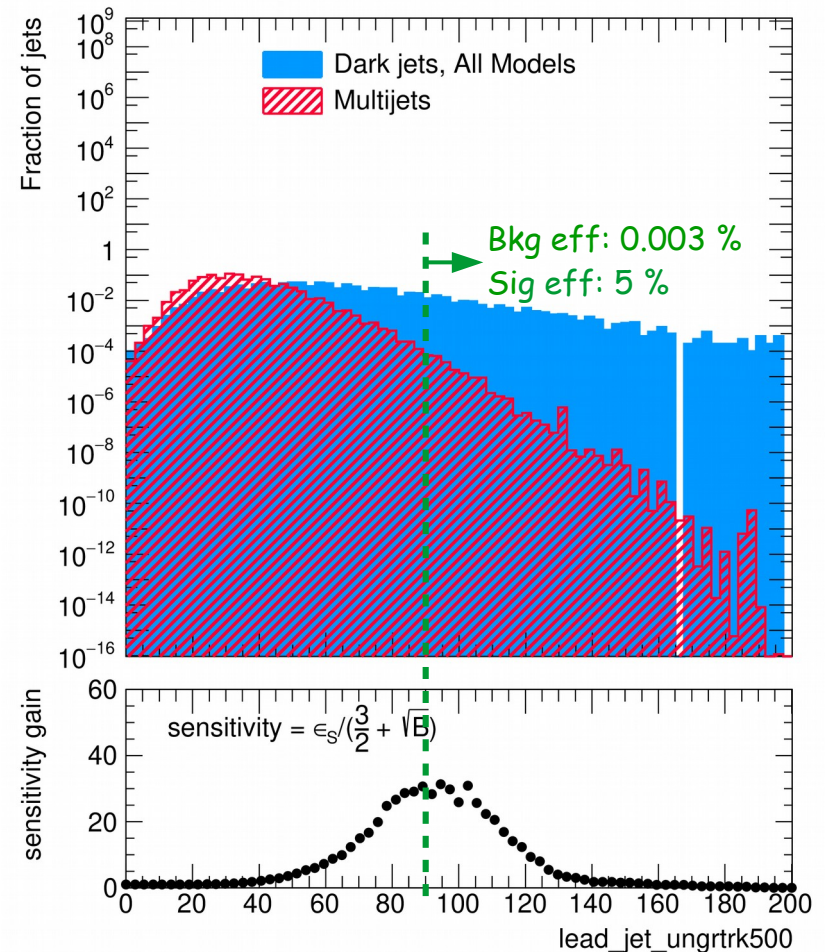
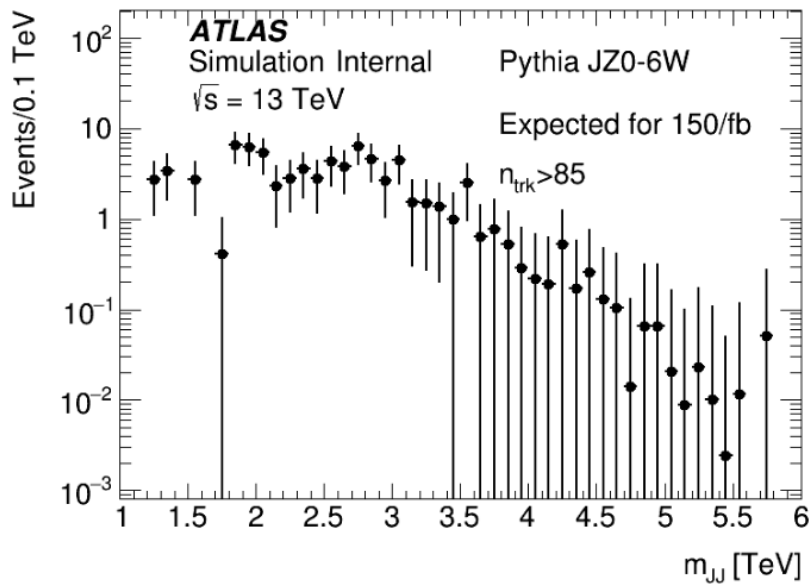
Generator-level results, no detector simulation:



# Signal Region 1: Challenge



- Best sensitivity for a range of models obtained with very strict cut
- Strict *n<sub>trk</sub>* cut significantly sculpts the background dijet mass spectrum
- Solution: Decorrelate *n<sub>trk</sub>* from dijet mass!

