

# Casting a ParticleNet to catch dark showers

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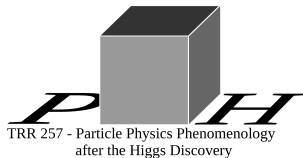
based on **arXiv:2005.xxxxx**

with Thorben Finke, Felix Kahlhoefer, Michael Krämer and Alexander Mück

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Physics of the Heaviest Particles at the LHC



- **What if the dark sector resembles QCD?**

⇒ DM could be meson/baryon in confining dark sector

- **Cosmology:** relic density from interactions within dark sector

Hochberg et al., 1411.3727

- **Astrophysics:** possible resolution of DM small-scale problems (SIDM)

Hochberg et al., 1402.5143

- **Novel LHC phenomenology:**

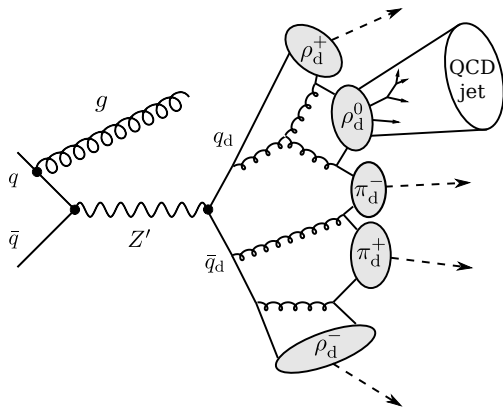
- dark showers
- semi-visible jets
- emerging jets

Cohen et al., 1707.05326

Schwaller et al., 1502.05409

# Dark showers at the LHC

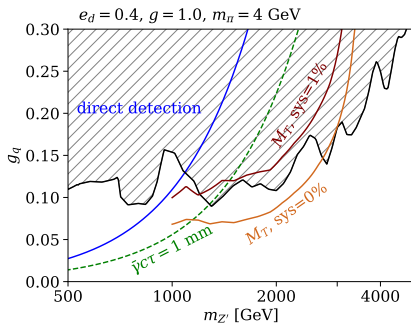
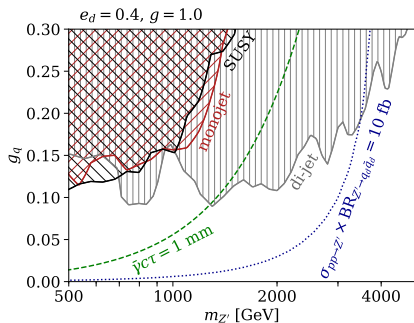
- Dark  $SU(3)$ , dark pion DM, consistent cosmology EB et al., 1907.04346
- Production of dark quarks at the LHC via heavy vector mediator
- Shower and hadronisation in dark sector (PYTHIA HIDDEN VALLEY)



$$m_{Z'} \sim \mathcal{O}(\text{TeV}), m_{\pi} \sim m_{\rho} \sim \mathcal{O}(\text{GeV})$$

- 10 - 20 dark mesons in an event
  - Most escape the detector as  $\cancel{E}_T$
  - $\rho_d^0$  decay to visible jets
- ⇒ **Semi-visible jets**
- ⇒ Jets +  $\cancel{E}_T$ , often aligned

# Existing and prospective LHC searches



## Two classes of events:

- If one dark shower stays invisible:  
⇒ Limits from existing monojet and SUSY searches
- If both dark showers become partly visible:  
⇒ Prospective semi-visible jet search: bump hunt in  $M_T$  for small  $\Delta\phi$

Cohen et al., 1707.05326

improves existing limits only under optimistic assumptions

# Can we do better with machine learning?

- Proposed semi-visible jet search does not use jet substructure
  - Semi-visible jets differ substantially from QCD
- ⇒ **Train a neural network classifier to distinguish dark showers from QCD**
- Wide range of supervised and unsupervised ML approaches for jet classification, most commonly benchmarked on top tagging  
Kasieczka et al., 1902.09914  
For example Convolutional Neural Networks on jet images
  - Dark showers more similar to QCD than tops: varying number of light dark mesons with varying missing energy between

