

# The Hidden Geometry of Particle Collisions

Jesse Thaler



Heidelberg Teilchentee / MITP Virtual Workshop Seminar — June 24, 2021

# The NSF AI Institute for Artificial Intelligence and Fundamental Interactions (IAIFI)

“eye-phi”



*Advance physics knowledge — from the smallest building blocks of nature  
to the largest structures in the universe — and galvanize AI research innovation*



[<http://iaifi.org>, MIT News Announcement]

# “Collision Course”

## Robustness of Energy Flow

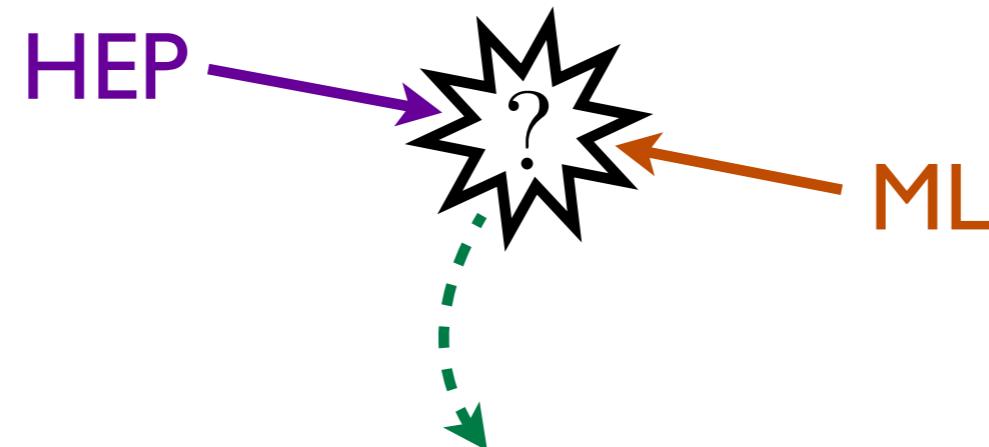
[Komiske, Metodiev, JDT, JHEP 2018]



Patrick Komiske

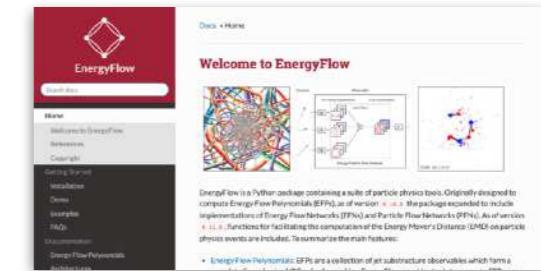


Eric Metodiev



## Power of Point Cloud Learning

[Zaheer, Kottur, Ravanbakhsh, Poczos, Salakhutdinov, Smola, NIPS 2017]



## Energy Flow Networks

<https://energyflow.network/>

[Komiske, Metodiev, JDT, JHEP 2019]

Machine Learning for  
Particle Physics  
21 June – 2 July 2021

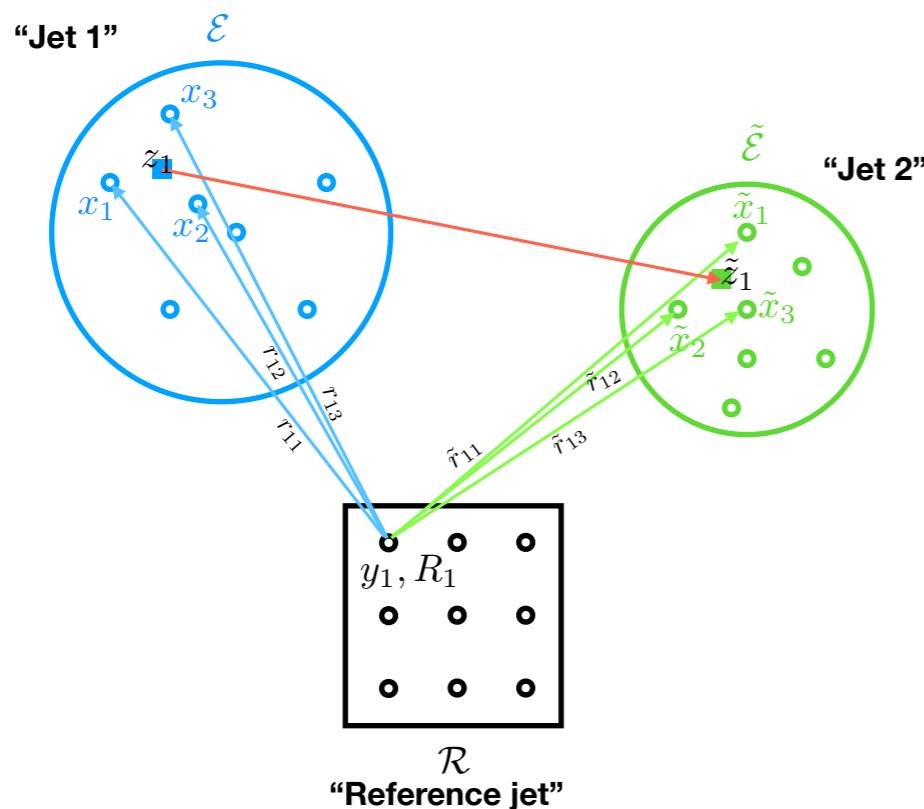
<https://indico.mitp.uni-mainz.de/event/199>



# Optimal Transport for HEP

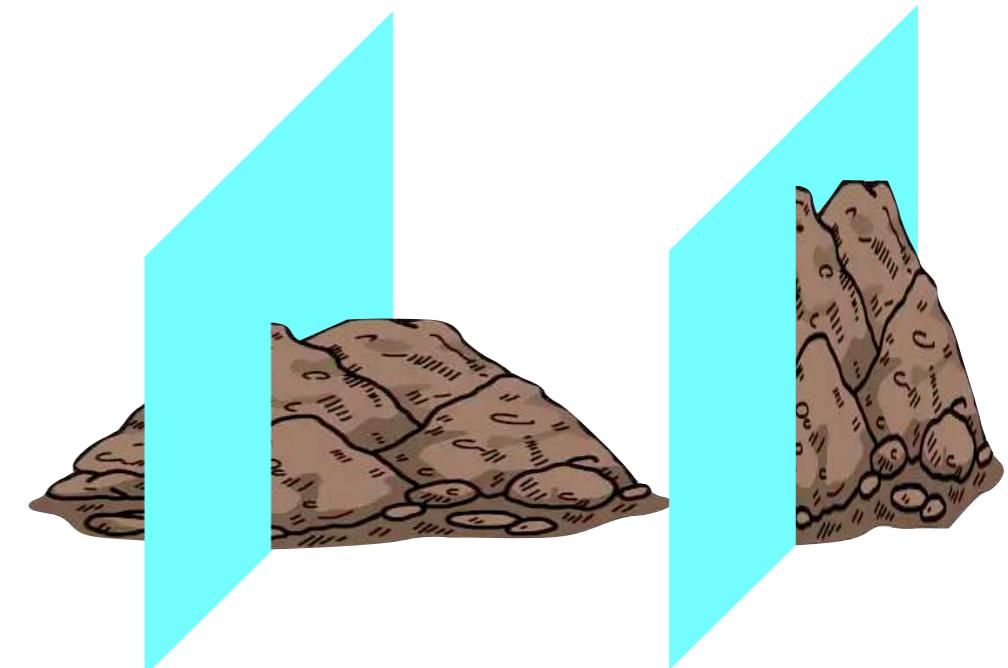
See talks by Nathaniel and Jessica on Monday!

## Linearized OT for Jet Classification



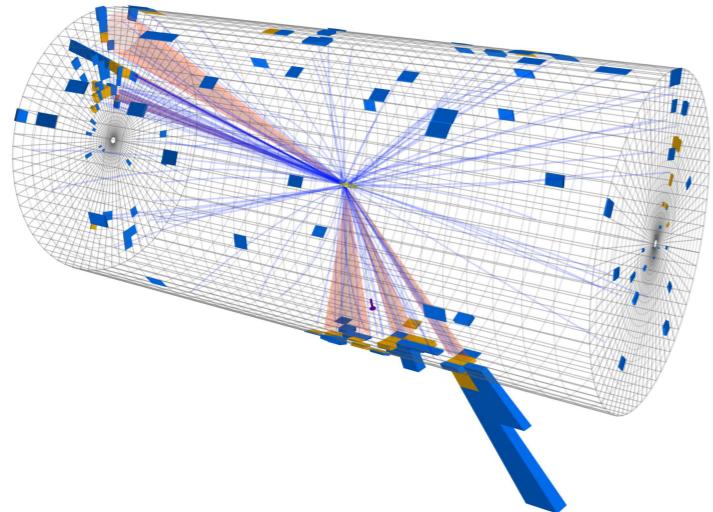
[Cai, Cheng, Craig, Craig, [PRD 2020](#)]

## Sliced Wasserstein for Simulation/Unfolding

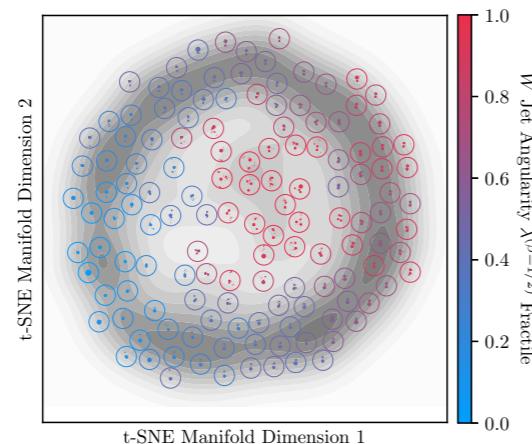


[Howard, Mandt, Whiteson, Yang, [arXiv 2021](#)]

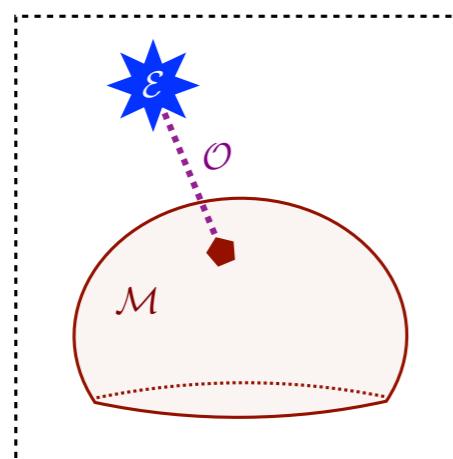
## What does moving “earth” have to do with collider physics?



From Manifest Geometry...

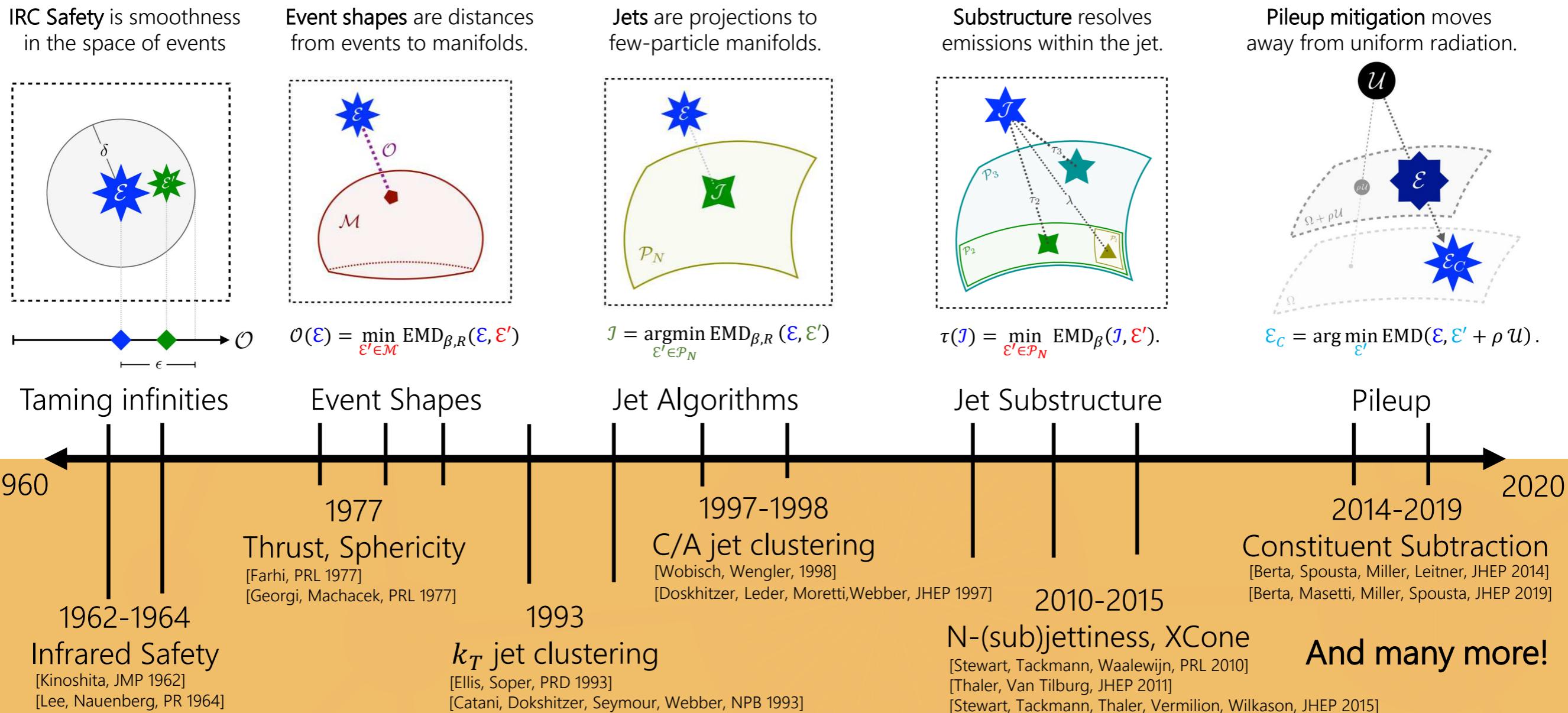


...to Emergent Geometry...



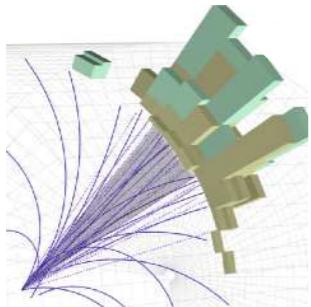
...to Hidden Geometry!

# Six Decades of Collider Physics Translated into a New Geometric Language!

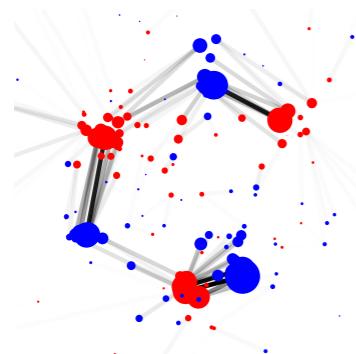


[timeline from Eric Metodiev]

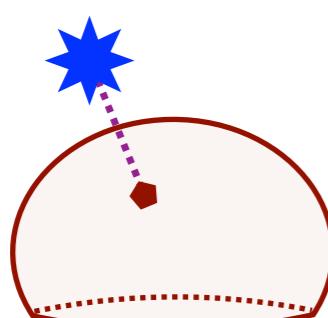
# Outline



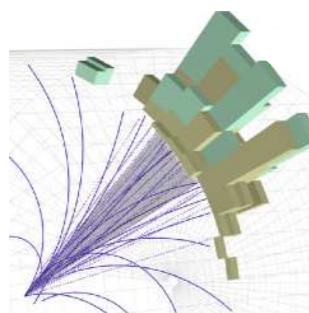
Going with the (Energy) Flow



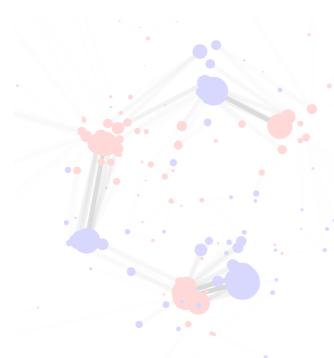
The Energy Mover's Distance



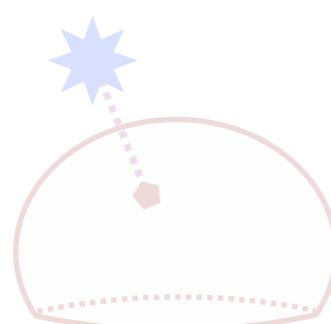
Revealing a Hidden Geometry



## Going with the (Energy) Flow



The Energy Mover's Distance



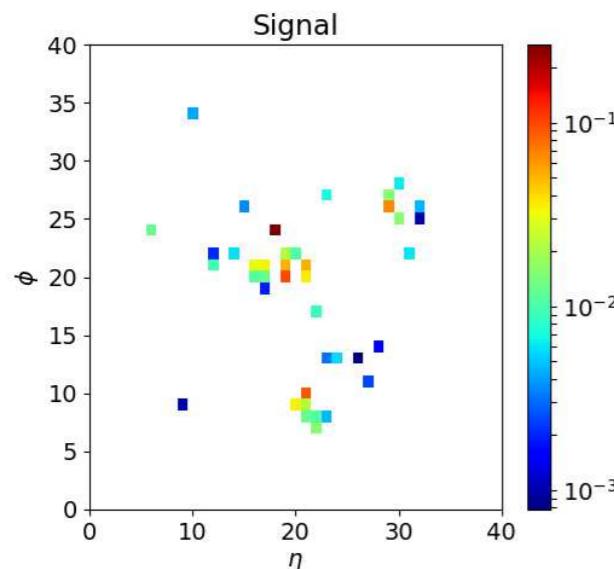
Revealing a Hidden Geometry

*Taking a step back to  
supervised machine learning...*

# Jet Representations

## Pixelized Image

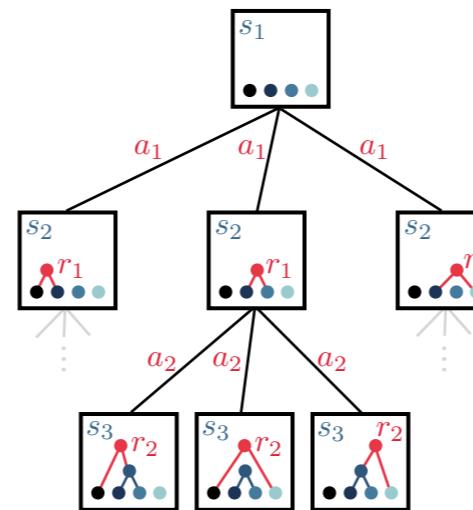
*Calorimetry*



[review in Kagan, [arXiv 2020](#)]

## Hierarchical Tree

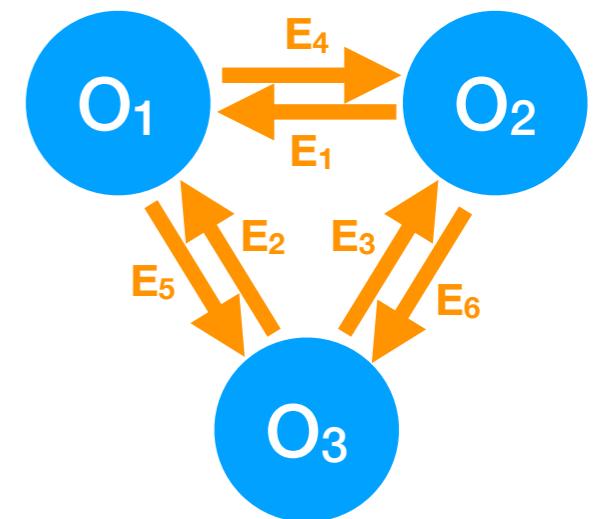
*Binary Splittings*



[e.g. Brehmer, Macaluso, Pappadopulo, Cranmer, [NeurIPS 2020](#)]

## Graphs

*Pairwise Interactions*

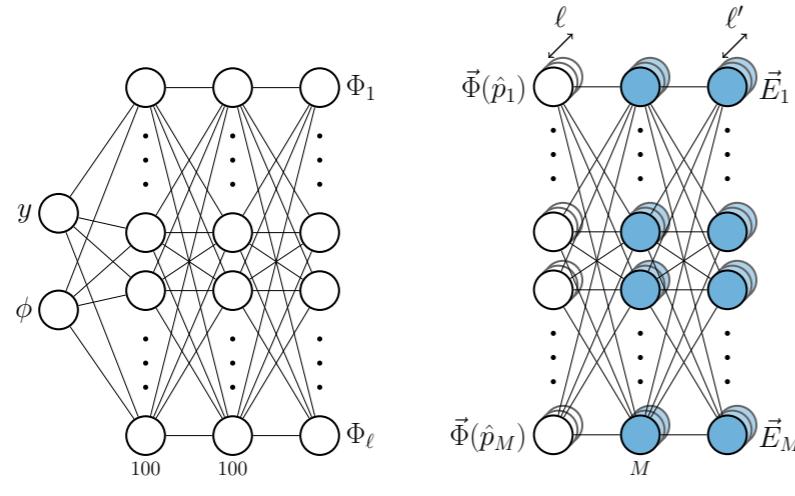


[e.g. Moreno, Cerri, Duarte, Newman, Nguyen, Periwal, Pierini, Serikova, Spiropulu, Vlimant, [EPJC 2020](#)]

*Imposes implicit **theoretical prior** (typically a good thing!)*  
*Influences choice of **network architecture***

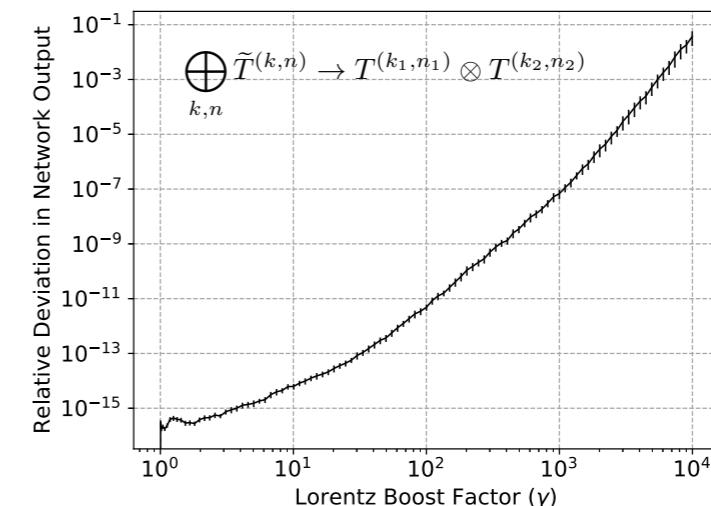
# From Principles to Network Architectures

## Permutation Equivariance



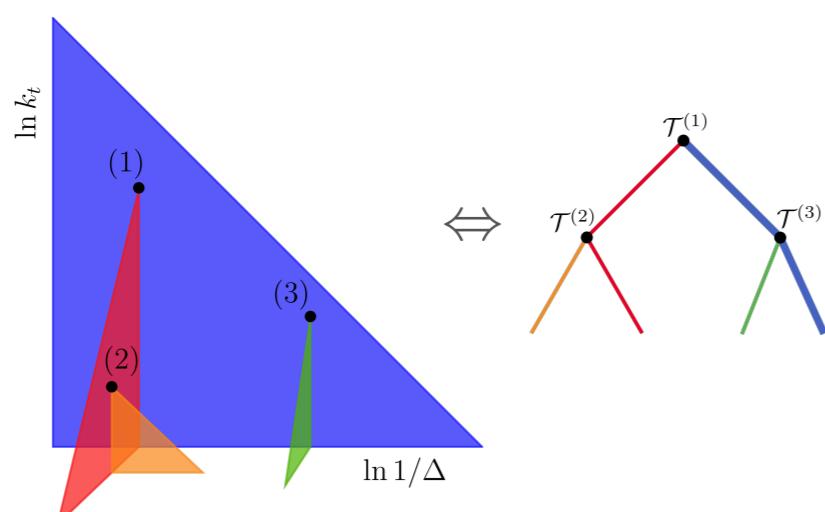
[Dolan, Ore, [PRD 2021](#)]