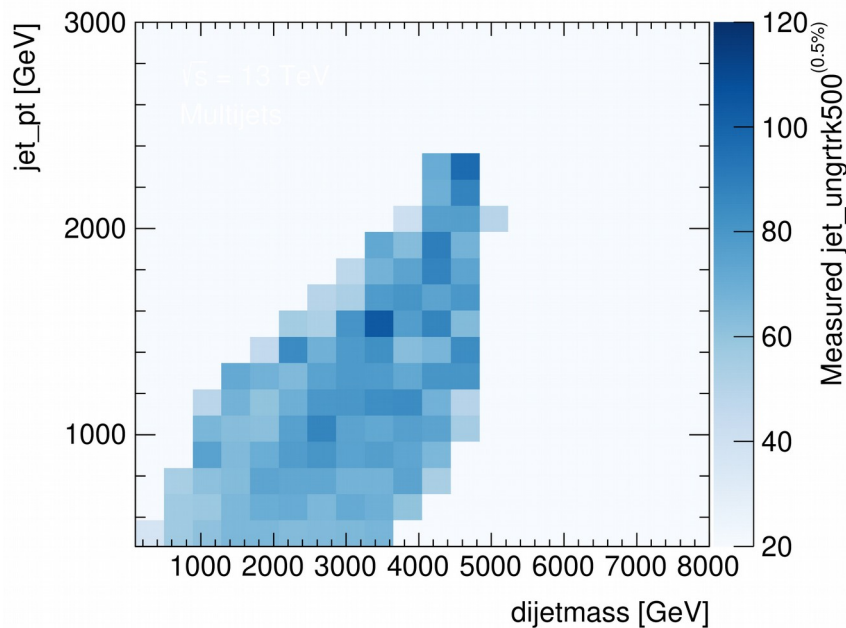




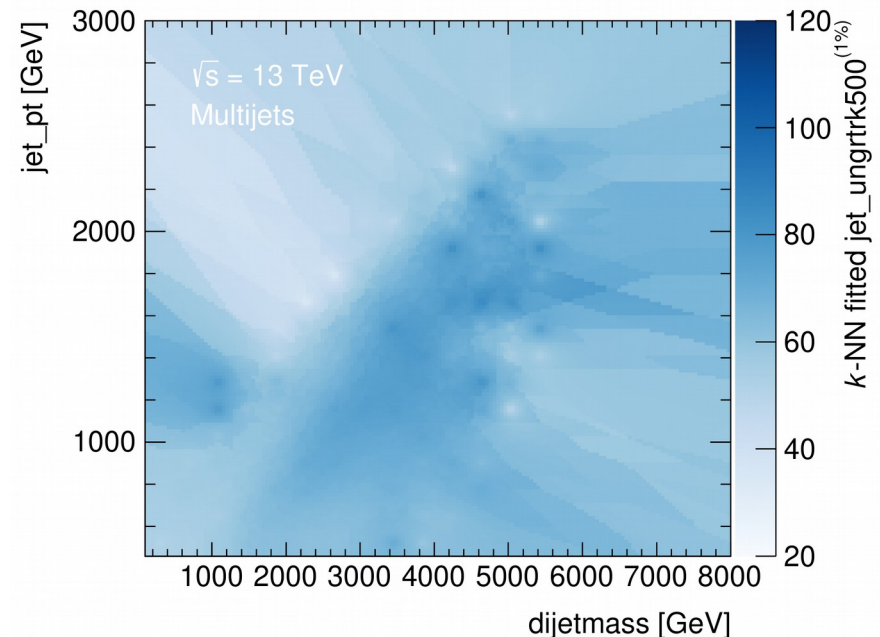
# Signal Region 1: Decorrelation method



- "kNN method" decorrelates a discriminating variable from other variables
  - Described and tested in for  $W$ -tagging in [ATL-PHYS-PUB-2017-004](#)
- Principle:
  - Choose a substructure variable with good classification power (here  $ntrk$ )
  - Decide on a fixed background single jet efficiency (0.5%, giving a signal eff. of 30%)
  - Evaluate the cut value on  $ntrk$  that gives desired bkg. eff. in bins of  $p_T$  and  $m_{jj}$
  - Fit the distribution using the  $k$ -nearest neighbours (kNN) algorithm
  - For each jet, the new decorrelated observable is computed as  $ntrk^{kNN} = ntrk - ntrk^{(0.5\%)}$
- Here  $ntrk^{kNN}$  is defined from data in a control region with signal efficiency  $\sim 0.4\%$