- Originally from computer vision
- Recently used as jet tagger: ParticleNet

Wang et al., 1801.07829  
Qu, Gouskos, 1902.08570

## Jets as point clouds

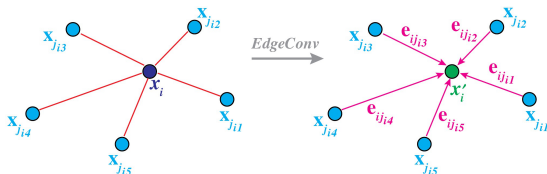
- Every constituent is a point in a high-dimensional feature space
- No particular ordering

## Edge convolution

- For each point construct graph of  $k$  nearest neighbours
- Carry out convolution over edges (features of pairs of neighbours)

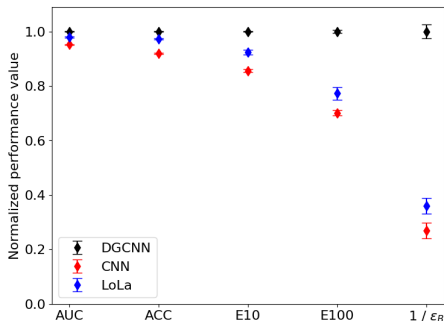
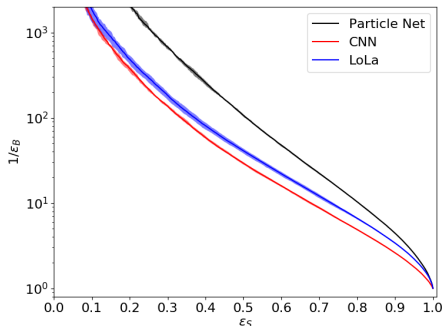
$$x'_i = \frac{1}{k} \sum_{j=1}^k h_{\Theta}(x_i, x_{j_i})$$

with points  $x_i \in \mathbb{R}^F$  and edge function  $h_{\Theta} : \mathbb{R}^F \times \mathbb{R}^F \rightarrow \mathbb{R}^{F'}$



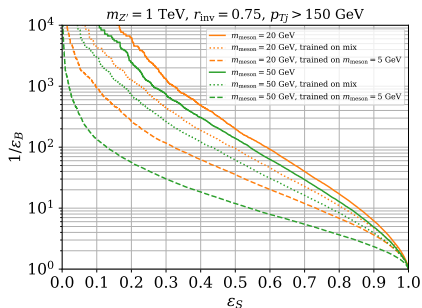
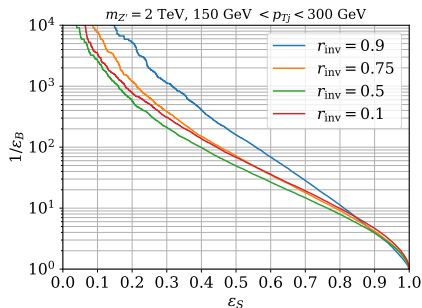
Wang et al., 1801.07829

# DGCNN performance in comparison to other networks



- Signal: semi-visible jets from dark showers, background: QCD jets
- DGCNN outperforms CNN operating on jet images as well as LoLa on 4-vectors
- Advantage of DGCNN is larger than for tops: dark shower identification the more difficult problem

# Varying dark sector parameters: $r_{\text{inv}}$ and $m_\rho$



- Invisible fraction  $r_{\text{inv}}$  of shower can differ from 0.75
- Dark showers with larger  $r_{\text{inv}}$  easier to identify (more unlike QCD), once  $p_T$  is factored out
- Performance depends sensitively on dark meson mass  
⇒ mixed training or parametrised network

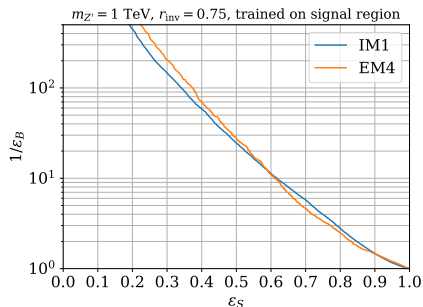


By how much can we improve an analysis with our dark shower tagger?