## Lorentz Equivariance



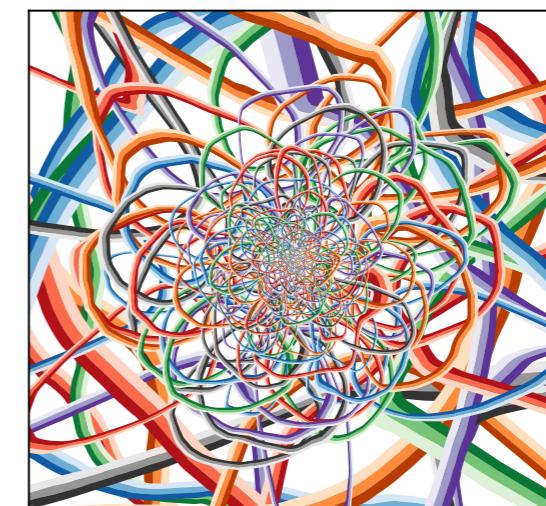
[Bogatskiy, Anderson, Oeffermann, Roussi, Miller, Kondor, [arXiv 2020](#)]

## Lund Plane Emissions



[Dreyer, Qu, [JHEP 2021](#)]

## Infrared and Collinear Safety

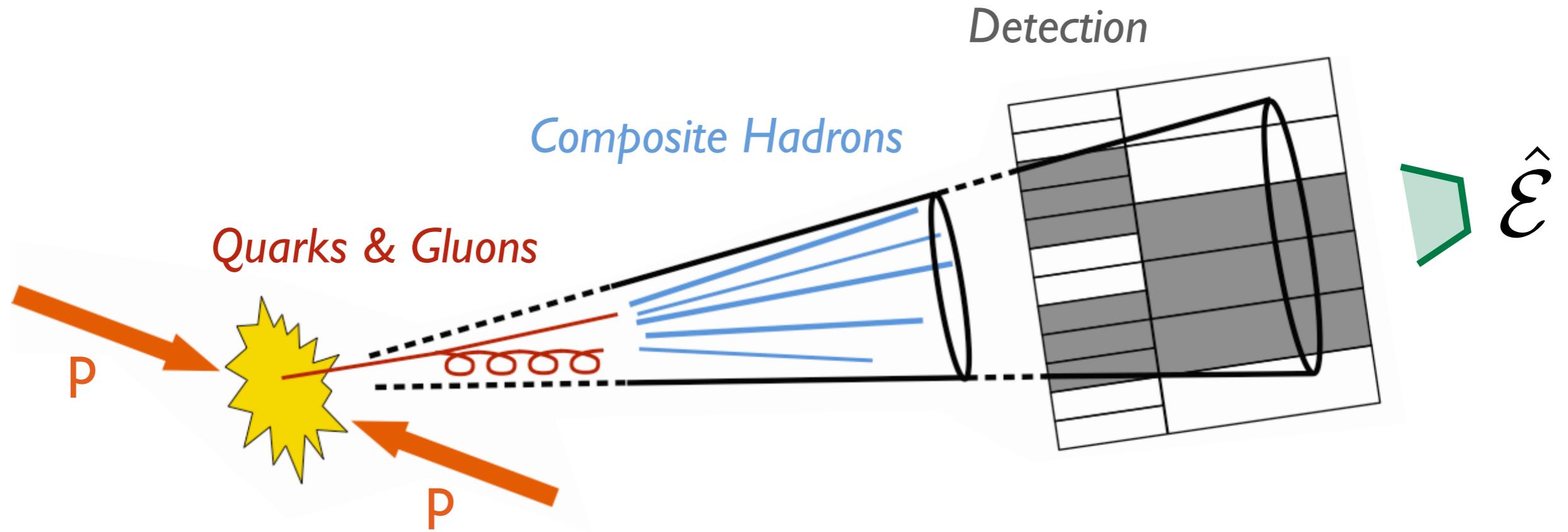


[Komiske, Metodiev, [JDT, JHEP 2019](#)]

# Energy Flow Representation

Emphasizes *infrared and collinear safety*

Theory



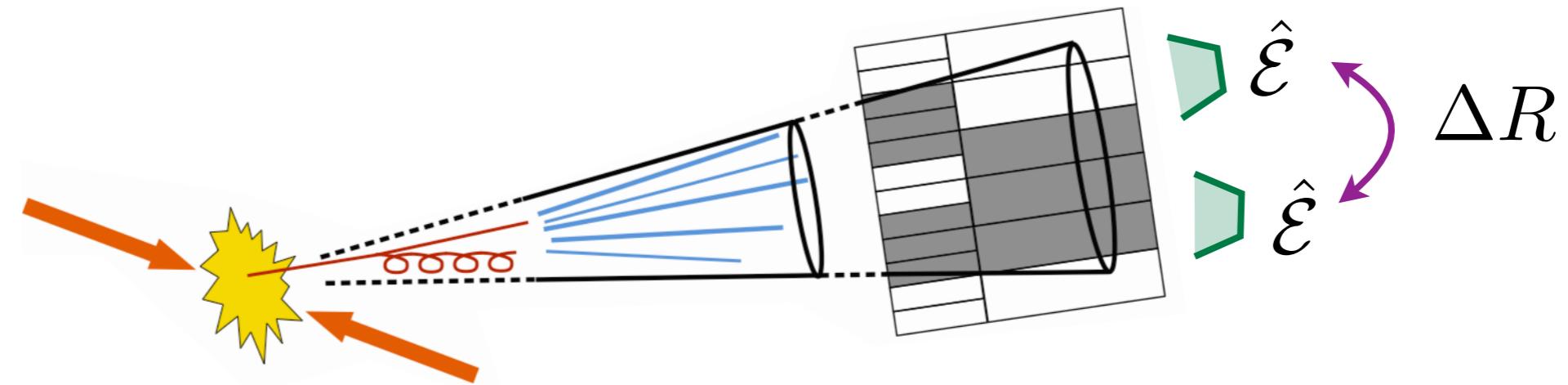
## Energy Flow:

Robust to hadronization and detector effects  
Well-defined for massless gauge theories

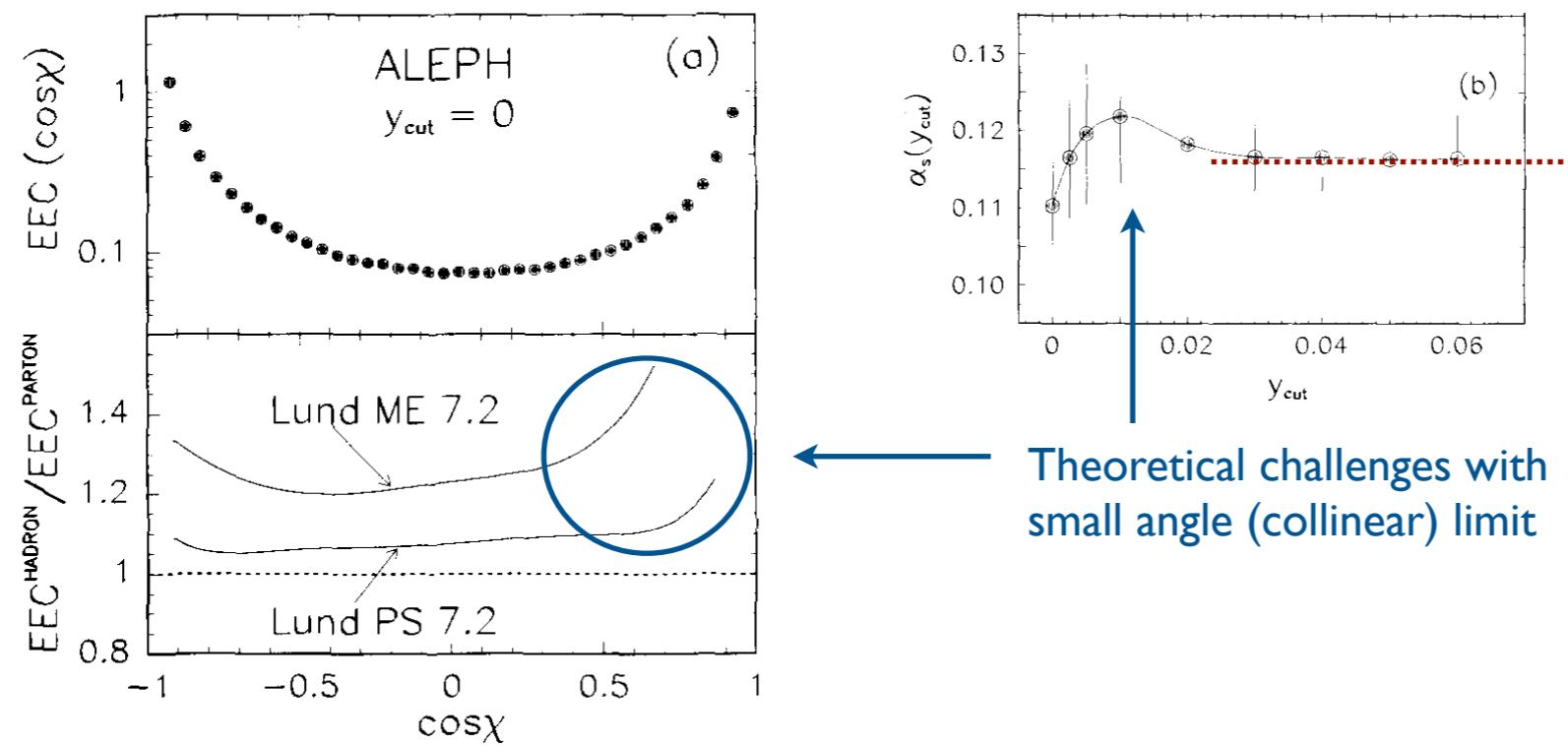
$$\hat{\mathcal{E}} \simeq \lim_{t \rightarrow \infty} \hat{n}_i T^{0i}(t, vt\hat{n})$$

[see e.g. Sveshnikov, Tkachov, [PLB 1996](#); Hofman, Maldacena, [JHEP 2008](#); Mateu, Stewart, [JDT, PRD 2013](#); Belitsky, Hohenegger, Korchemsky, Sokatchev, Zhiboedov, [PRL 2014](#); Chen, Moult, Zhang, Zhu, [PRD 2020](#)]

# Energy-Energy Correlators

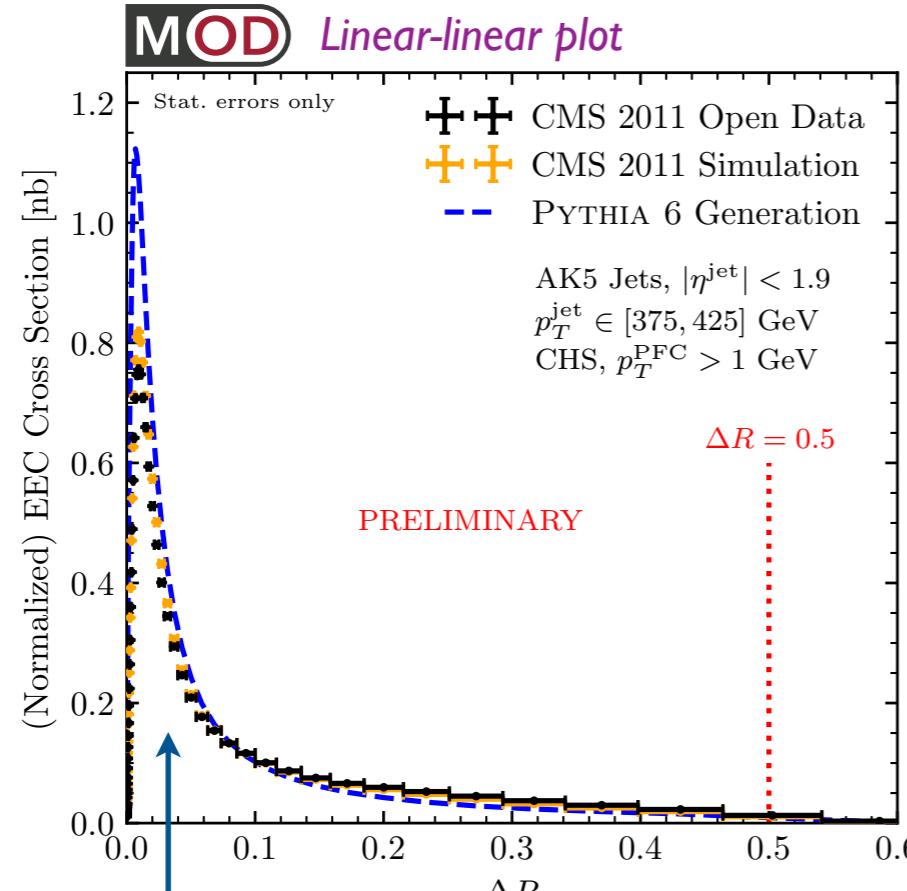


A long history in probing collinear dynamics of QCD



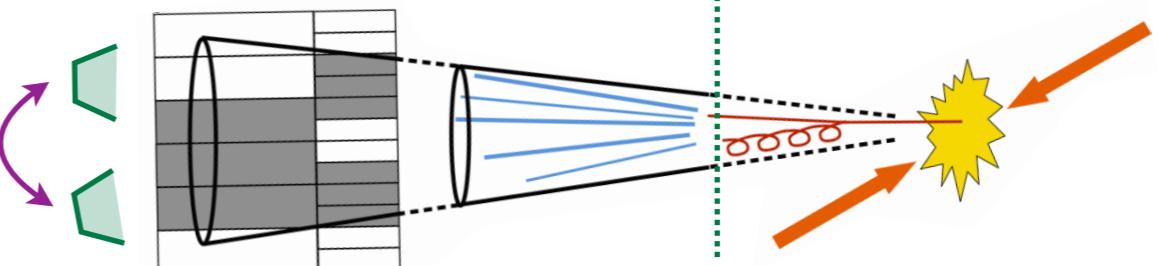
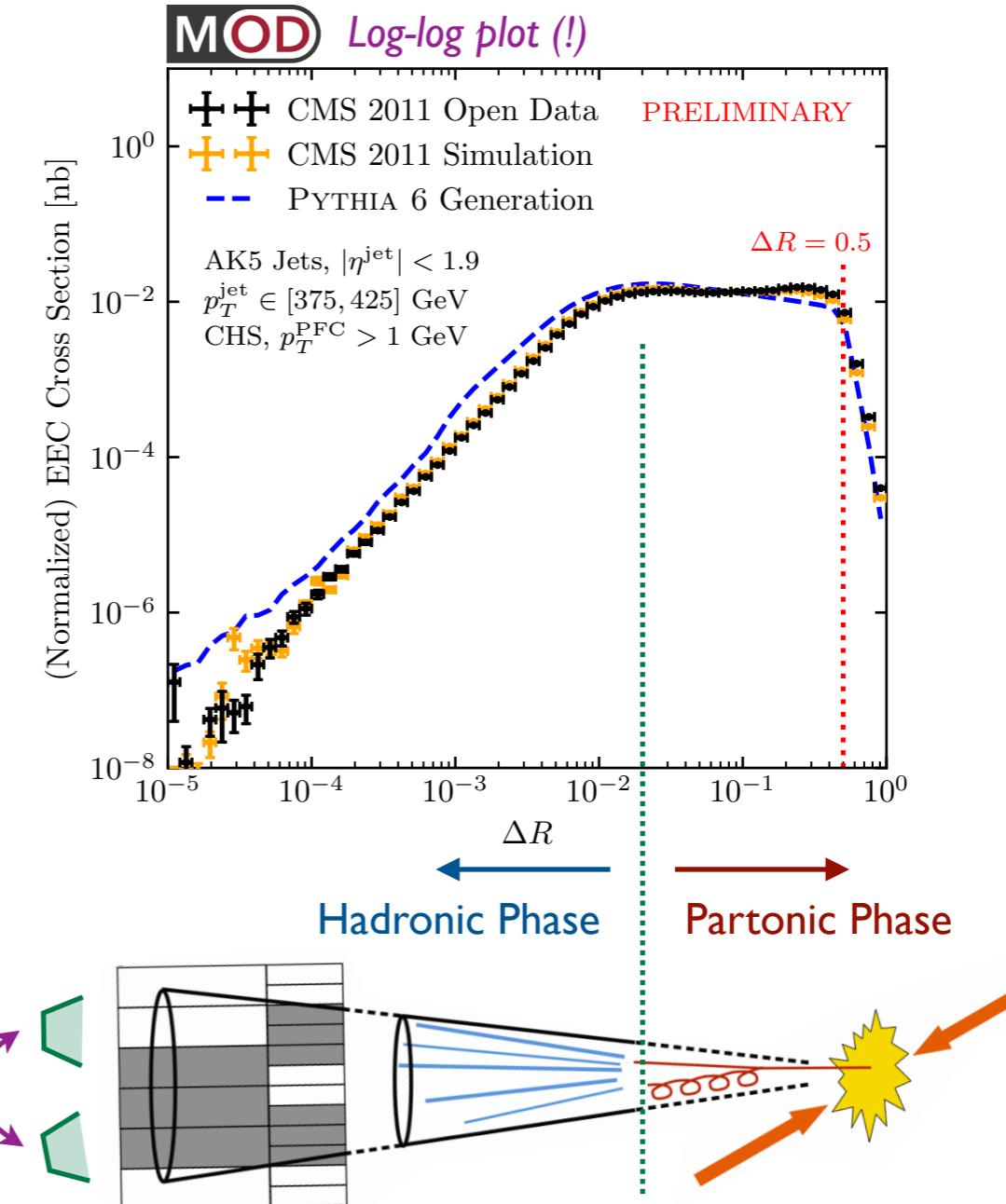
[Basham, Brown, Ellis, Love, [PRL 1978](#); ALEPH, [PLB 1991](#); see Chen, Moult, Zhang, Zhu, [PRD 2020](#)]

# QCD Phase Transition in Jets?



Are we learning something about small angle limit of QCD?

First Jet EEC Plot from the LHC (!)



[Komiske, Moult, JDT, Zhu, in progress; see talks by Moult, [BOOST 2019](#), [BOOST 2020](#)]



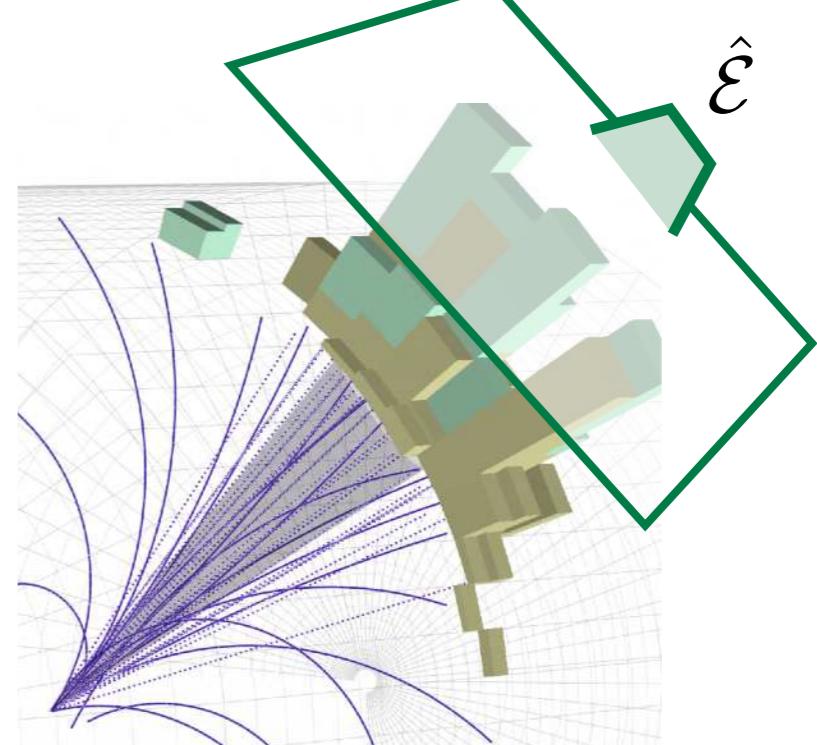
# Jets as Weighted Point Clouds

- Energy-Weighted Directions

$$\vec{p} = \{E, \hat{n}_x, \hat{n}_y, \hat{n}_z\}$$

↑      |  
Energy      Direction

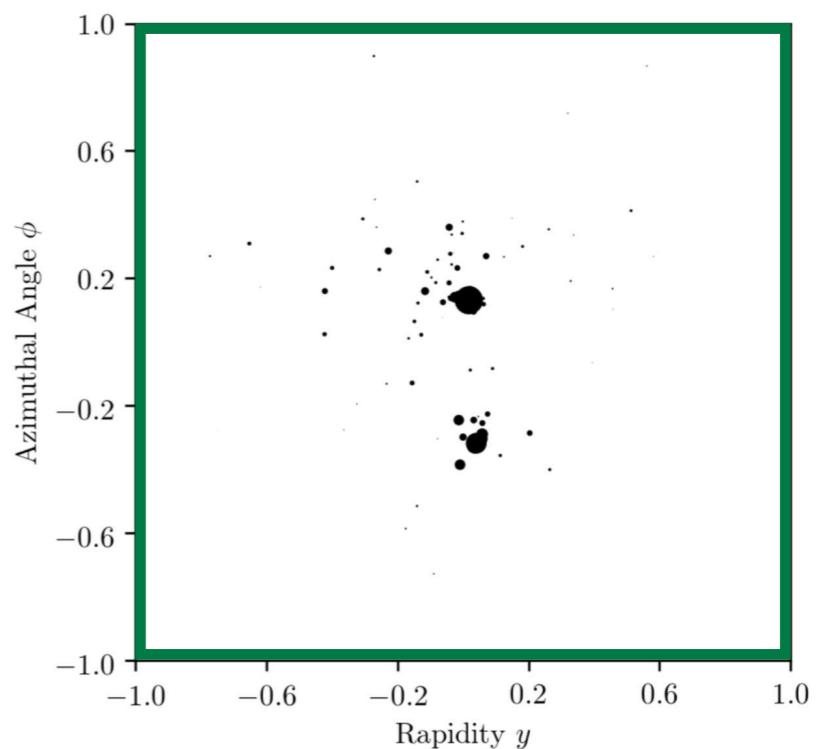
(suppressing “unsafe” charge/flavor information)

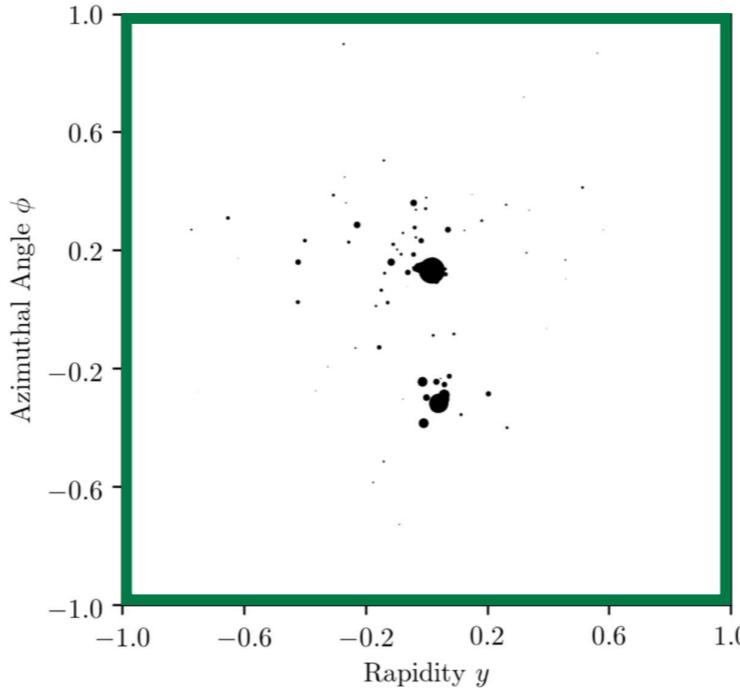


- Equivalently: Energy Density

$$\rho(\hat{n}) = \sum_{i \in \mathcal{J}} E_i \delta^{(2)}(\hat{n} - \hat{n}_i)$$