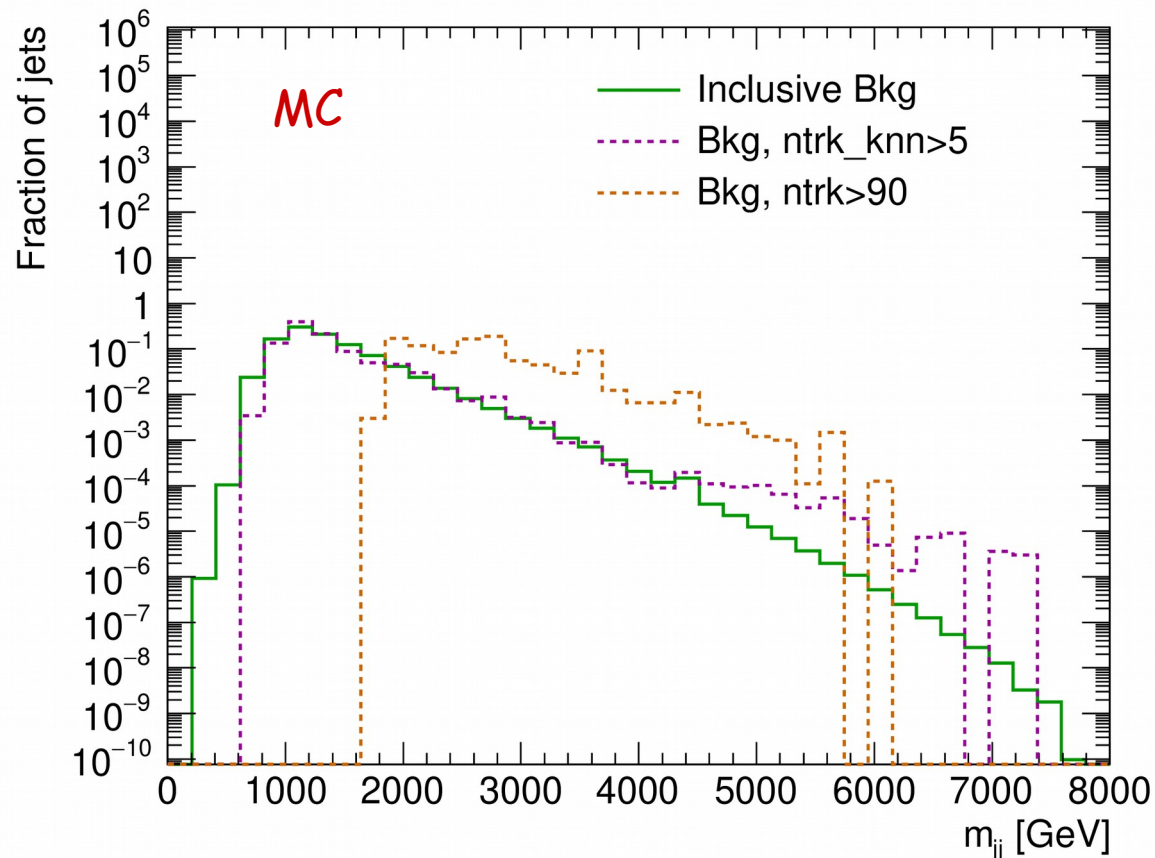
kNN fit  
→



# Signal Region 1: Results (preliminary)



- Gives a smoothly falling spectrum up to  $\sim 4$  TeV in MC
- Breaks down at higher  $m_{jj}$  because of too low statistics in data sample
- Cut: Leading and sub-leading jet  $ntrk^{kNN} > 5$ 
  - Background efficiency: 0.008 %
  - Signal efficiency (all tested models): 25 %



# Conclusion and Plans

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- **Loads of uncovered phase space in hidden sector models**
  - We can probe part of it by searching for dark jet
  - Many fun challenges for jet substructure fans
- **Analysis to-do list:**
  - Evaluate systematics
  - Optimise background fit
  - Validate with signal injection
- **Other plans:**
  - Write gFEX software
  - Write thesis

# Backup



- Table of benchmark models

arXiv:1712.09279

	$N_d$	$n_f$	$\Lambda_d$ (GeV)	$\tilde{m}_{q'}$ (GeV)	$m_{\pi_d}$ (GeV)	$m_{\rho_d}$ (GeV)	$\pi_d$ Decay Mode	$\rho_d$ Decay Mode
<i>A</i>	3	2	15	20	10	50	$\pi_d \rightarrow c\bar{c}$	$\rho_d \rightarrow \pi_d\pi_d$
<i>B</i>	3	6	2	2	2	4.67	$\pi_d \rightarrow s\bar{s}$	$\rho_d \rightarrow \pi_d\pi_d$
<i>C</i>	3	2	15	20	10	50	$\pi_d \rightarrow \gamma'\gamma'$ with $m_{\gamma'} = 4.0$ GeV	$\rho_d \rightarrow \pi_d\pi_d$
<i>D</i>	3	6	2	2	2	4.67	$\pi_d \rightarrow \gamma'\gamma'$ with $m_{\gamma'} = 0.7$ GeV	$\rho_d \rightarrow \pi_d\pi_d$

# Backup



- Sensitivity distribution of  $ntrk^{kNN}$
- Cut at peak sensitivity does not leave enough background events to fit