⇒ **Monojet search as example**

ATLAS-CONF-2017-060

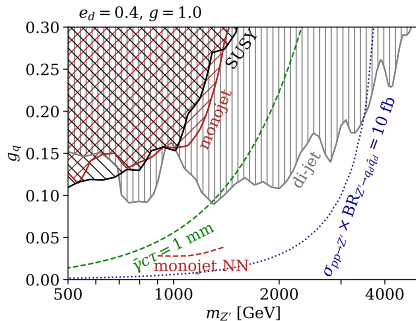
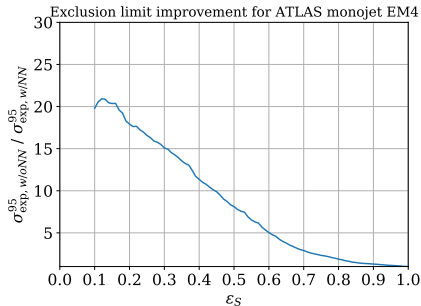
- Train on dark showers and dominant background ( $Z$ +jets), separately for each signal region



- Apply tagger after ordinary monojet cuts
- Require at least one jet tagged as dark shower

# Sensitivity improvement and expected limit

Preliminary:



- Highly suppressed background, uncertainty becomes statistics-dominated
- Substantial increase in sensitivity
- Can bridge gap to displaced regime

- Strongly interacting dark matter is a well motivated scenario predicting exciting new LHC signatures
- Difficult to identify with conventional methods
- Great opportunity for machine learning
- Graph nets seem particularly suited to this task
- Can increase the sensitivity of searches by a lot
- Still on the lookout for an unsupervised technique that works here . . .

# Backup

$$SU(3)_{\text{dark}} \times U(1)'_{\text{mediator}}$$

- 2 flavours of dark quarks  $q_d$
- $Z'$  mediator  $\sim \mathcal{O}(\text{TeV})$  coupling to  $q_{\text{SM}}$  and  $q_d$



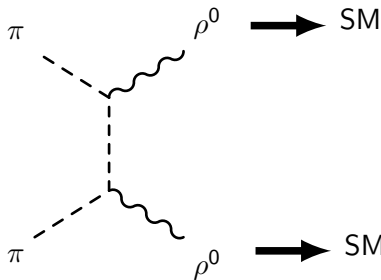
confinement at  $\Lambda_d$



- $\pi_d^0, \pi_d^\pm, \rho_d^0, \rho_d^\pm \sim \mathcal{O}(\text{GeV})$
- **Dark pions are DM (stable)**
- $Z' \pi_d^+ \pi_d^-, Z' \rho_d^+ \rho_d^-$  coupling
- $Z' - \rho_d^0$  mixing  $\Rightarrow \rho_d^0$  **unstable**

# Freeze-out

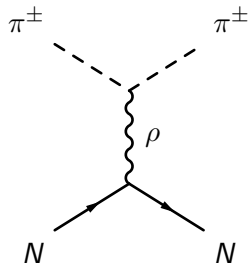
- $\rho_d$  in equilibrium in early Universe if  $\Gamma_{\rho^0} > H$
- $\pi_d$ - $\rho_d$  decoupling sets DM relic density
- Dominant freeze-out process:  $\pi_d \pi_d \rightarrow \rho_d \rho_d$  (forbidden DM, D'Agnolo et al., 1505.07107)



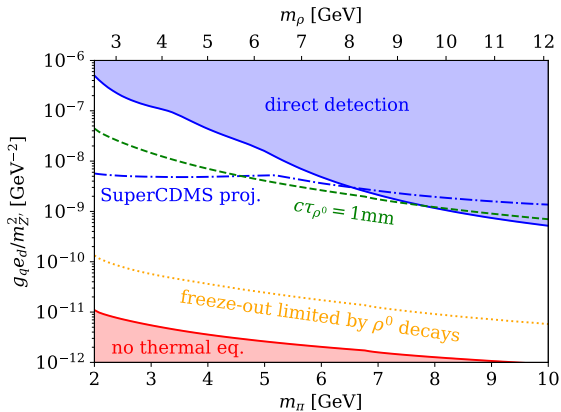
$$\sigma_{\pi\pi \rightarrow \rho\rho} \propto \frac{g^2}{m_\pi^2} e^{-2\Delta x_f}$$
$$\Delta = \frac{m_\rho - m_\pi}{m_\pi} \sim 0.2 - 0.5$$

- Relic density can be easily produced by adjusting  $m_\rho/m_\pi$ .