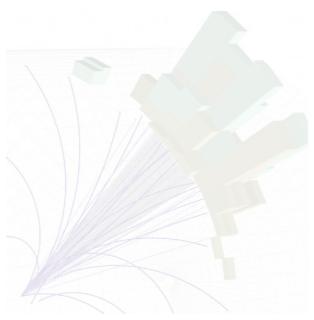
↑      ↑  
Energy      Direction



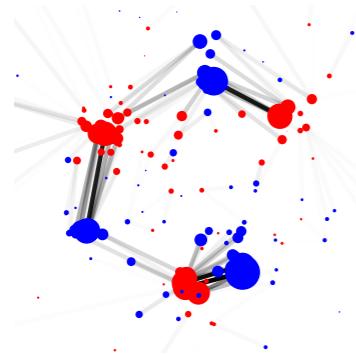


*When restricted to IRC safe information,  
jets/events are naturally represented  
as energy densities*

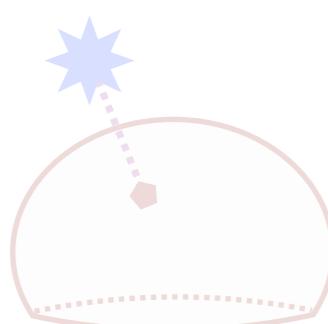
There is no obvious way to include unsafe information in this picture,  
since flow of charge/flavor is theoretically delicate



## Going with the (Energy) Flow



## The Energy Mover's Distance



## Revealing a Hidden Geometry

*If you ask your local computational geometry  
expert how to process densities...*

# The Earth Mover's Distance

See Nathaniel's talk for a more extensive history

## Optimal Transport:

[Peleg, Werman, Rom, [IEEE 1989](#); Rubner, Tomasi, Guibas, [ICCV 1998](#), [ICCV 2000](#); Pele, Werman, [ECCV 2008](#); Pele Taskar, [GSI 2013](#)]

Minimum “work” (stuff  $\times$  distance) to make one distribution look like another distribution



Déblai

Remblai

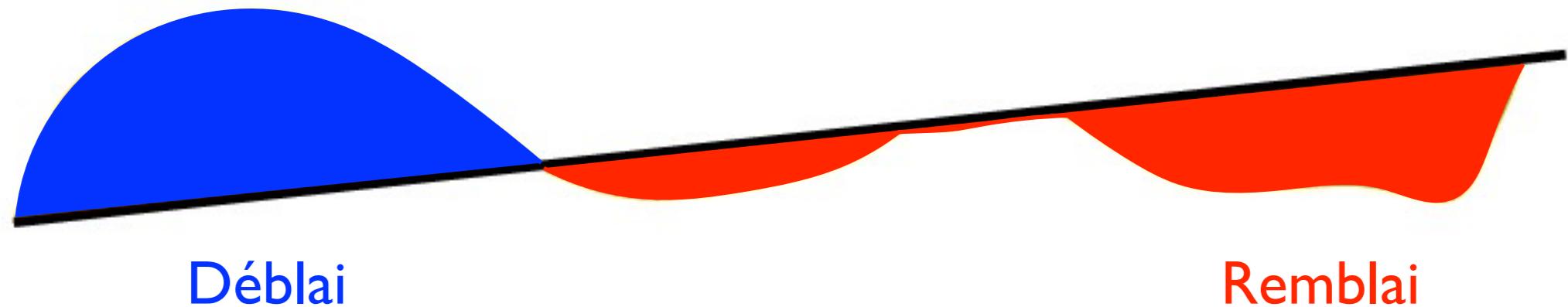
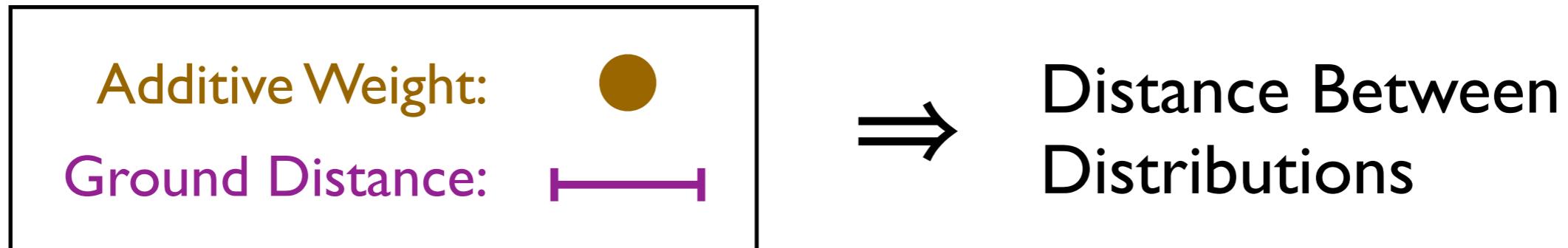
[h/t Niles-Weed, [ML4Jets 2020](#); Monge, 1781; Kantorovich, 1939; Vaserštejn, 1969; [Wikipedia](#)]

# The Earth Mover's Distance

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Minimum “work” (**stuff** × **distance**) to make  
**one distribution** look like **another distribution**



[h/t Niles-Weed, [ML4Jets 2020](#); Monge, 1781; Kantorovich, 1939; Vaserštejn, 1969; [Wikipedia](#)]

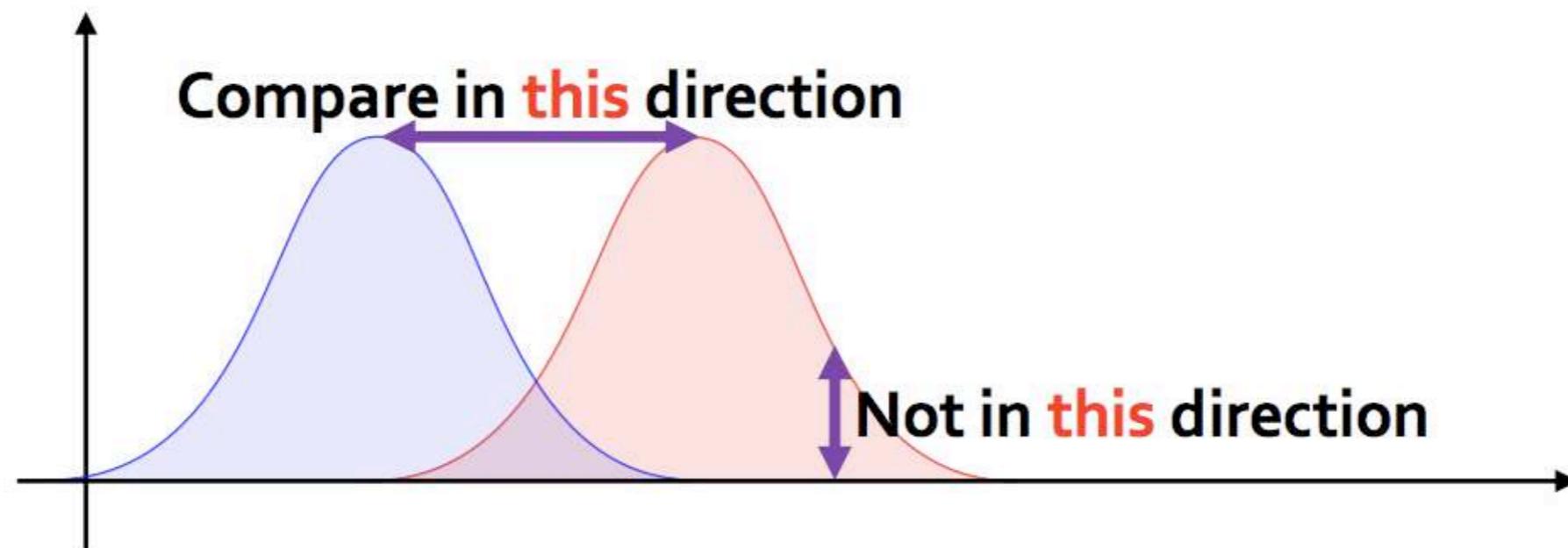
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Minimum “work” (**stuff  $\times$  distance**) to make  
**one distribution** look like **another distribution**

“Horizontal” comparison (EMD) yields better  
dynamic range than “vertical” comparison (e.g. KL)

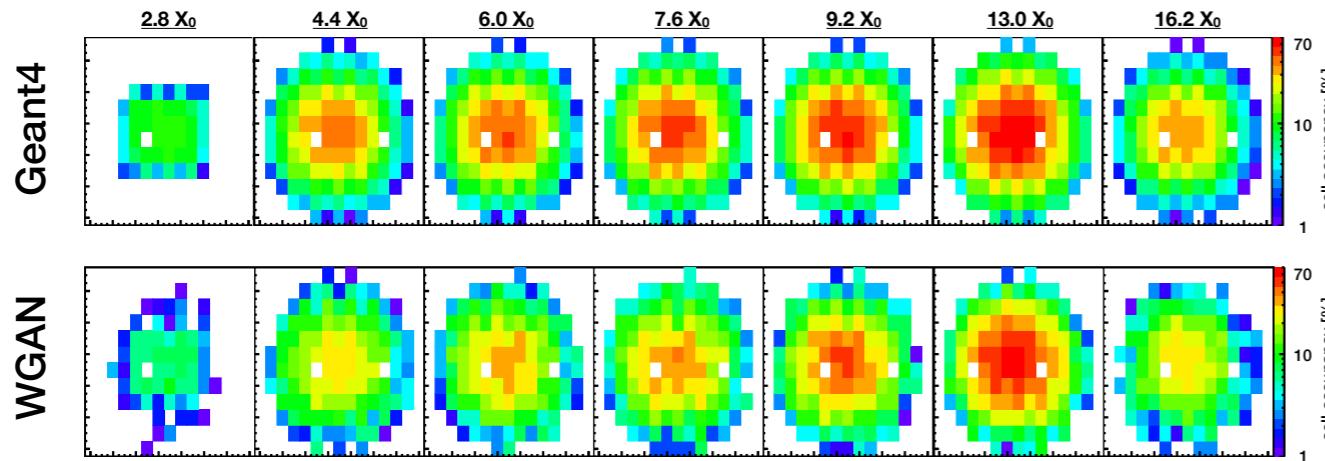


[figure from Kun, [Math n Programming](#)]

[h/t Niles-Weed, [ML4Jets 2020](#); Monge, 1781; Kantorovich, 1939; Vaserštejn, 1969; [Wikipedia](#)]

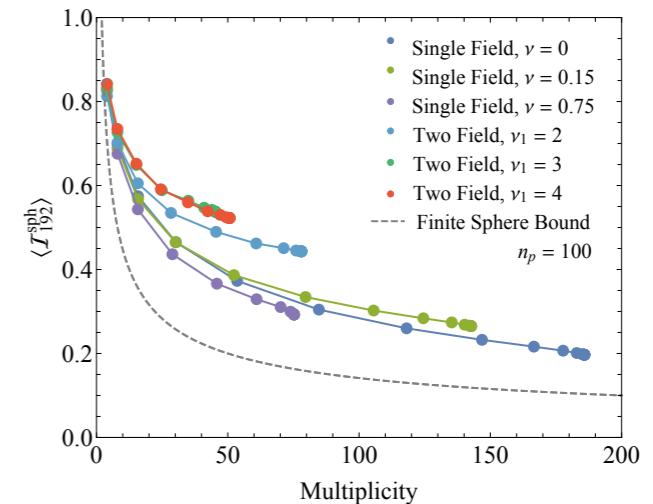
# Wasserstein in HEP

## Generative Modeling



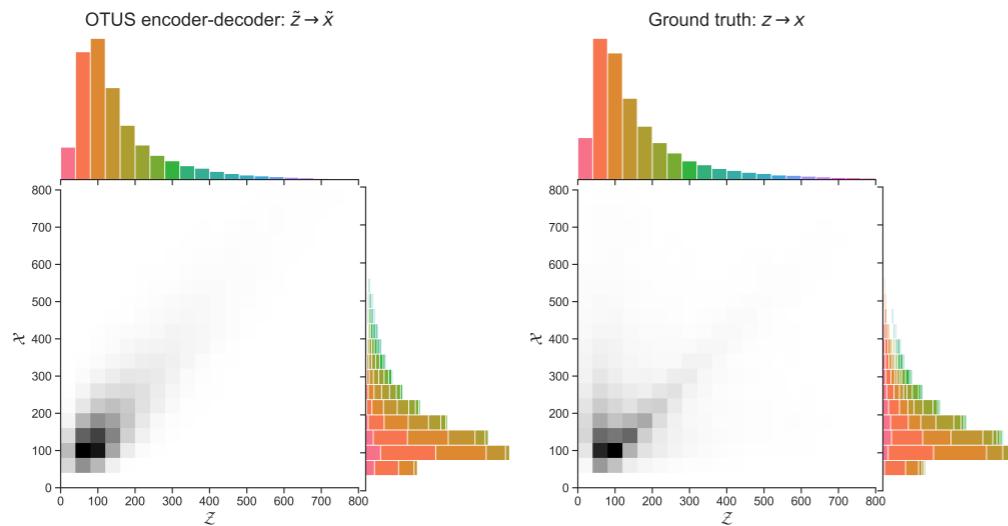
[Erdmann, Geiger, Glombitza, Schmidt, [CSBS 2018](#); Erdmann, Glombitza, Quast, [CSBS 2019](#);  
Chekalina, Orlova, Ratnikov, Ulyanov, Ustyuzhanin, Zakharov, [CHEP 2018](#)]

## BSM Characterization



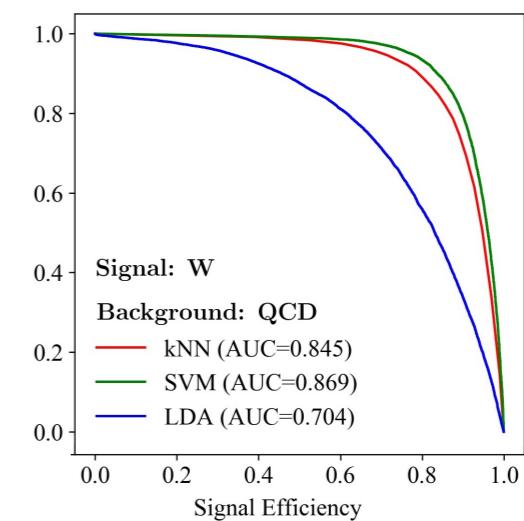
[Cesarotti, Reece, Strassler, [JHEP 2021](#), [arXiv 2020](#)]

## Estimated Simulation/Unfolding



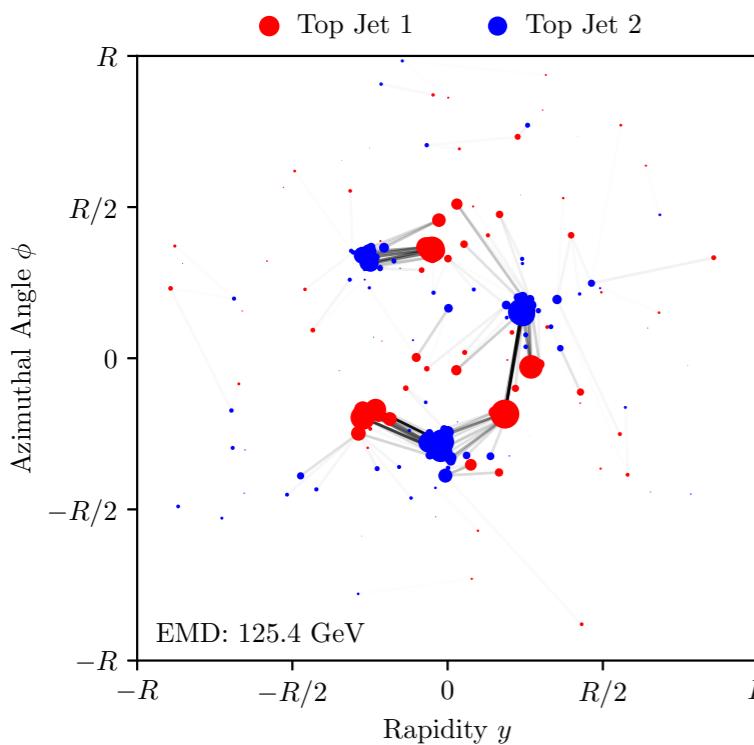
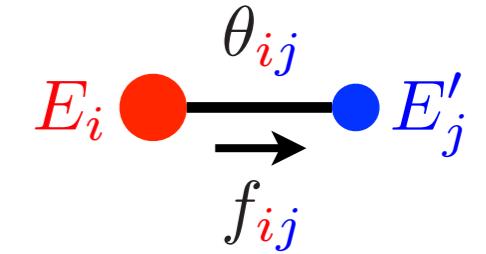
[Howard, Mandt, Whiteson, Yang, [arXiv 2021](#)]

## Jet Classification



[Cai, Cheng, Craig, Craig, [PRD 2020](#)]

# The Energy Mover's Distance

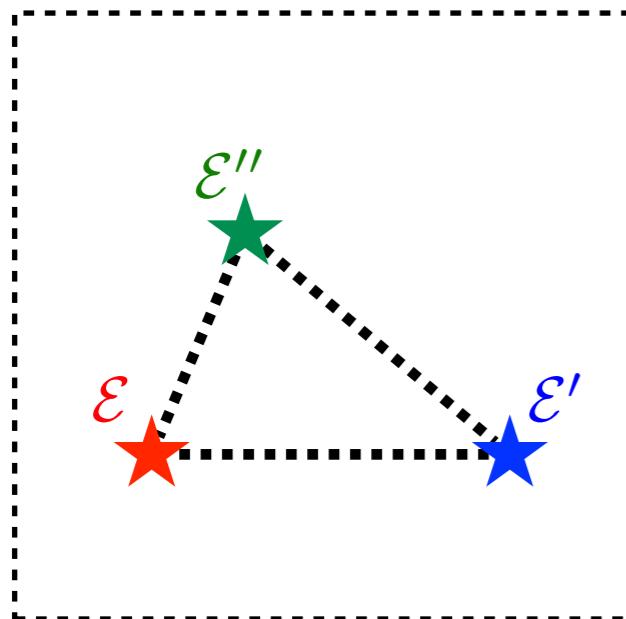


Optimal transport between energy flows...

$$\text{EMD}(\mathcal{E}, \mathcal{E}') = \min_{\{f\}} \sum_i \sum_j f_{ij} \frac{\theta_{ij}}{R} + \left| \sum_i E_i - \sum_j E'_j \right|$$

↑  
in GeV

Cost to move energy      Cost to create energy



...defines a metric on the space of events

$$0 \leq \text{EMD}(\mathcal{E}, \mathcal{E}') \leq \text{EMD}(\mathcal{E}, \mathcal{E}'') + \text{EMD}(\mathcal{E}', \mathcal{E}'')$$

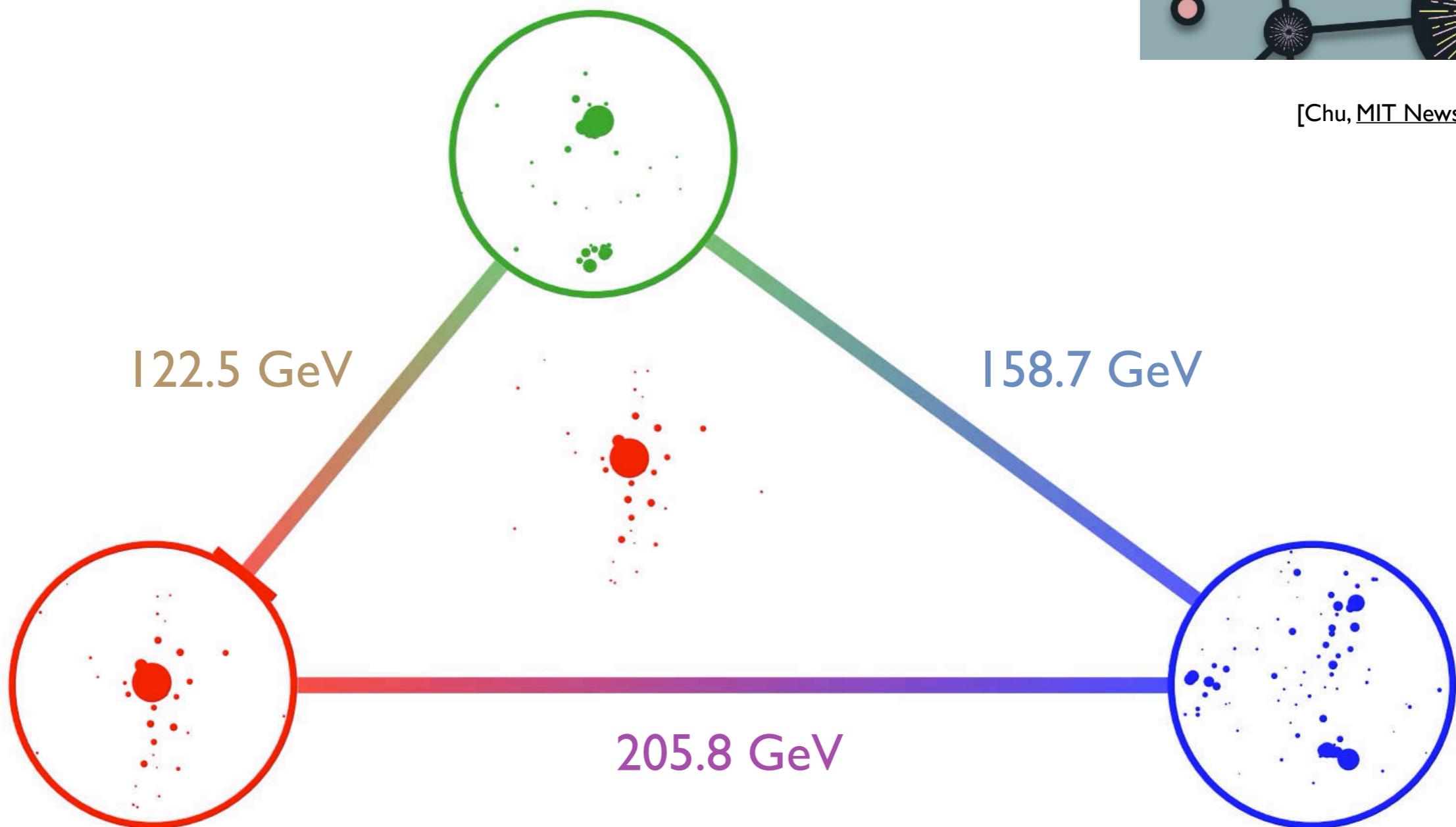
(assuming  $R \geq \theta_{\max}/2$ , i.e.  $R \geq$  jet radius for conical jets)

[Komiske, Metodiev, JDT, [PRL 2019](#);  
 see also Pele, Werman, [ECCV 2008](#); Pele, Taskar, [GSI 2013](#);  
 [see flavored variant in Crispim Romão, Castro, Milhano, Pedro, Vale, [EPJC 2021](#)]  
 [see computational speed up in Cai, Cheng, Craig, [PRD 2020](#)]

# Similarity of Three Energy Flows?



[Chu, MIT News July 2019]



[Komiske, Metodiev, JDT, [PRL 2019](#); code at Komiske, Metodiev, JDT, [energyflow.network](#);  
see alternative graph network approach in Mullin, Pacey, Parker, White, Williams, [JHEP 2021](#)]

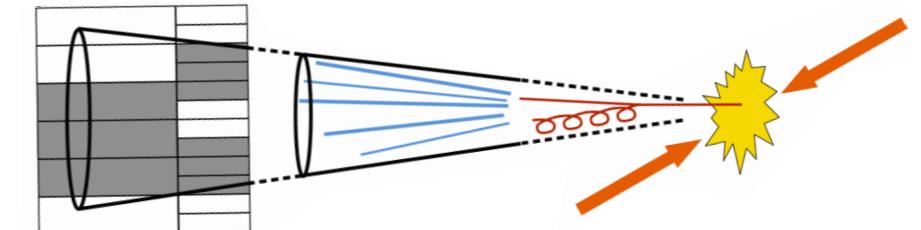
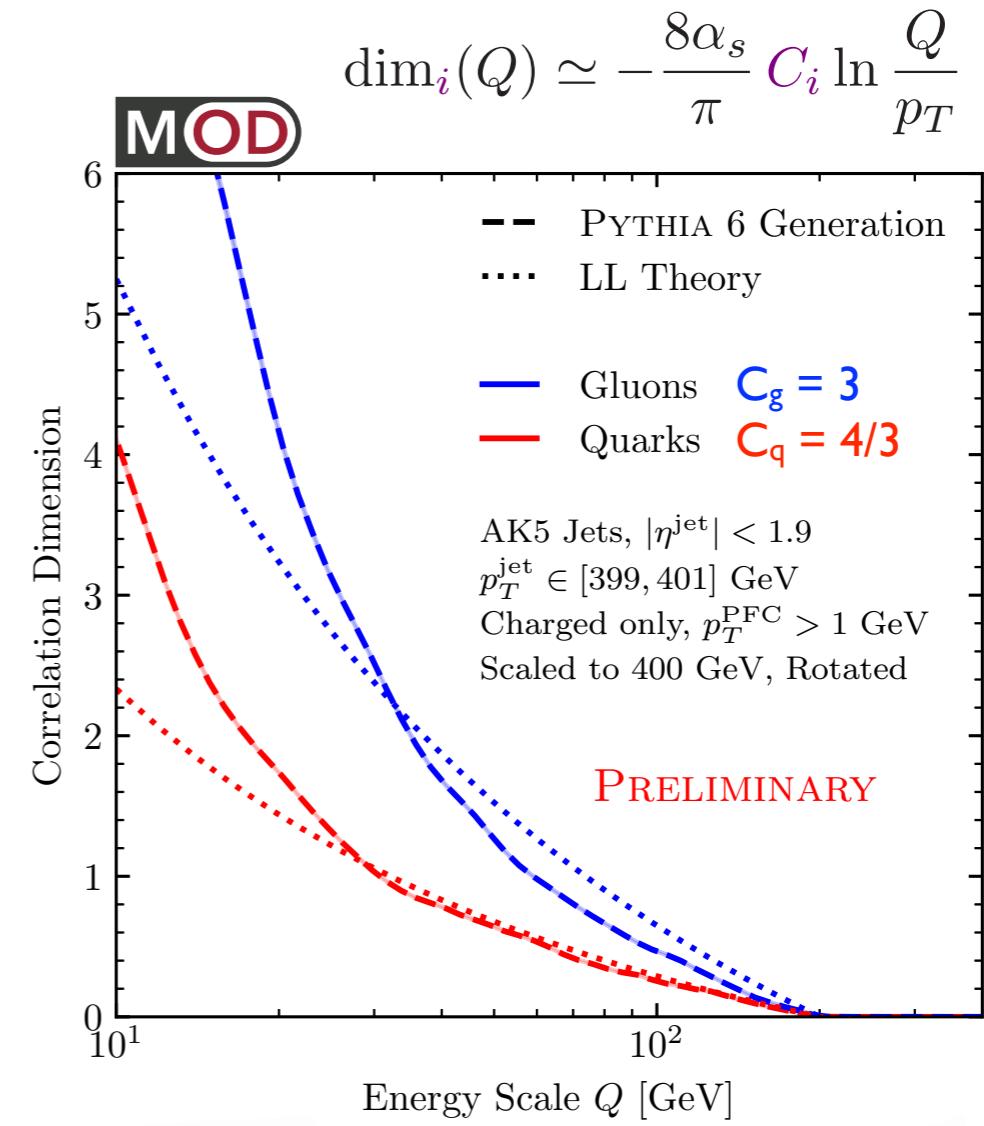
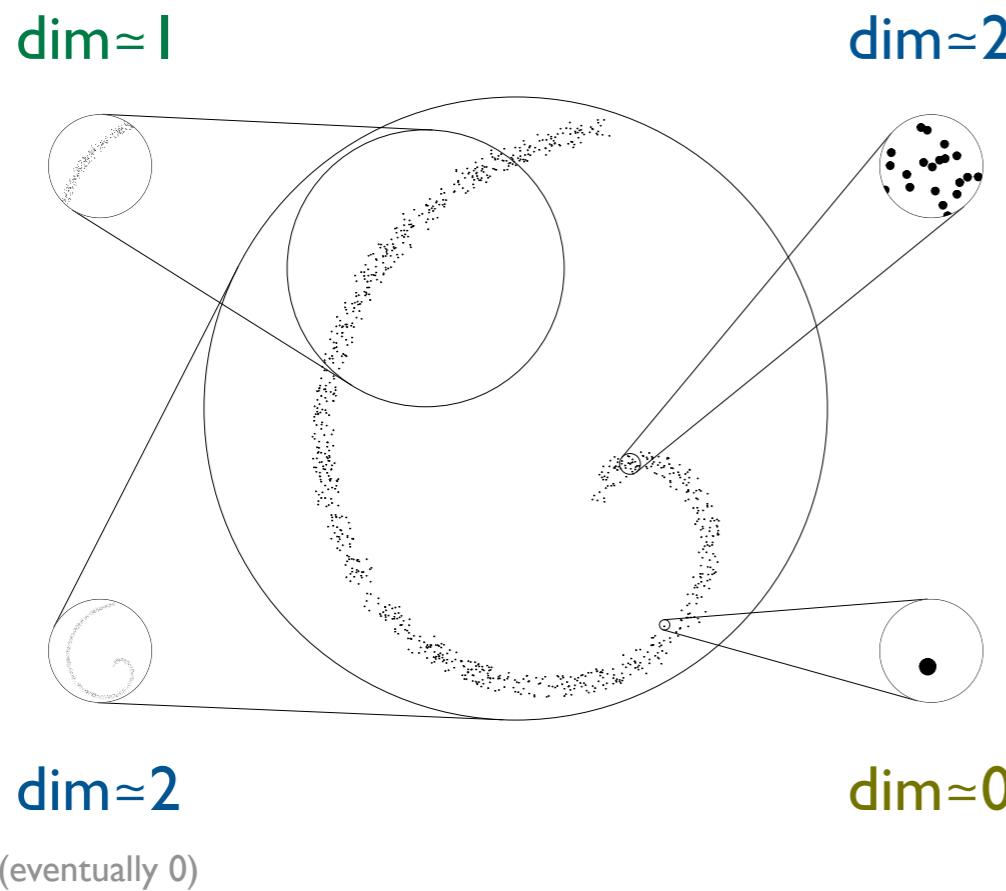
# Dimensionality of Space of Jets



$$N_{\text{neighbors}}(r) \sim r^{\dim}$$

$$\Rightarrow \dim(r) \sim r \frac{\partial}{\partial r} \ln N_{\text{neighbors}}(r)$$

[Grassberger, Procaccia, [PRL 1983](#); Kégl, [NIPS 2002](#)]



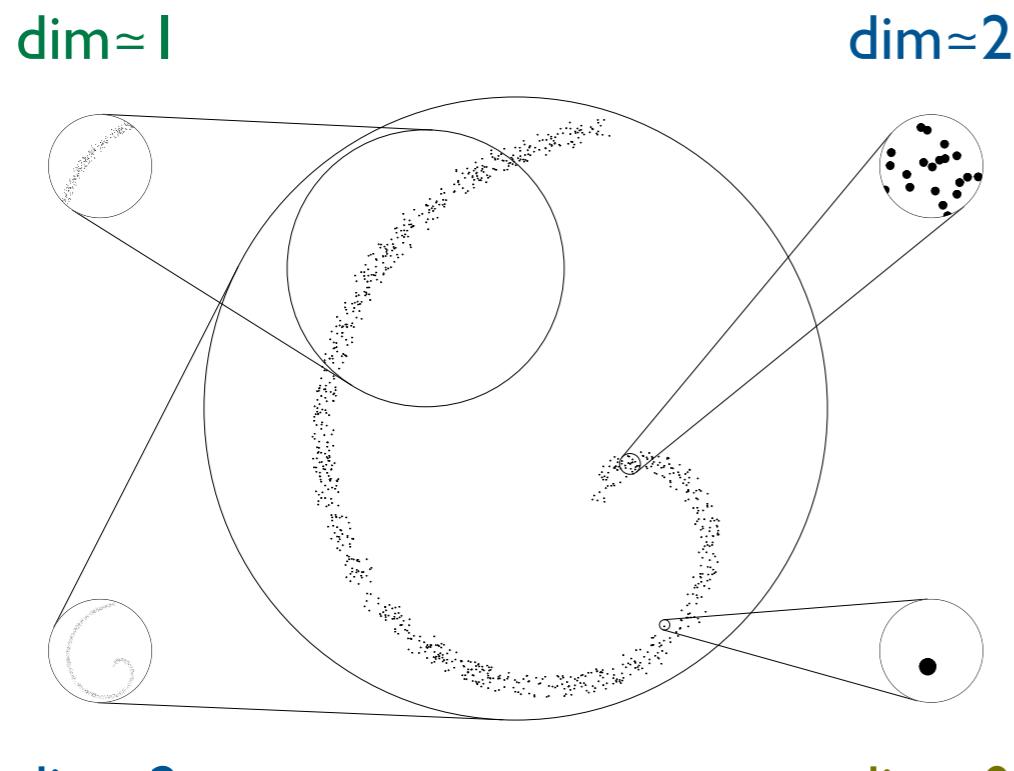
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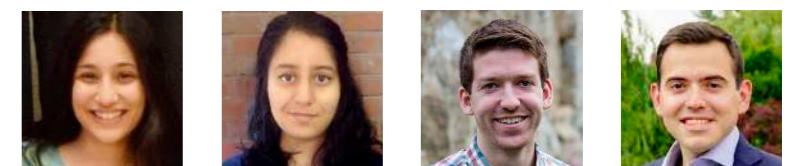
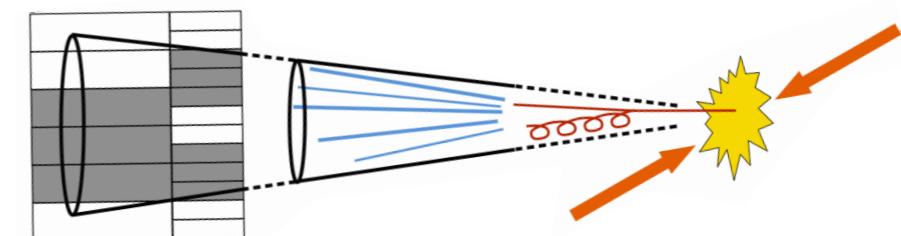
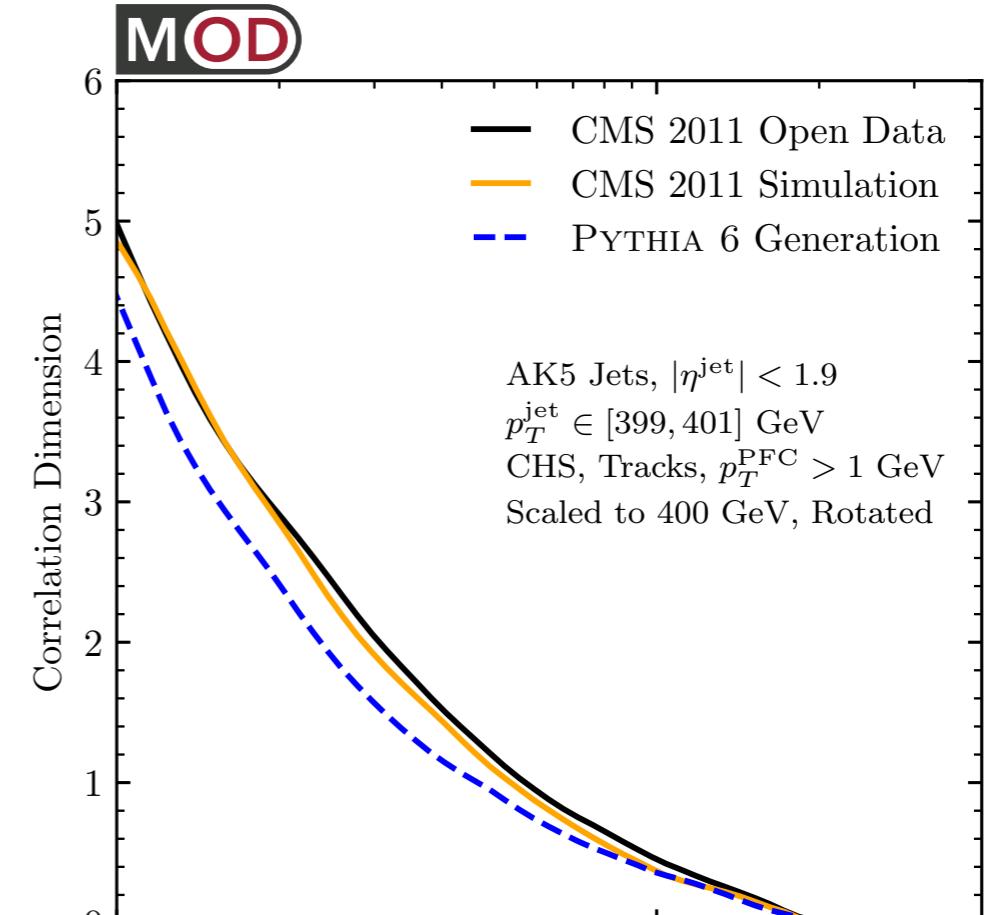
$$\Rightarrow \dim(r) \sim r \frac{\partial}{\partial r} \ln N_{\text{neighbors}}(r)$$

[Grassberger, Procaccia, [PRL 1983](#); Kégl, [NIPS 2002](#)]



(eventually 0)

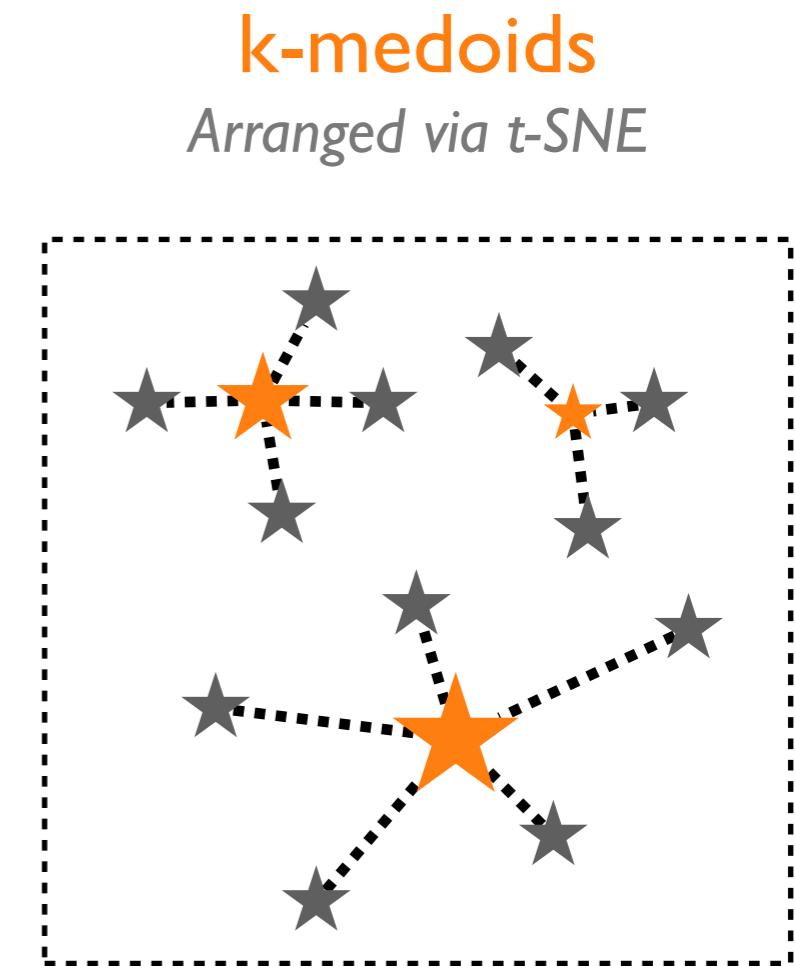
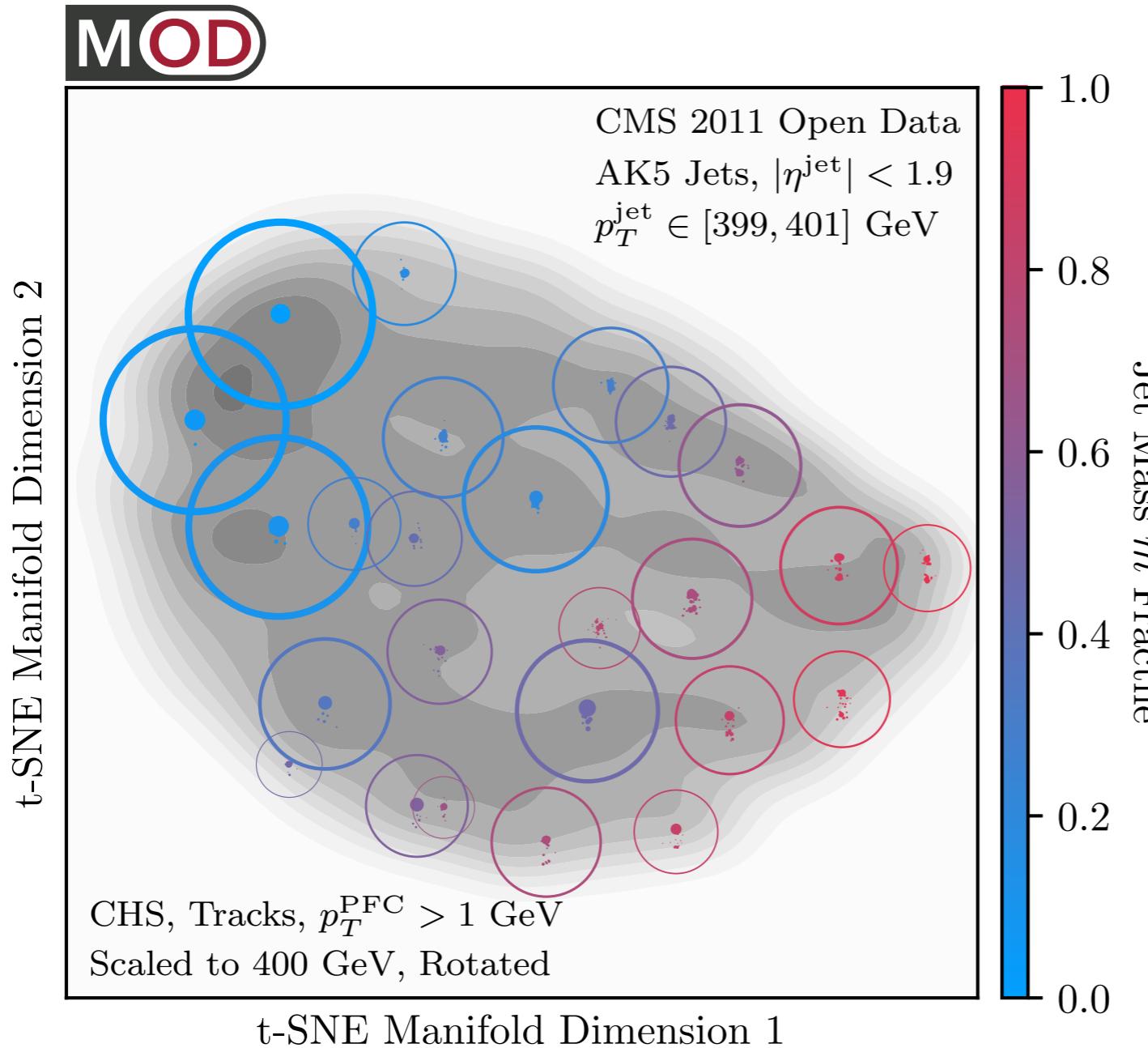
[Komiske, Mastandrea, Metodiev, Naik, [JDT, PRD 2020](#);  
using [CMS Open Data](#)]



# Most Representative Jets



[<http://opendata.cern.ch/>]

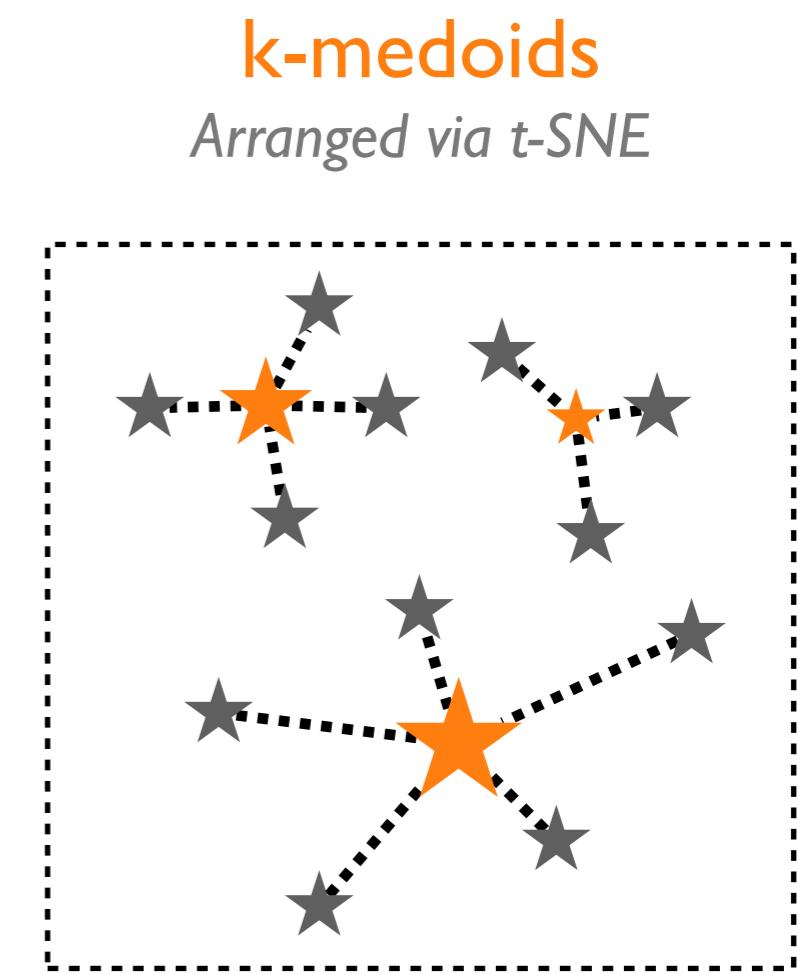
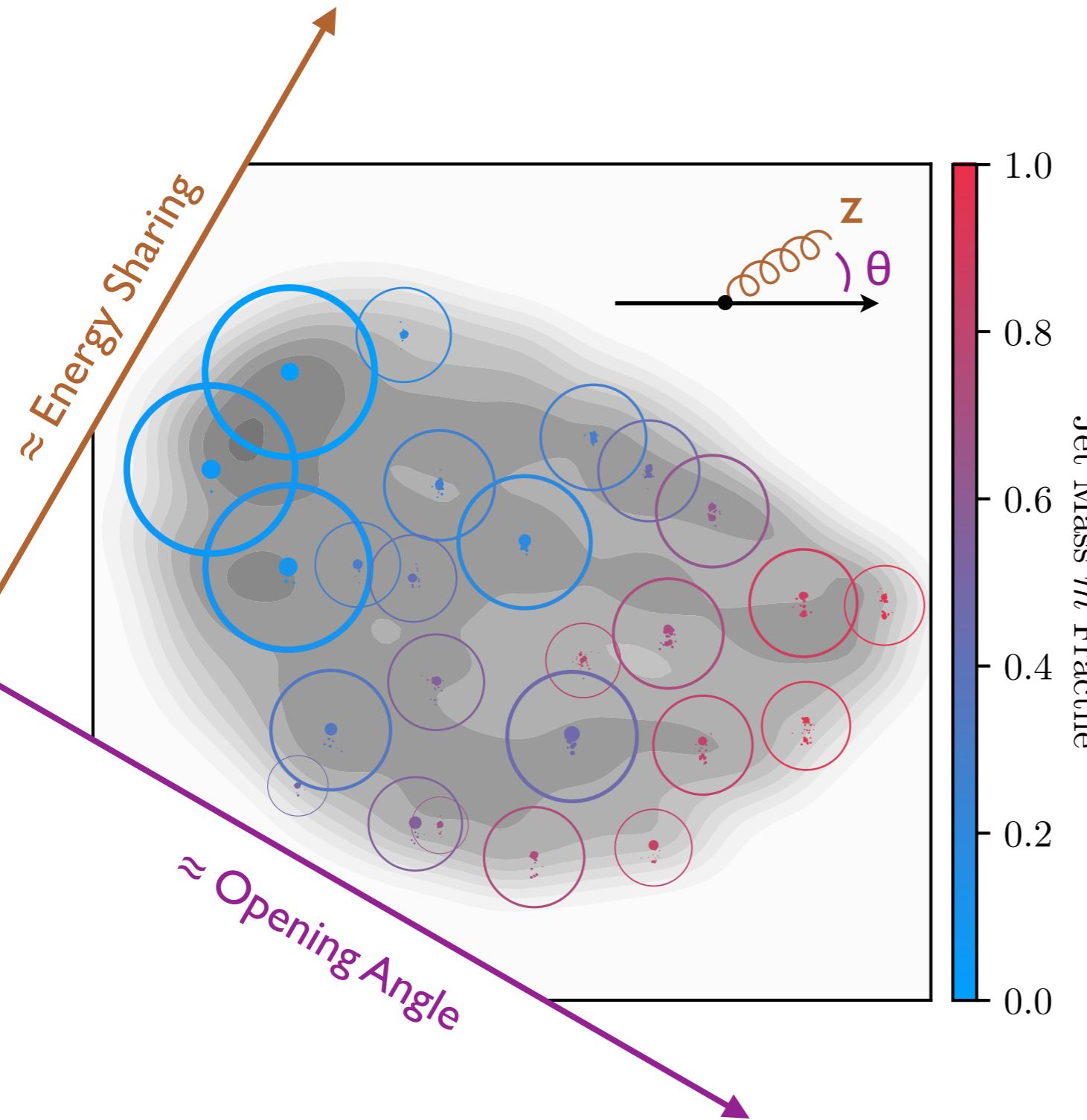