# Direct detection



$$\sigma^{\text{SI}} \propto \frac{e_d^2 g_q^2}{m_{Z'}^4}$$



Combination of CRESST-III, CDMSLite, PICO-60, PandaX and XENON1T

The Lagrangian of the underlying dark  $SU(3)$  gauge theory coupled to  $Z'$  reads

$$\begin{aligned}\mathcal{L} = & -\frac{1}{4}F_{\mu\nu}^a F^{\mu\nu,a} + \bar{q}_d i \not{D} q_d - \bar{q}_d M_q q_d \\ & -\frac{1}{4}Z'_{\mu\nu} Z'^{\mu\nu} + \frac{1}{2}m_{Z'}^2 Z'_\mu Z'^\mu - g_q Z'_\mu \sum_{q_{SM}} \bar{q}_{SM} \gamma^\mu q_{SM} ,\end{aligned}$$

where  $q_d = (q_{d,1}, q_{d,2})$  and  $M_q = \text{diag}(m_q, m_q)$ . The dark quark covariant derivative has the form

$$D_\mu q_d = \left( \partial_\mu + ig_D A_\mu + ie_D Z'_\mu Q \right) q_d ,$$



The chiral EFT Lagrangian (up to fourth order in the pion fields) is given by

$$\begin{aligned}\mathcal{L}_{\text{EFT}} = & \text{Tr} (D_\mu \pi D^\mu \pi) - \frac{2}{3f_\pi^2} \text{Tr} (\pi^2 D_\mu \pi D^\mu \pi - \pi D_\mu \pi \pi D^\mu \pi) \\ & + m_\pi^2 \text{Tr} (\pi^2) + \frac{m_\pi^2}{3f_\pi^2} \text{Tr} (\pi^4) + \mathcal{O} \left( \frac{\pi^6}{f_\pi^4} \right) \\ & - \frac{1}{4} \text{Tr} (V_{\mu\nu} V^{\mu\nu}) + m_V^2 \text{Tr} (V^2) - \frac{e_d}{g} Z'_{\mu\nu} \text{Tr} (QV^{\mu\nu}) . \quad (1)\end{aligned}$$

The pion covariant derivative is given by

$$D_\mu \pi = \partial_\mu \pi + ig [\pi, V_\mu] + ie_d [\pi, Q] Z'_\mu .$$

$$\begin{pmatrix} \tilde{Z}' \\ \tilde{\rho}_0 \end{pmatrix} = \begin{pmatrix} \sec \epsilon & \sin \epsilon \frac{m_\rho^2}{m_{Z'}} \\ -\tan \epsilon + \frac{1}{2} \sin 2\epsilon \frac{m_\rho^2}{m_{Z'}^2} & 1 - \sin^2 \epsilon \frac{m_\rho^2}{m_{Z'}^2} \end{pmatrix} \begin{pmatrix} Z' \\ \rho^0 \end{pmatrix}, \quad (2)$$

where  $\epsilon = \arcsin(2 e_d/g)$ .

$$\mathcal{L}_{\text{EFT}} \supset \frac{2 e_d g_q}{g} \frac{m_\rho^2}{m_{Z'}^2} \rho_\mu^0 \sum_{q_{\text{SM}}} \overline{q_{\text{SM}}} \gamma^\mu q_{\text{SM}}. \quad (3)$$