[Komiske, Mastandrea, Metodiev, Naik, JDT, [PRD 2020](#); using van der Maaten, Hinton, [JMLR 2008](#)]

# Most Representative Jets

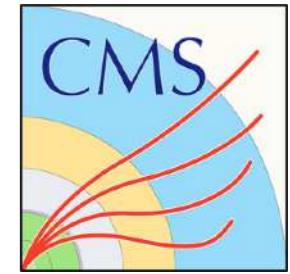


[<http://opendata.cern.ch/>]

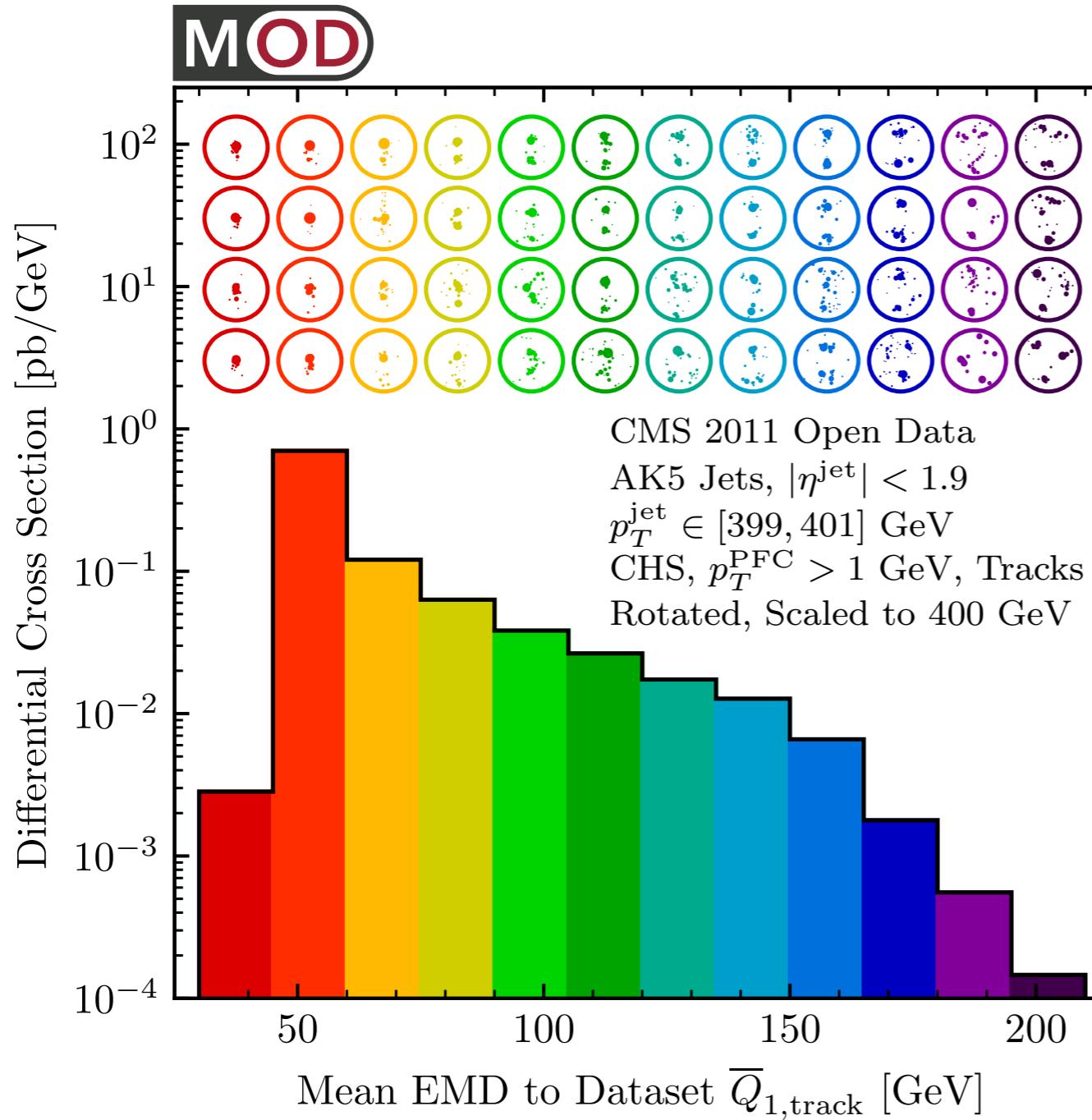


[Komiske, Mastandrea, Metodiev, Naik, JDT, PRD 2020; using van der Maaten, Hinton, JMLR 2008]

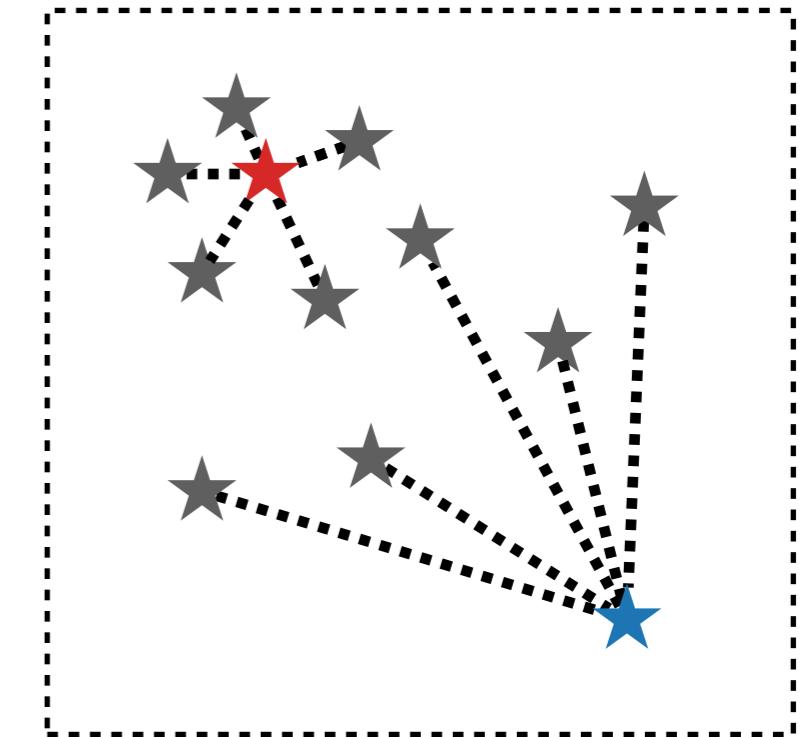
# Least Representative Jets



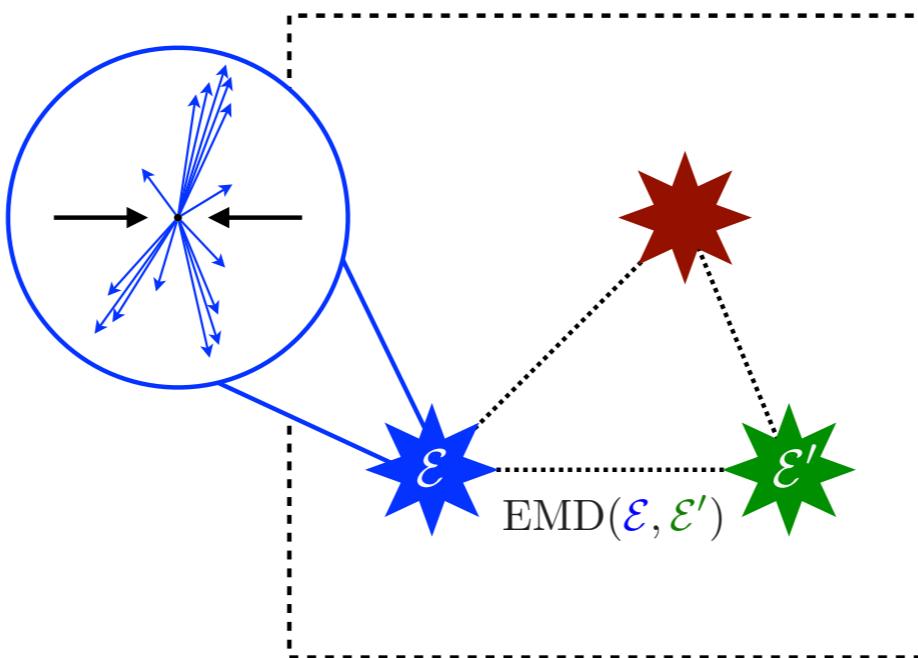
[<http://opendata.cern.ch/>]



New Physics?  
Or tails of QCD?

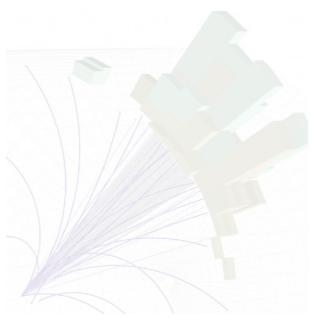


[Komiske, Mastandrea, Metodiev, Naik, JDT, [PRD 2020](#)]

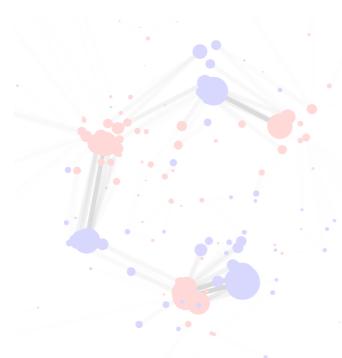


*Viewed through the data science lens,  
the EMD unlocks a suite of  
geometric analysis strategies*

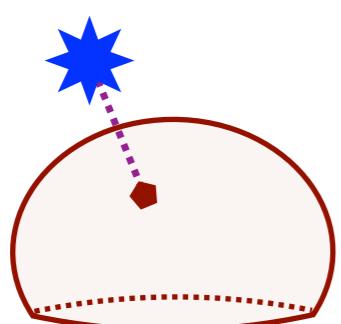
At this point, it should not be obvious why optimal transport distances should be particularly well-suited to collider applications



## Going with the (Energy) Flow



## The Energy Mover's Distance

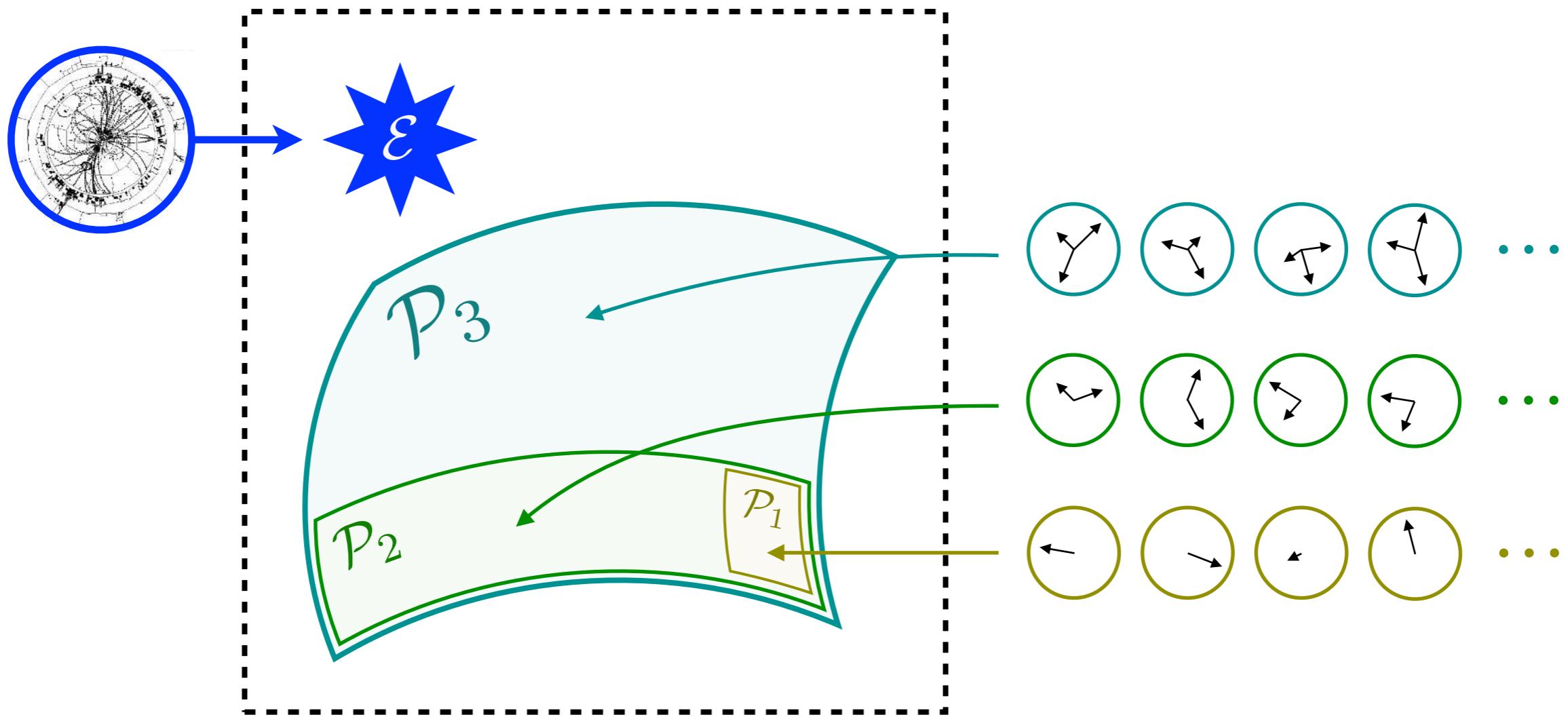


## Revealing a Hidden Geometry

*Given a metric space, the first geometric object  
you might think to construct is...*

# Introducing N-particle Manifolds

$\mathcal{P}_N$  = set of all N-particle configurations



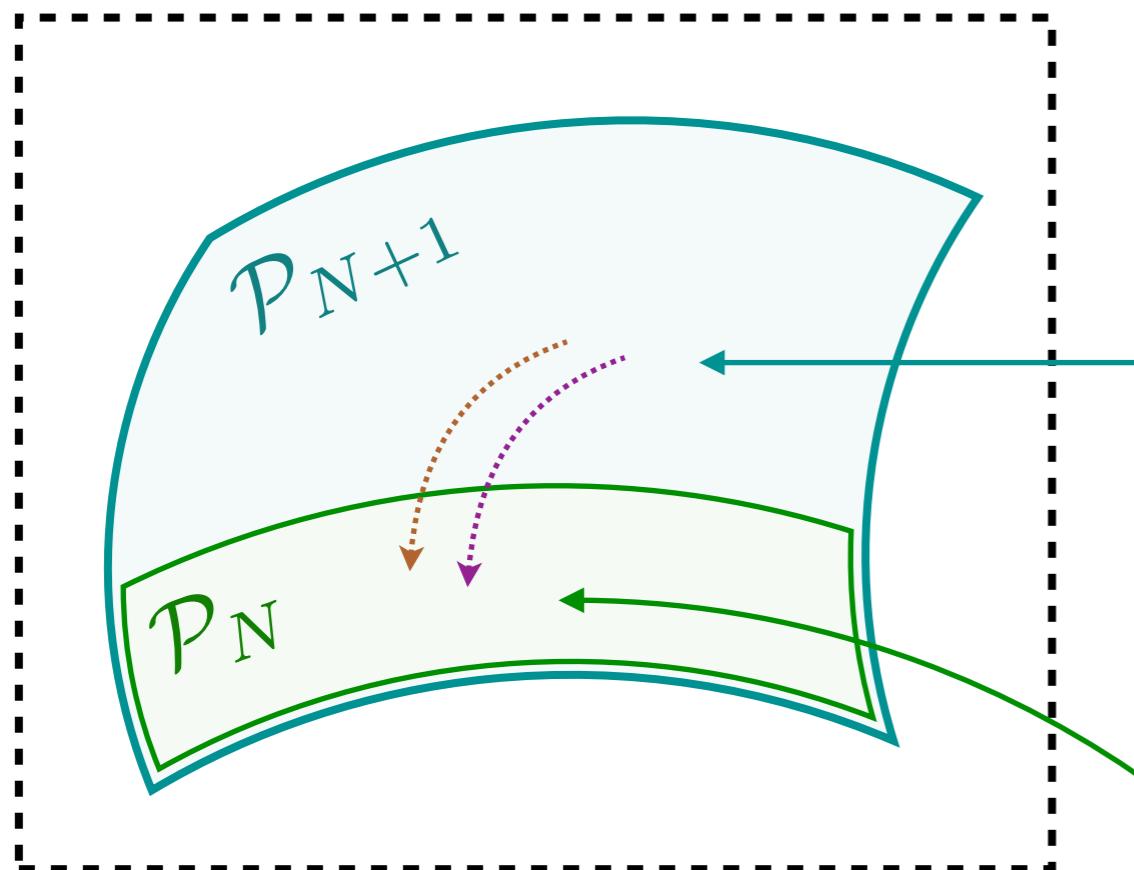
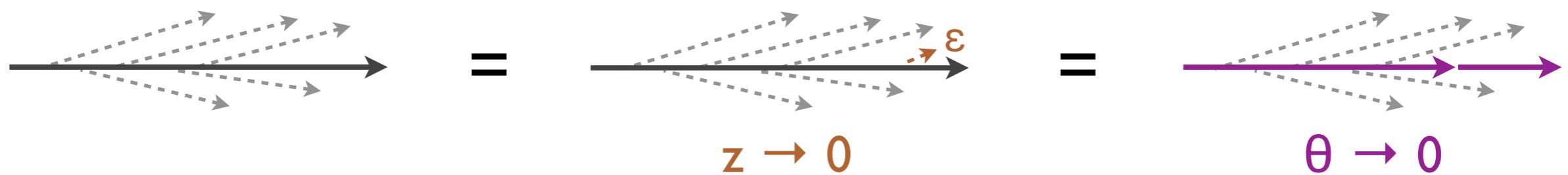
$\mathcal{P}_N \supset \mathcal{P}_{N-1} \supset \dots \supset \mathcal{P}_2 \supset \mathcal{P}_1$  by soft/collinear limits

[see related discussion in Larkoski, Melia, [PRD 2020](#)]

# When are Two Events the Same?

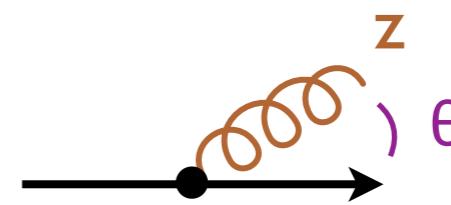
$$\mathcal{E}(\hat{n}) = \sum_i E_i \delta(\hat{n} - \hat{n}_i)$$

*Energy Flow unchanged by infinitesimal soft/collinear emissions*



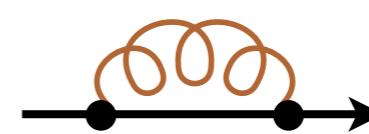
Infrared divergences “live” together!

Real:



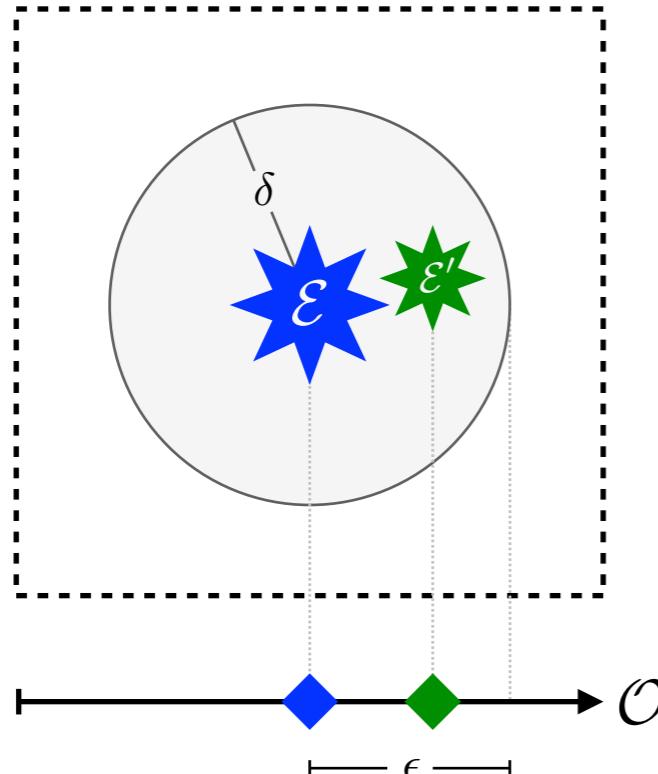
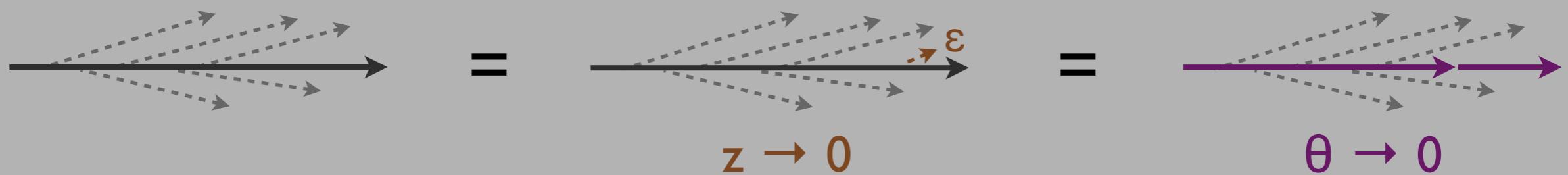
$$dP_{i \rightarrow ig} \simeq \frac{2\alpha_s}{\pi} C_i \frac{dz}{z} \frac{d\theta}{\theta}$$

Virtual:



# When are Two Events the Same?

*Energy Flow unchanged by infinitesimal soft/collinear emissions*



## Infrared & Collinear Safety

≈ calculable in perturbative quantum field theory

*is\** ← (see backup for subtleties)

## Continuity in EMD Space

[Komiske, Metodiev, JDT, [JHEP 2020](#)]

[Sterman, Weinberg, [PRL 1977](#); Sterman, [PRD 1979](#)]

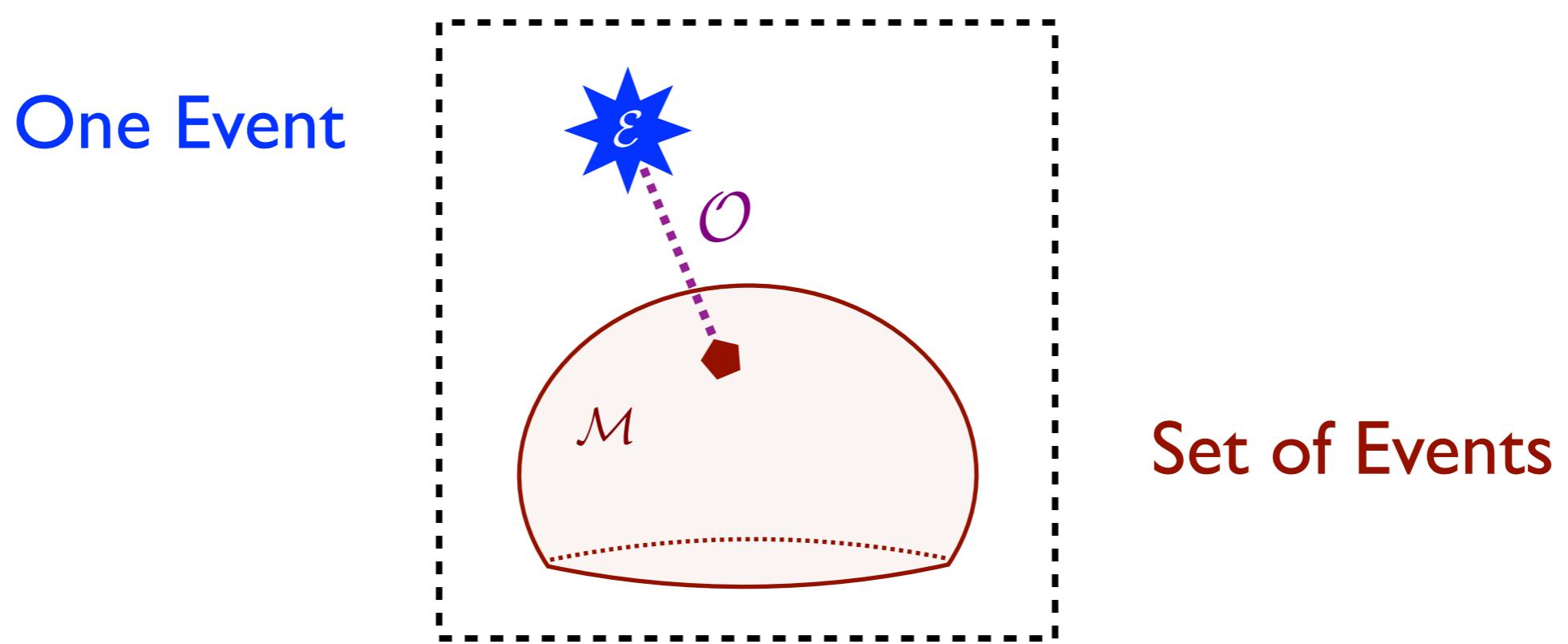
[see also Banfi, Salam, Zanderighi, [JHEP 2005](#); Larkoski, Marzani, JDT, [PRD 2015](#)]

*The EMD seems to define the “natural” geometry for massless gauge theories*

Open question: Can you define  $|\mathcal{M}_{AB \rightarrow 12\dots n}|^2$  directly in this space?

What does it mean to “integrate” in this space?

# Manifolds for Observables



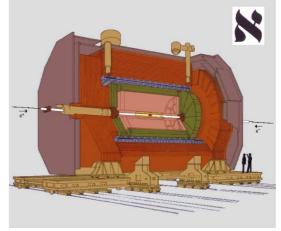
Distance of Closest Approach  $\Rightarrow$  Observable

$$O(\mathcal{E}) = \min_{\mathcal{E}' \in \mathcal{M}} \text{EMD}(\mathcal{E}, \mathcal{E}')$$

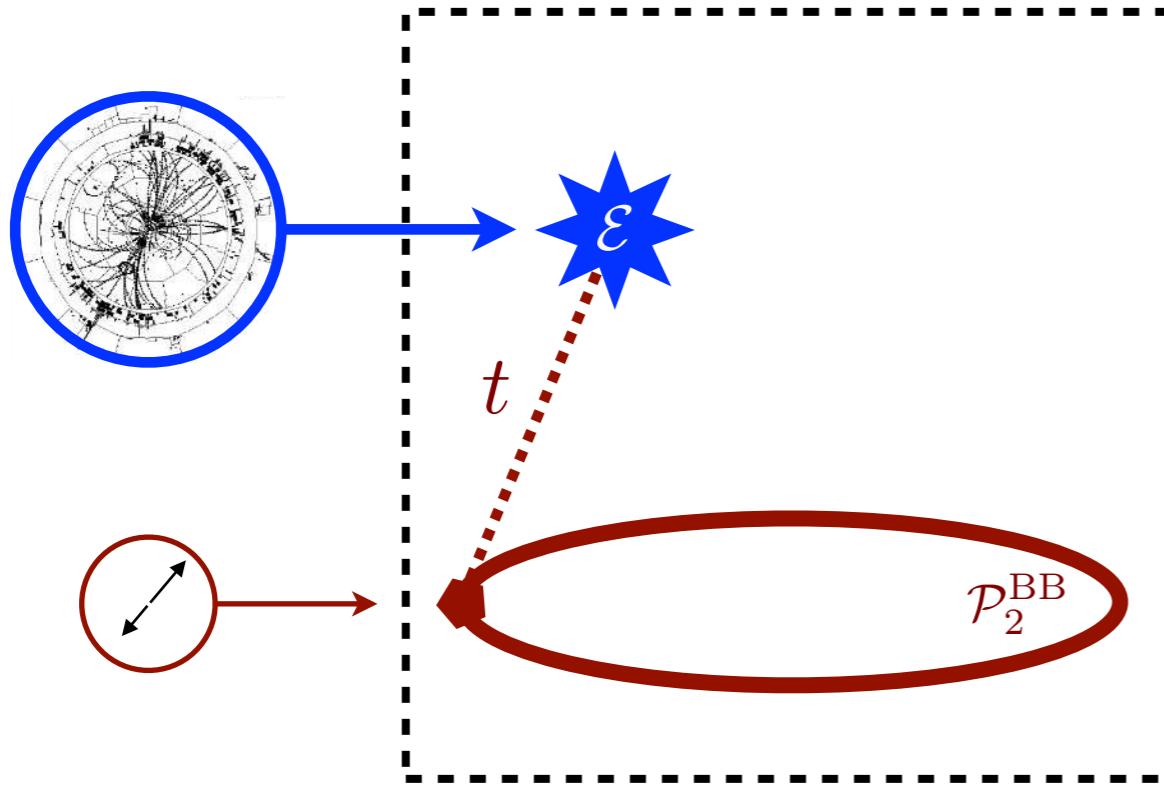
[Komiske, Metodiev, JDT, [JHEP 2020](#)]

# E.g. Thrust

How dijet-like is an event?



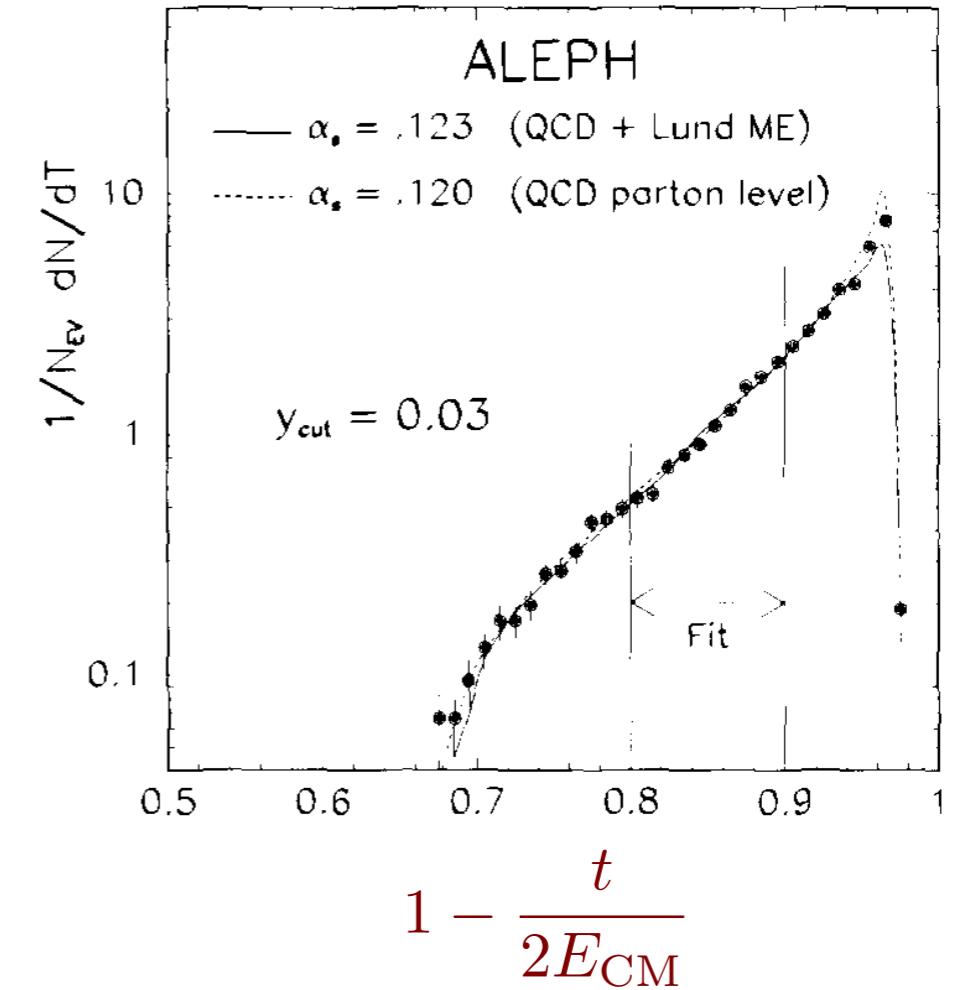
$$t(\mathcal{E}) = \min_{\mathcal{E}' \in \mathcal{P}_2^{\text{BB}}} \text{EMD}_2(\mathcal{E}, \mathcal{E}')$$



All Back-to-Back Two Particle Configurations

$$\mathcal{P}_2^{\text{BB}} = \left\{ \begin{array}{c} \text{red circles with internal arrows} \\ \cdots \end{array} \right\}$$

(using  $\beta=2$  EMD variant)



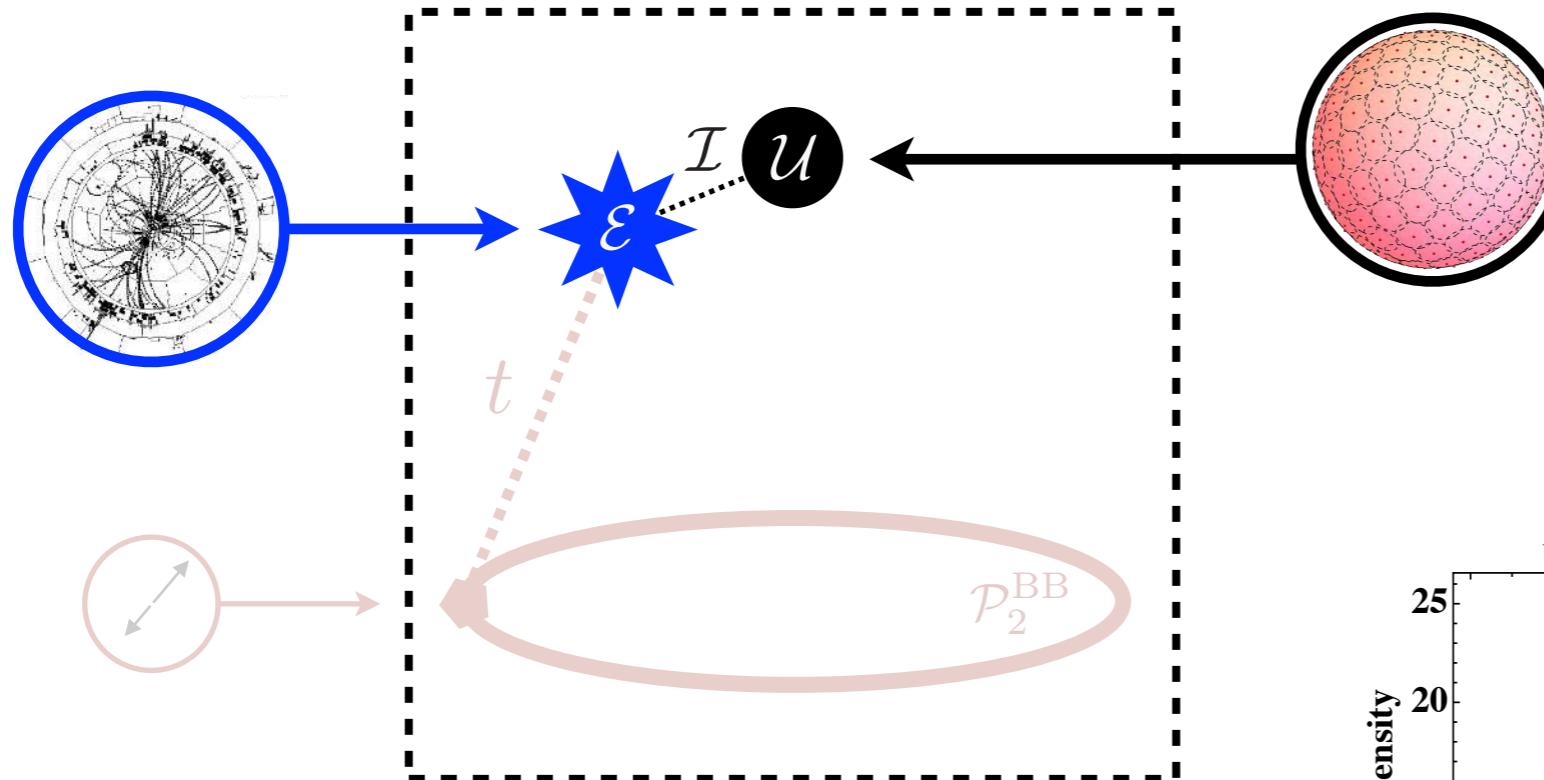
$$1 - \frac{t}{2E_{\text{CM}}}$$

$$\text{cf. } T(\mathcal{E}) = \max_{\hat{n}} \frac{\sum_i |\vec{p}_i \cdot \hat{n}|}{\sum_j |\vec{p}_j|}$$

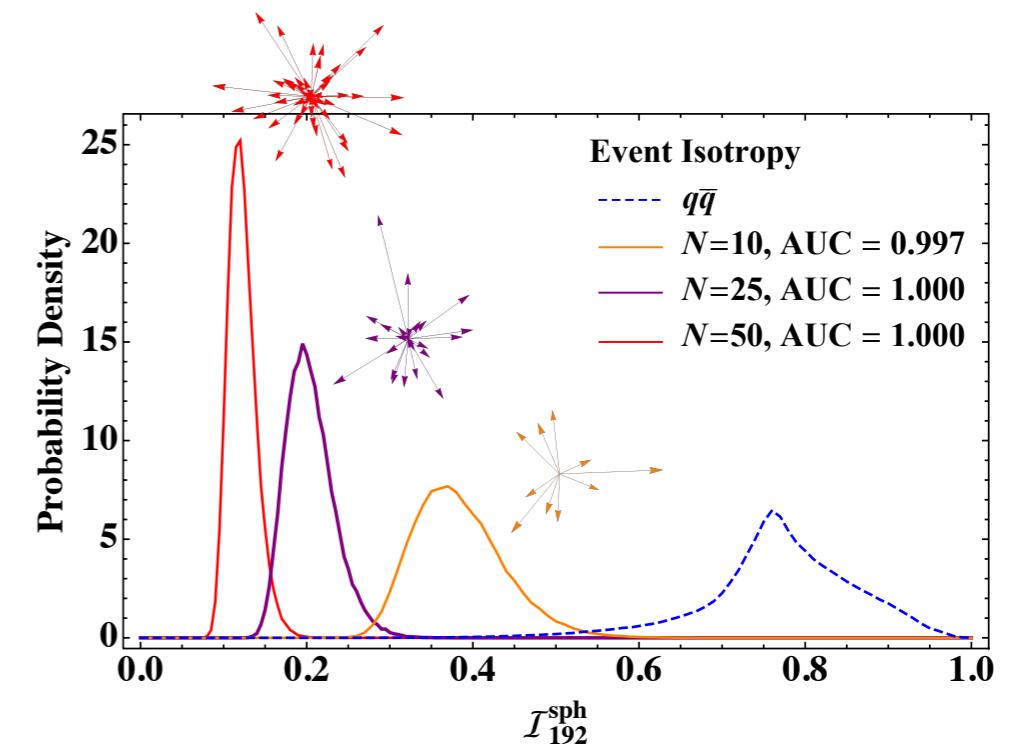
[Komiske, Metodiev, JDT, JHEP 2020]  
 [Brandt, Peyrou, Sosnowski, Wroblewski, PL 1964; Farhi, PRL 1977; ALEPH, PLB 1991]

# New! Event Isotropy

How isotropic is an event?



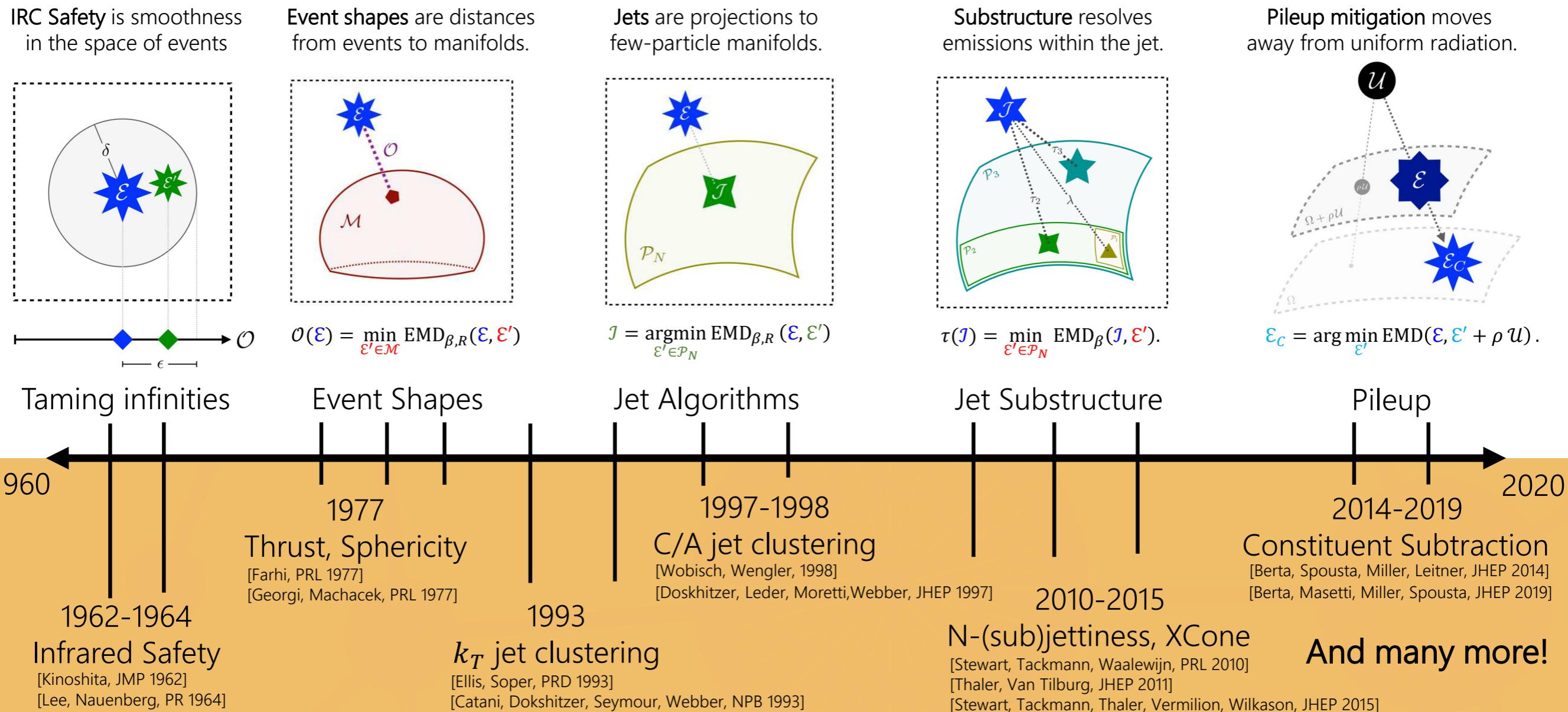
$$\mathcal{I}(\mathcal{E}) = \text{EMD}(\mathcal{E}, \mathcal{U})$$



[Cesarotti, JDT, [JHEP 2020](#);  
see also Cesarotti, Reece, Strassler, [arXiv 2020](#)]



# Six Decades of Collider Physics Translated into a New Geometric Language!

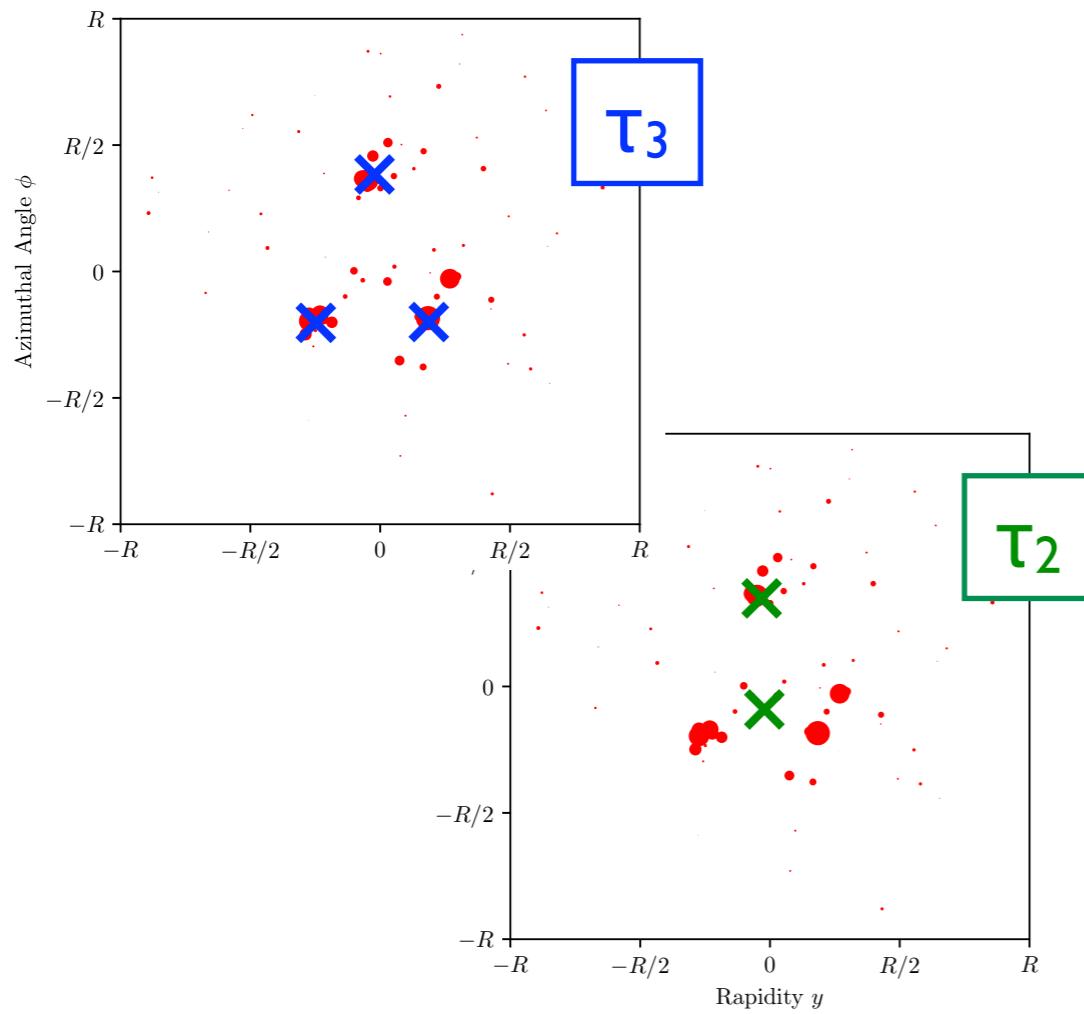


[timeline from Eric Metodiev]