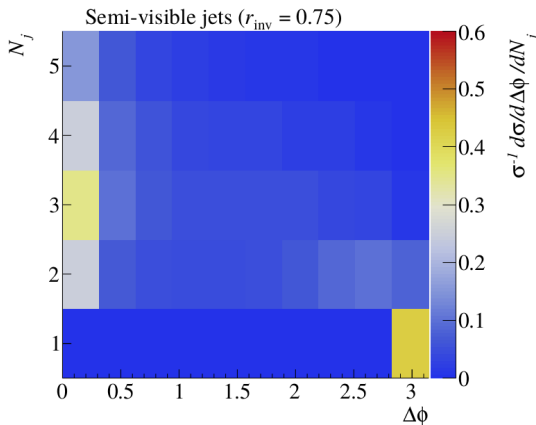
$$\mathcal{L}_{\text{EFT}} \subset \left( -2 e_d \sqrt{1 - \frac{4 e_d^2}{g^2}} \frac{m_\rho^2}{m_{Z'}^2} Z'_\mu + g \rho_\mu^0 \right) [\pi^+ (\partial^\mu \pi^-) - (\partial^\mu \pi^+) \pi^-]. \quad (4)$$

- Charged pions stable
- $\pi^0$  generically unstable
  - extremely dangerous in early universe (CMB, BBN, relic density, ...)
  - stabilised by  $G_d$ -parity if  $N_F$  even and  $Q \propto \text{diag}(1, -1)$
- $\rho^\pm$  effectively stable if  $m_\rho < 2m_\pi$  (tiny three-body decay width)
- $\rho^0$ - $Z'$  mixing induces  $\rho^0$  decays to  $q_{\text{SM}}\bar{q}_{\text{SM}}$

$$\rightarrow c\tau_{\rho^0} \approx 3.2 \text{ mm} \times \left(\frac{g_q}{0.01}\right)^{-2} \left(\frac{e_d}{0.4}\right)^{-2} \left(\frac{m_\rho}{5 \text{ GeV}}\right)^{-5} \left(\frac{m_{Z'}}{1 \text{ TeV}}\right)^4$$

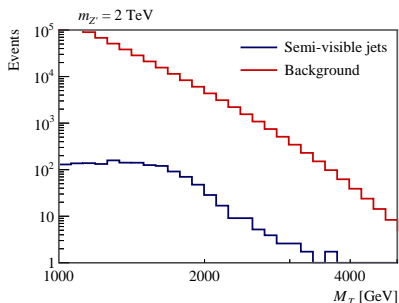
# Semi-visible jets

- Signature: jets +  $\cancel{E}_T$ , but with a special twist
- Angular separation  $\Delta\phi = \min_{i \leq 4} \{ \Delta\phi_{j_i, \cancel{E}_T} \}$ :  $\cancel{E}_T$  often aligned with jet



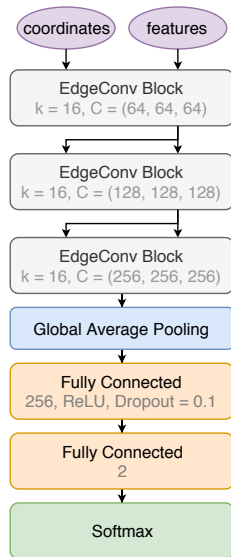
# Prospective search for semi-visible jets

- Peak in  $M_{jj}$  washed out by variation in dark meson production
- Instead **search for peak in  $M_T$**  [Cohen et al. 1707.05326]:
  - Recluster jets into two large jets ( $R = 1.1$ )
  - Inverted angular cut:  $\Delta\phi < 0.4$
  - Bump hunt in  $M_T = \left( M_{jj}^2 + 2 \left( \sqrt{M_{jj}^2 + p_{Tjj}^2} \cancel{E}_T - \vec{p}_{Tjj} \cdot \vec{\cancel{E}}_T \right) \right)^{1/2}$
- Backgrounds from Cohen et al. [1707.05326]



## Architecture (based on ParticleNet)

- Stacked EdgeConv blocks
  - Input features for each constituent:  
 $\Delta\eta$ ,  $\Delta\phi$ ,  $\log p_T$ ,  $\log E$ ,  $\log(p_T/p_T(\text{jet}))$ ,  
 $\log(E/E(\text{jet}))$ ,  $\Delta R$
  - $k$  nearest neighbours based on  $\eta$  and  $\phi$  in first block
  - in later blocks based on distance in new feature space  $\Rightarrow$  dynamic GCNN
- Finally fully-connected layers for classification



Qu, Gouskos, 1902.08570