# N-subjettiness

*Ubiquitous jet substructure observable used for almost a decade...*

$$\tau_N(\mathcal{J}) = \min_{N \text{ axes}} \sum_i E_i \min \{\theta_{1,i}, \theta_{2,i}, \dots, \theta_{N,i}\}$$



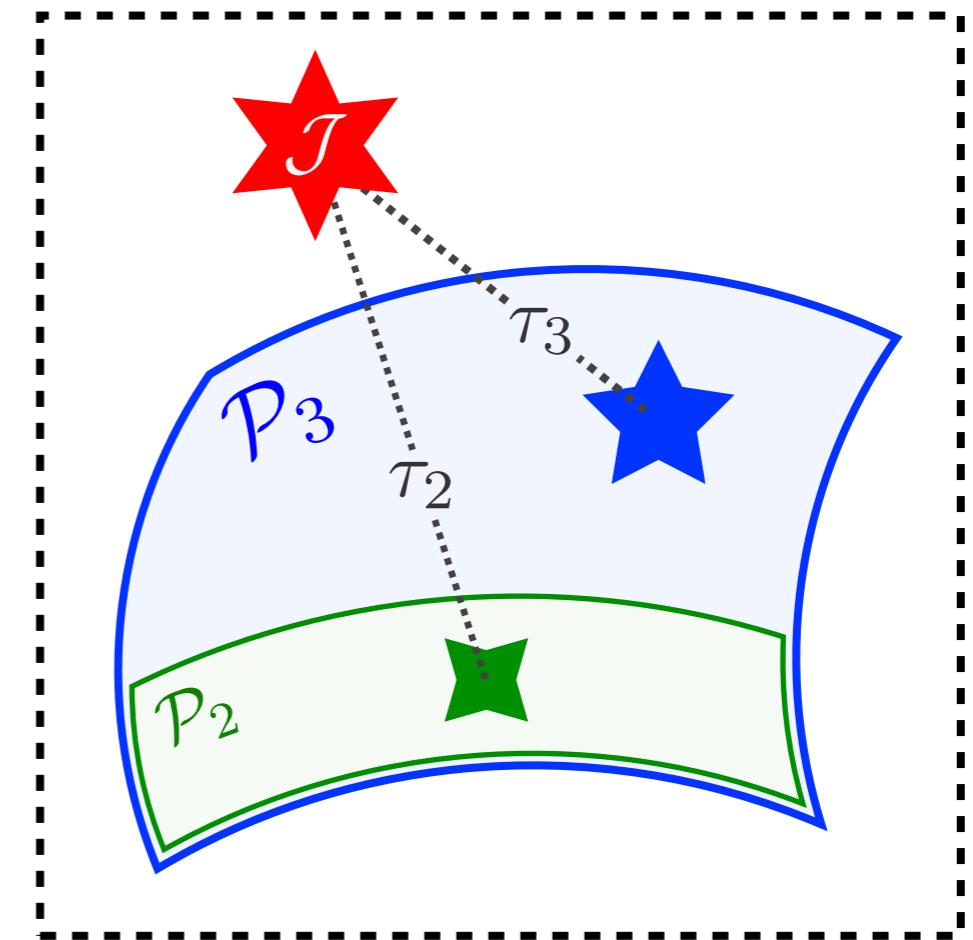
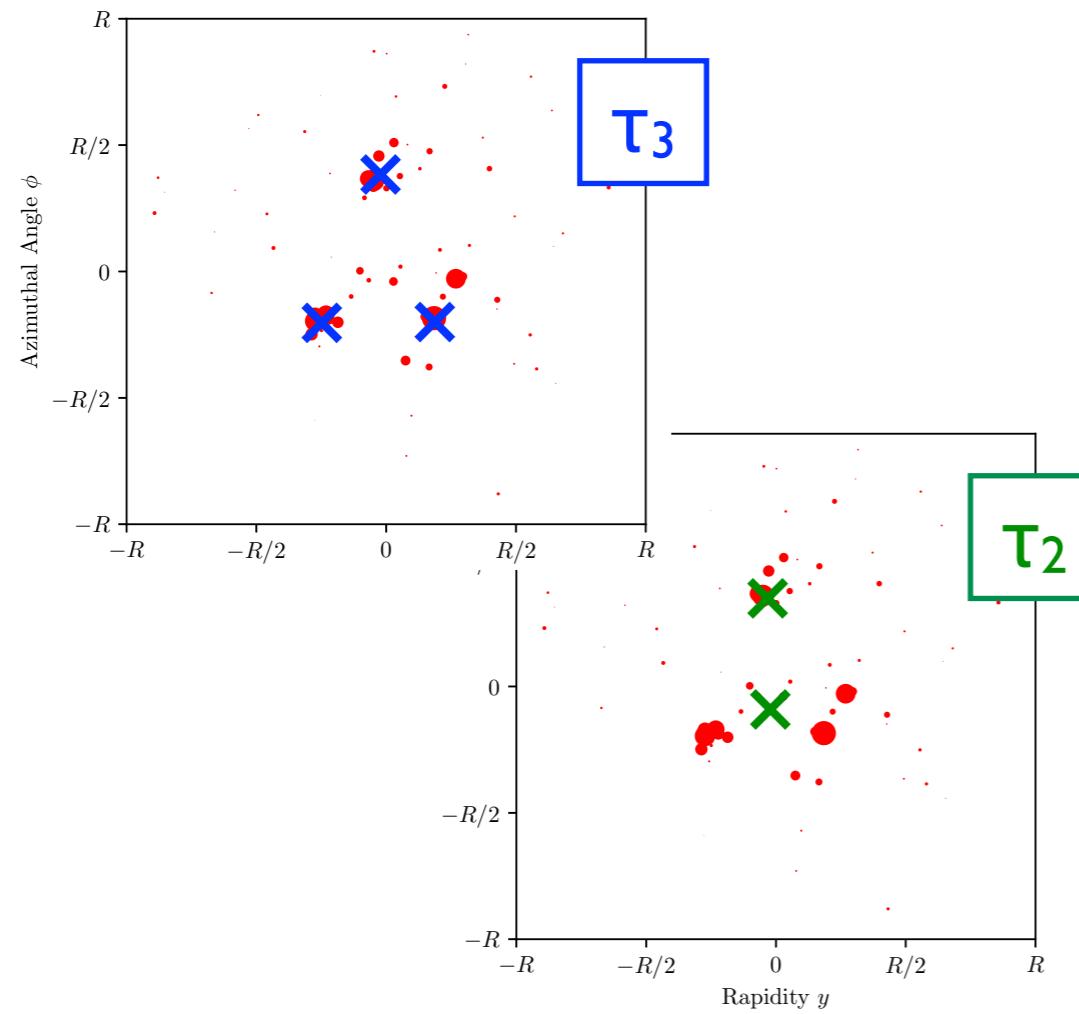
[JDT, Van Tilburg, [JHEP 2011](#), [JHEP 2012](#);  
based on Brandt, Dahmen, [ZPC 1979](#); Stewart, Tackmann, Waalewijn, [PRL 2010](#)]



# N-subjettiness = Point to Manifold EMD

*...is secretly an optimal transport problem*

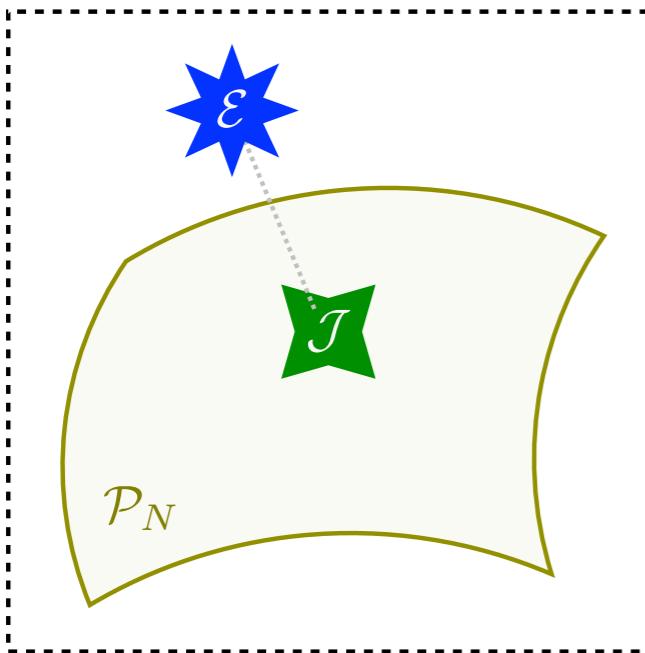
$$\tau_N(\mathcal{J}) = \min_{\mathcal{J}' \in \mathcal{P}_N} \text{EMD}(\mathcal{J}, \mathcal{J}')$$



[JDT, Van Tilburg, JHEP 2011, JHEP 2012;  
rephrased in the language of Komiske, Metodiev, JDT, PRL 2019]



# More Fun with N-particle Manifolds



## N-jettiness

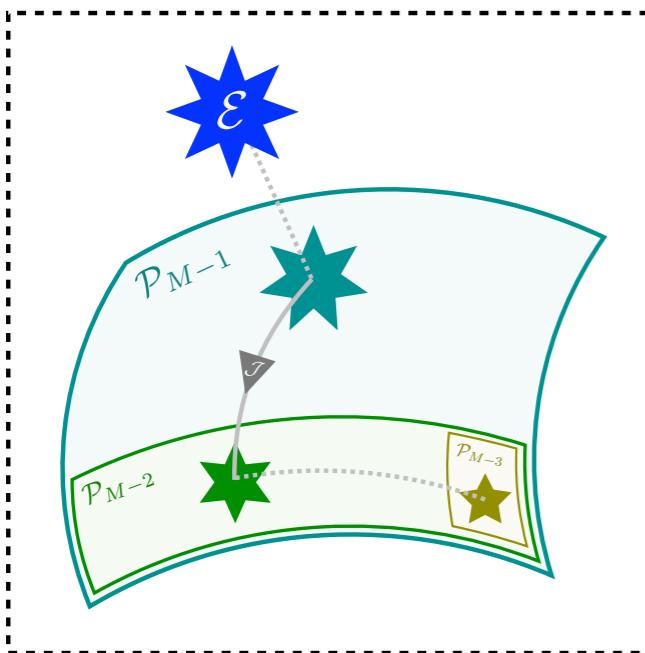
*Distance of closest approach to  $N$ -particle manifold*

[Brandt, Dahmen, [ZPC 1979](#); Stewart, Tackmann, Waalewijn, [PRL 2010](#)]

## Exclusive Cone Jet Finding

*Point of closest approach on  $N$ -particle manifold*

[Stewart, Tackmann, JDT, Vermilion, Wilkason, [JHEP 2015](#)]



## Sequential Jet Recombination

*Iteratively stepping between various  $N$ -particle manifolds*

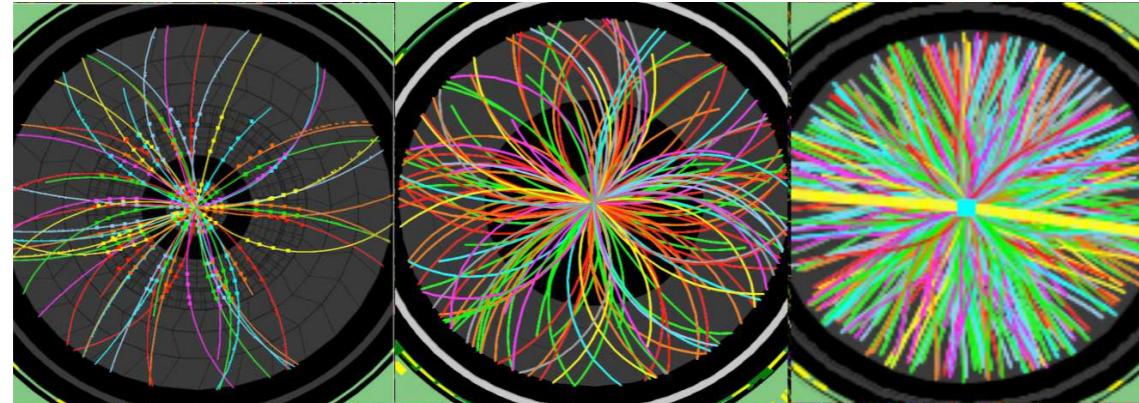
[Catani, Dokshitzer, Seymour, Webber, [NPB 1993](#); Ellis, Soper, [PRD 1993](#)]

[Dokshitzer, Leder, Moretti, Webber, [JHEP 1997](#); Wobisch, Wengler, [arXiv 1999](#)]

[Butterworth, Couchman, Cox, Waugh, [CPC 2003](#); Larkoski, Neill, JDT, [JHEP 2014](#)]

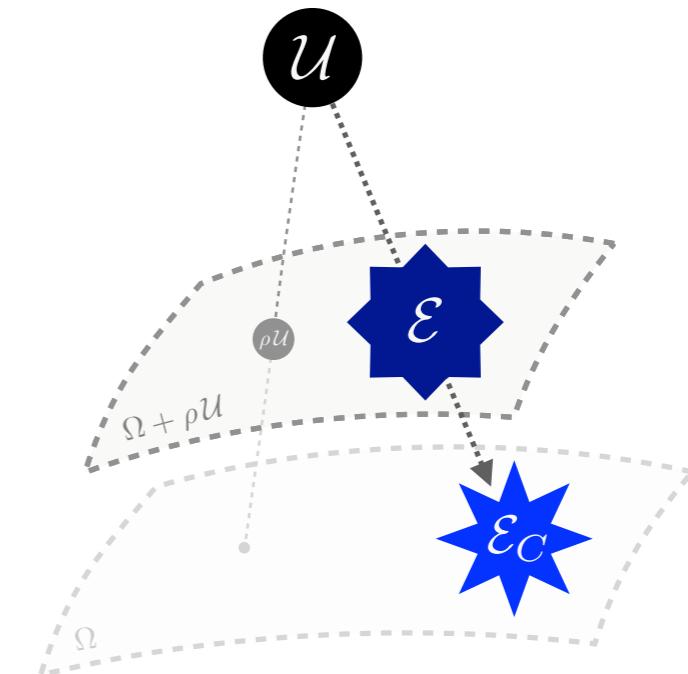
[Komiske, Metodiev, JDT, [JHEP 2020](#)]

# Pileup Mitigation



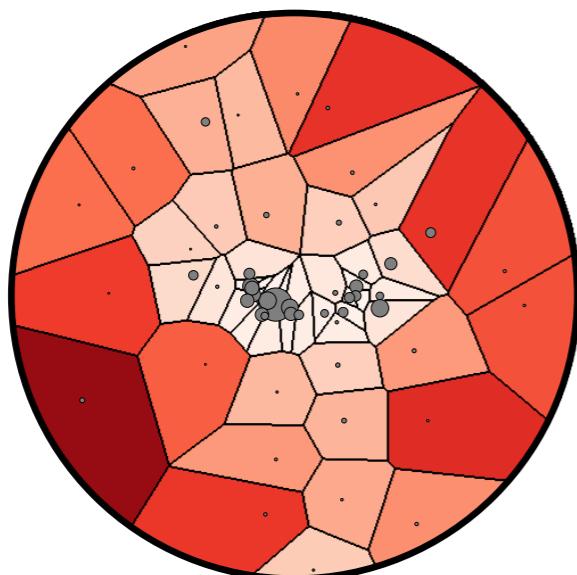
[see review in Soyez, PR 2019]

Uniform event contamination from overlapping proton-proton collisions



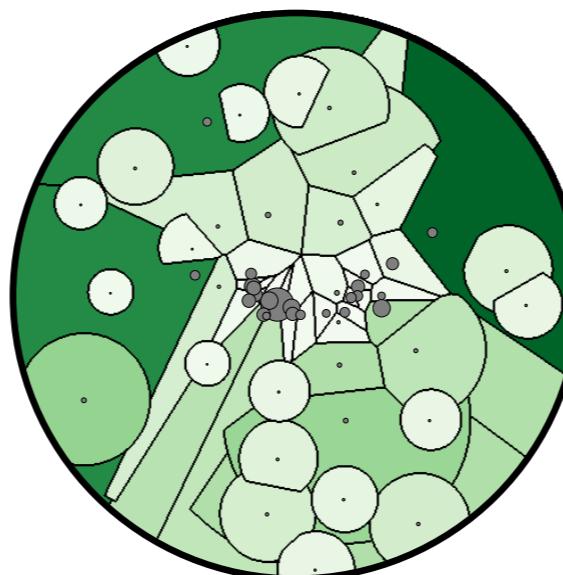
Pileup Mitigation:  
“Move away” from uniform event

Voronoi



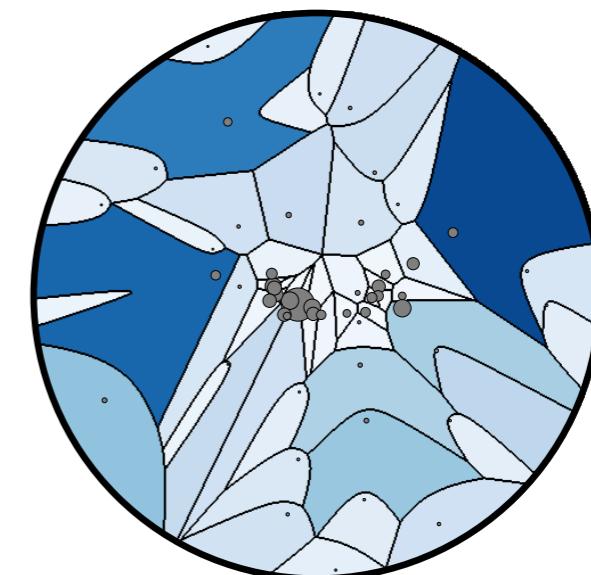
[Cacciari, Salam, Soyez, JHEP 2008]

Constituent Subtraction

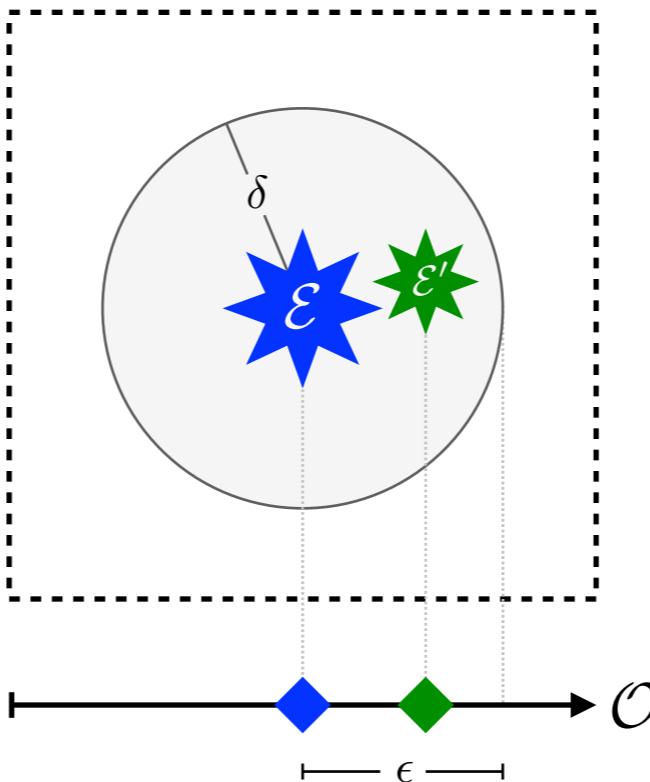


[Berta, Spousta, Miller, Leitner, JHEP 2014]

Apollonius



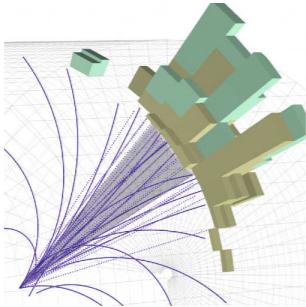
[Komiske, Metodiev, JDT, JHEP 2020]



We are just beginning to leverage the  
*conceptual richness* of optimal transport  
for high-energy physics application

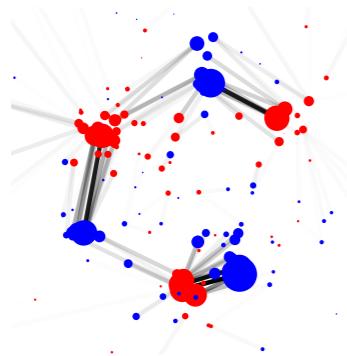
Ask me how far down this rabbit hole goes!

# Summary



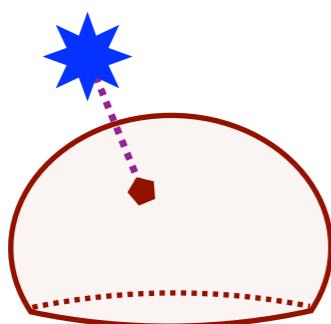
## Going with the (Energy) Flow

*Restricting our attention to IRC safe information  
is a theoretically motivated data analysis strategy*



## The Energy Mover's Distance

*Optimal transport allows us to triangulate the space  
of collider events and define an emergent geometry*



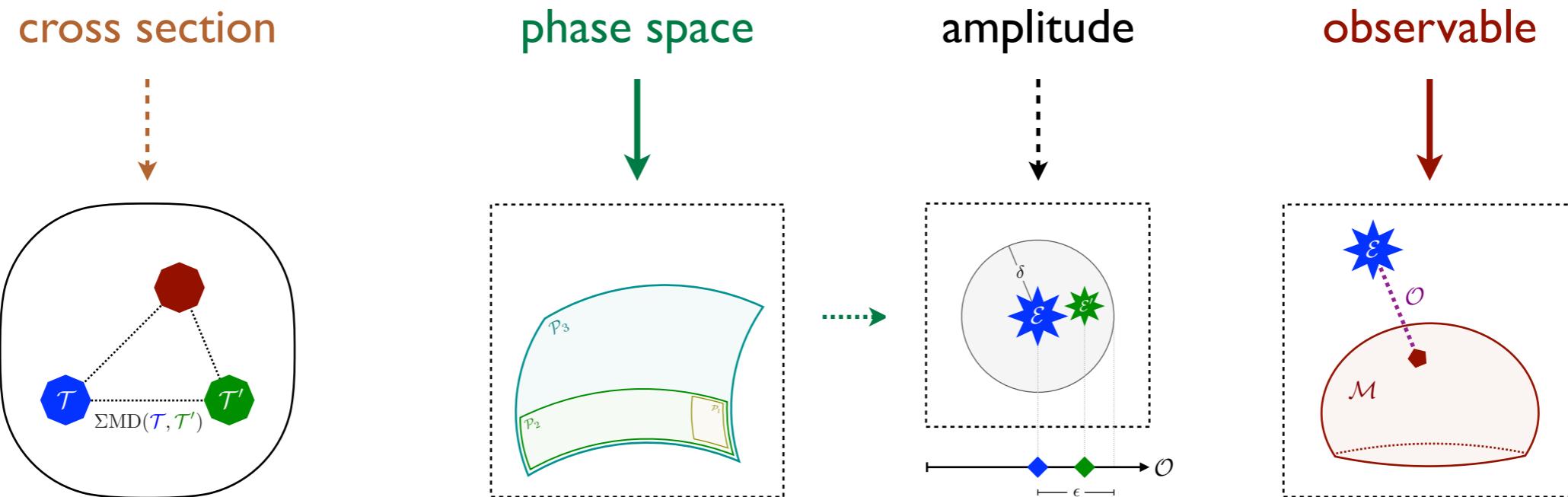
## Revealing a Hidden Geometry

*We can gain new perspectives on concepts/techniques  
in QFT and collider physics from the last half century*

*How far down does this rabbit hole go?*

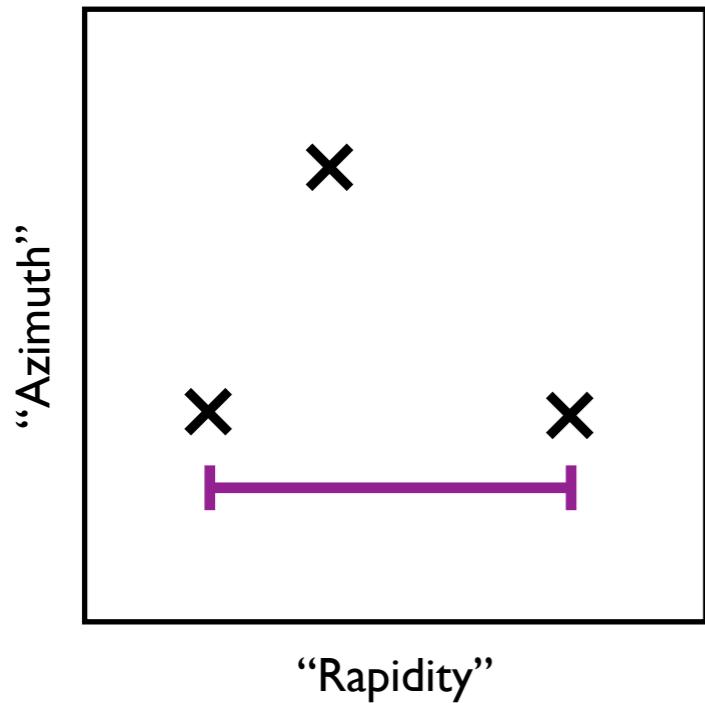
# Master Formula for Collider Physics

$$\sigma_{\text{obs}} \simeq \frac{1}{2E_{\text{CM}}^2} \sum_{n=2}^{\infty} \int d\Phi_n |\mathcal{M}_{AB \rightarrow 12\dots n}|^2 f_{\text{obs}}(\Phi_n)$$



[Komiske, Metodiev, JDT, [JHEP 2020](#)]

# Direction Space



**x** = Direction

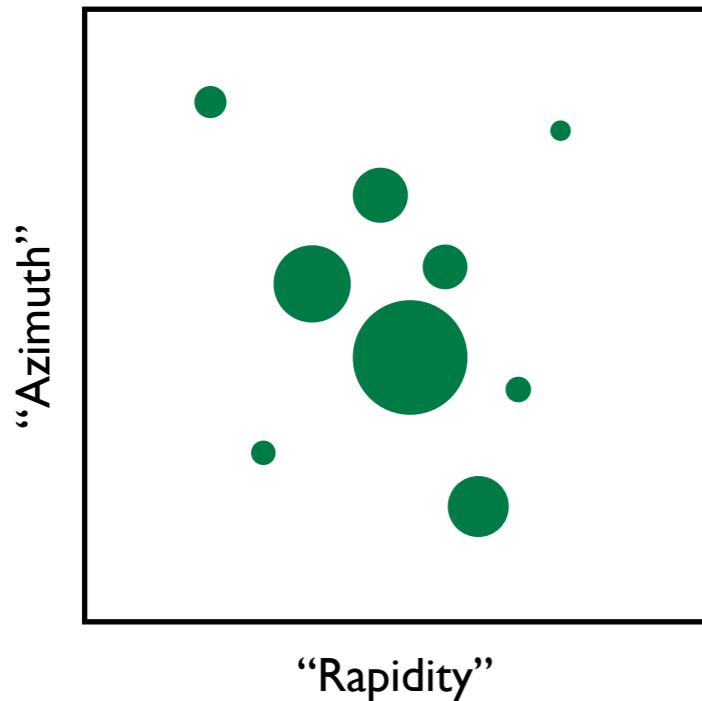
= Angular Distance

$$n_i^\mu = \frac{p_i^\mu}{E_i} = (1, \hat{n})^\mu$$

$$\theta_{ij} = \sqrt{2n_i^\mu n_{j\mu}}$$

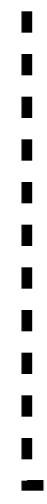
(for massless particles)

# Direction Space Distribution



● = Weighted Direction

— = Angular Distance



★ = Event

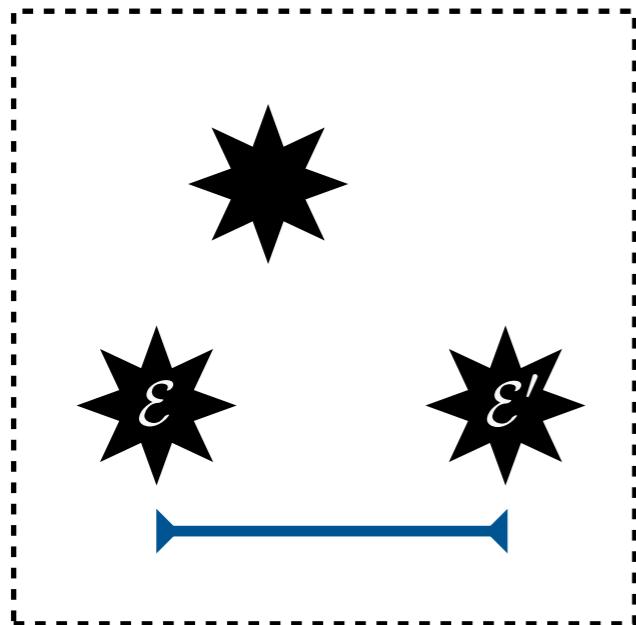
$$n_i^\mu = \frac{p_i^\mu}{E_i} = (1, \hat{n})^\mu$$

$$w_i = E_i$$

$$\theta_{ij} = \sqrt{2n_i^\mu n_{j\mu}}$$

(for massless particles)

# Event Space



★ = Event  
↔ = EMD  
Energy Mover's Distance

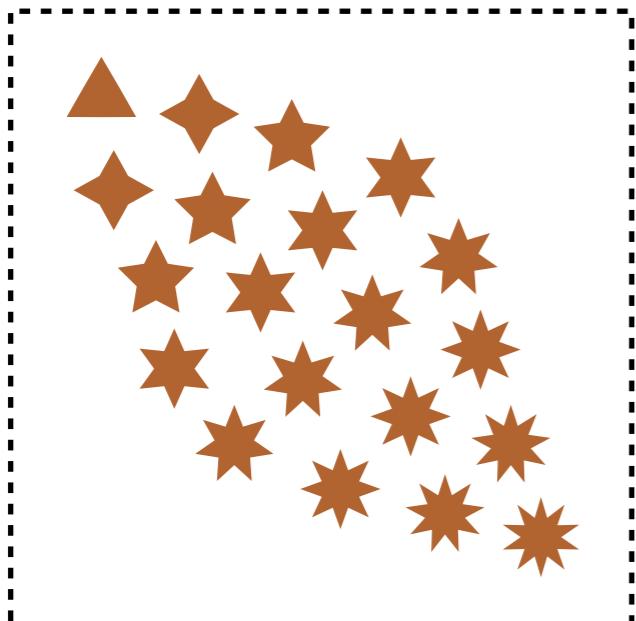
$$\mathcal{E}(\hat{n}) = \sum_i E_i \delta(\hat{n} - \hat{n}_i)$$

$$\text{EMD}(\mathcal{E}, \mathcal{E}') = \min_{\{f\}} \sum_i \sum_j f_{ij} \theta_{ij}$$

(for equal total energy)

[Komiske, Metodiev, JDT, PRL 2019]

# Event Space Distribution



= Weighted Event

$$\mathcal{E}(\hat{n}) = \sum_i E_i \delta(\hat{n} - \hat{n}_i)$$

$$w_a = \sigma_a$$

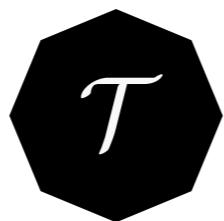
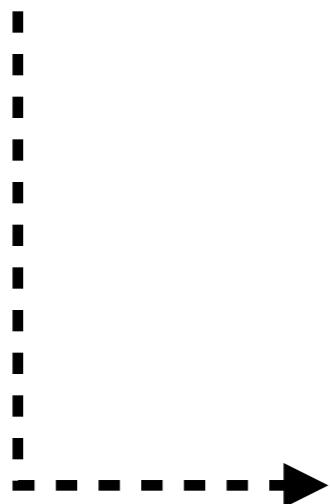


= EMD

Energy  
Mover's Distance

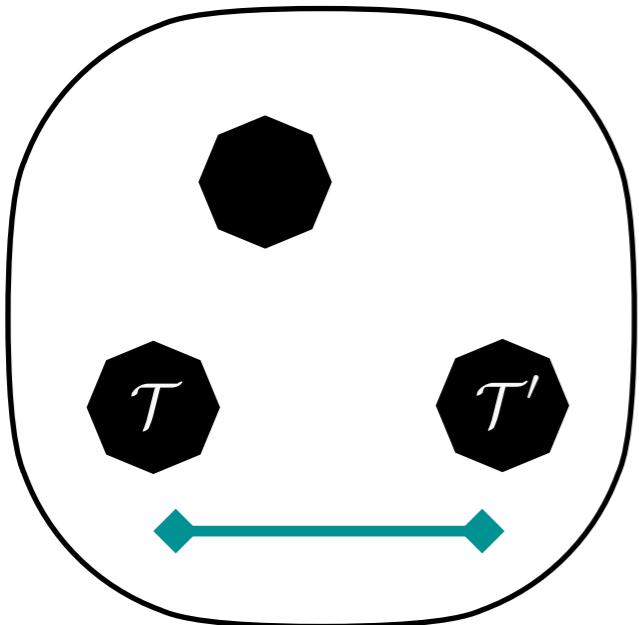
$$\text{EMD}(\mathcal{E}, \mathcal{E}') = \min_{\{f\}} \sum_i \sum_j f_{ij} \theta_{ij}$$

(for equal total energy)



= Theory

# Theory Space



● = Theory

↔ =  $\Sigma\text{MD}$   
Cross-Section  
Mover's Distance

$$\mathcal{T}(\mathcal{E}) = \sum_a \sigma_a \delta(\mathcal{E} - \mathcal{E}_a)$$

$$\Sigma\text{MD}(\mathcal{T}, \mathcal{T}') = \min_{\{\mathcal{F}\}} \sum_a \sum_b \mathcal{F}_{ab} \text{EMD}(\mathcal{E}_a, \mathcal{E}'_b)$$

(for equal total xsec)

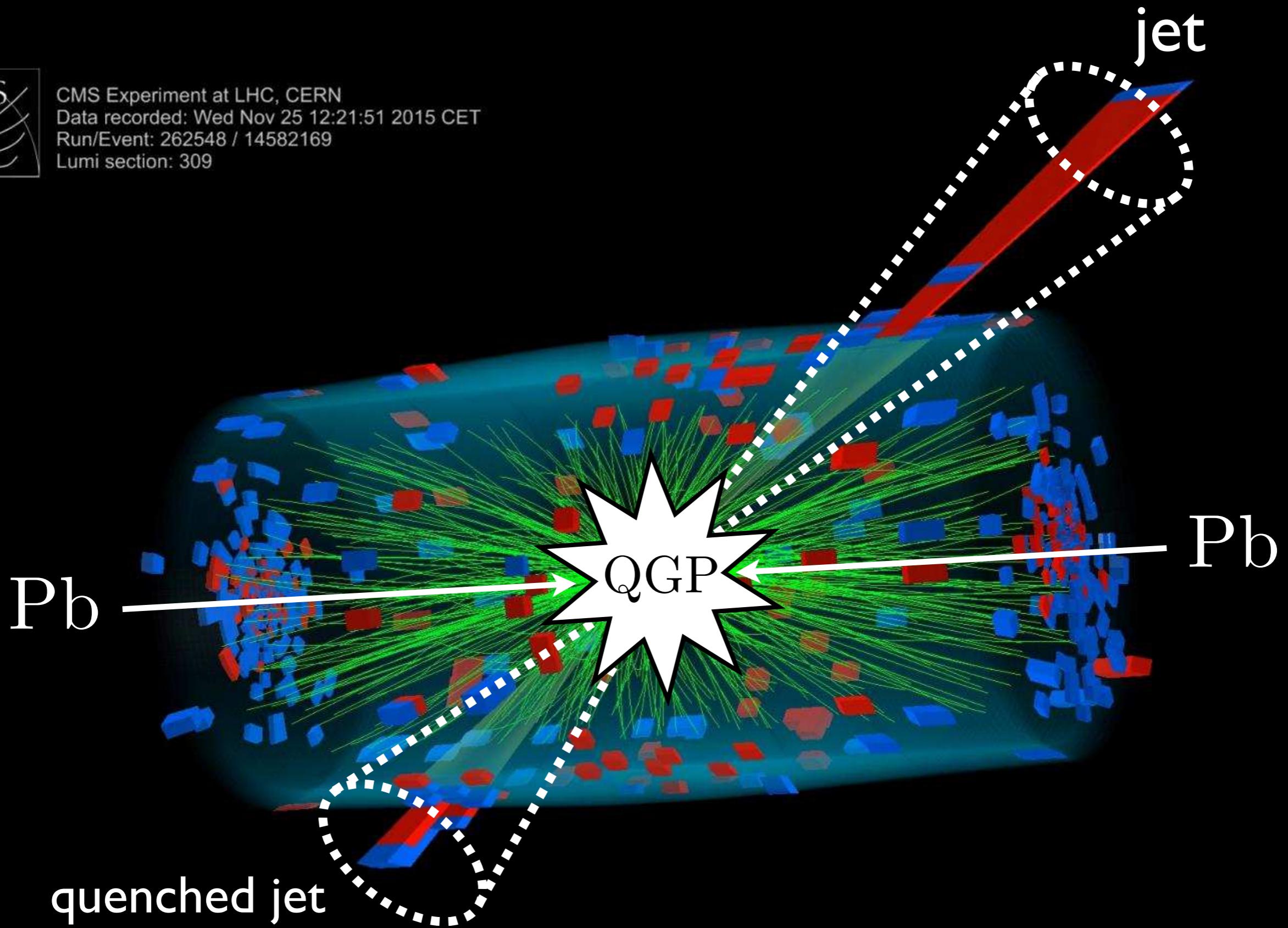
*A distance between theories!*

(e.g. EMD : N-jettiness ::  $\Sigma\text{MD}$  : k-eventiness)

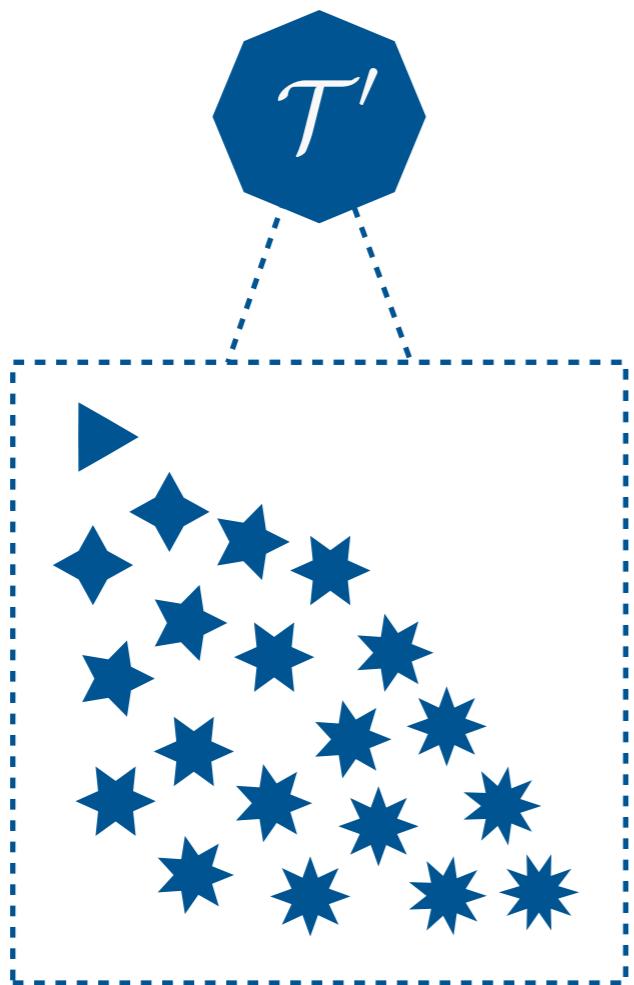
[Komiske, Metodiev, JDT, [JHEP 2020](#)]



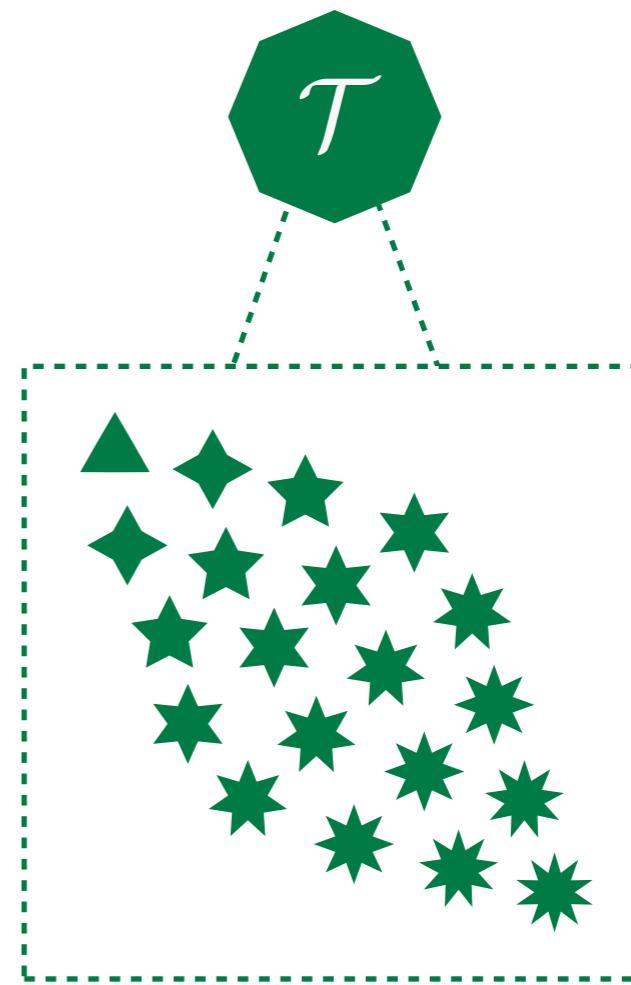
CMS Experiment at LHC, CERN  
Data recorded: Wed Nov 25 12:21:51 2015 CET  
Run/Event: 262548 / 14582169  
Lumi section: 309



## Theory Prime: In-Medium QCD

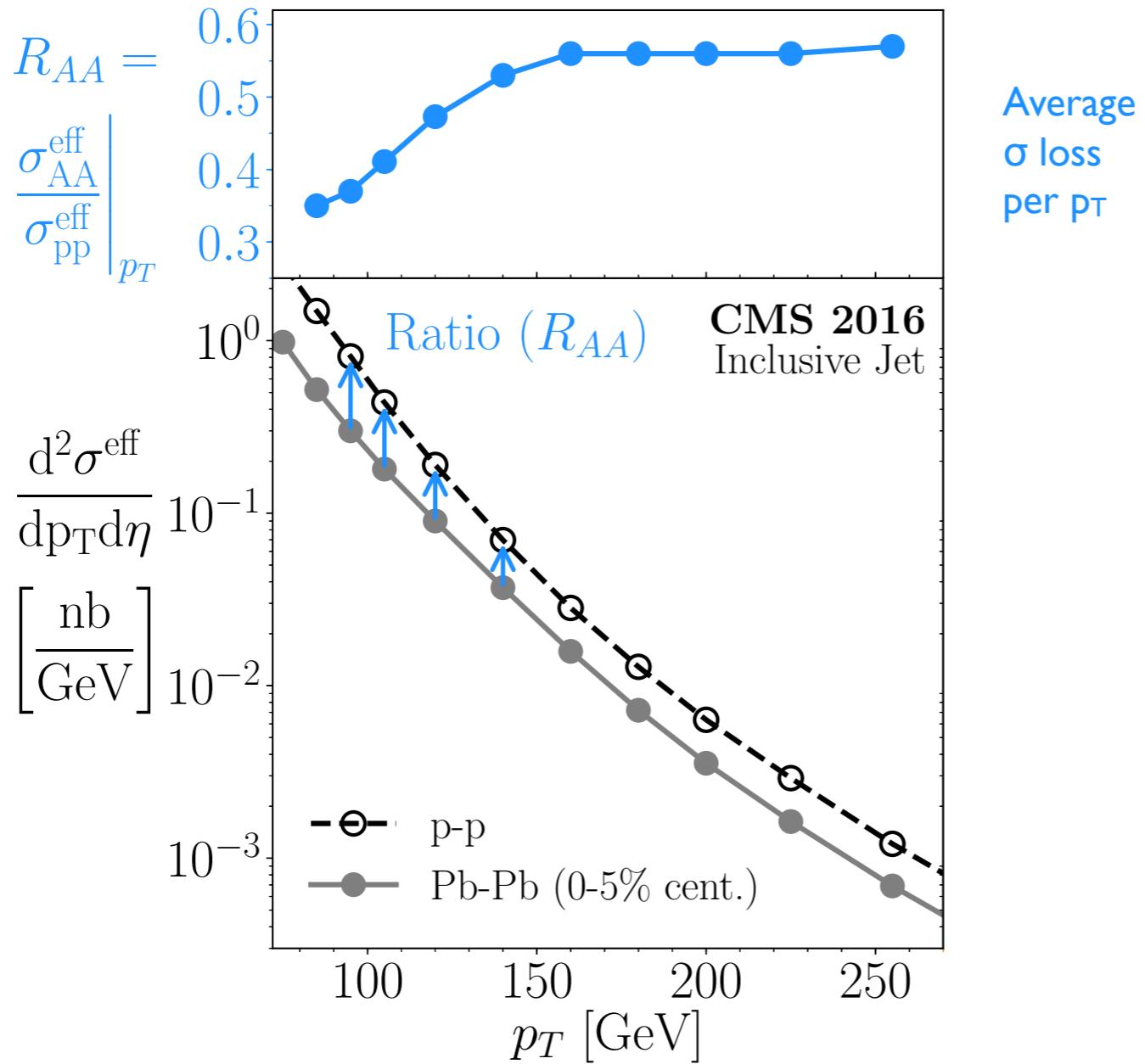


## Theory: Vacuum QCD



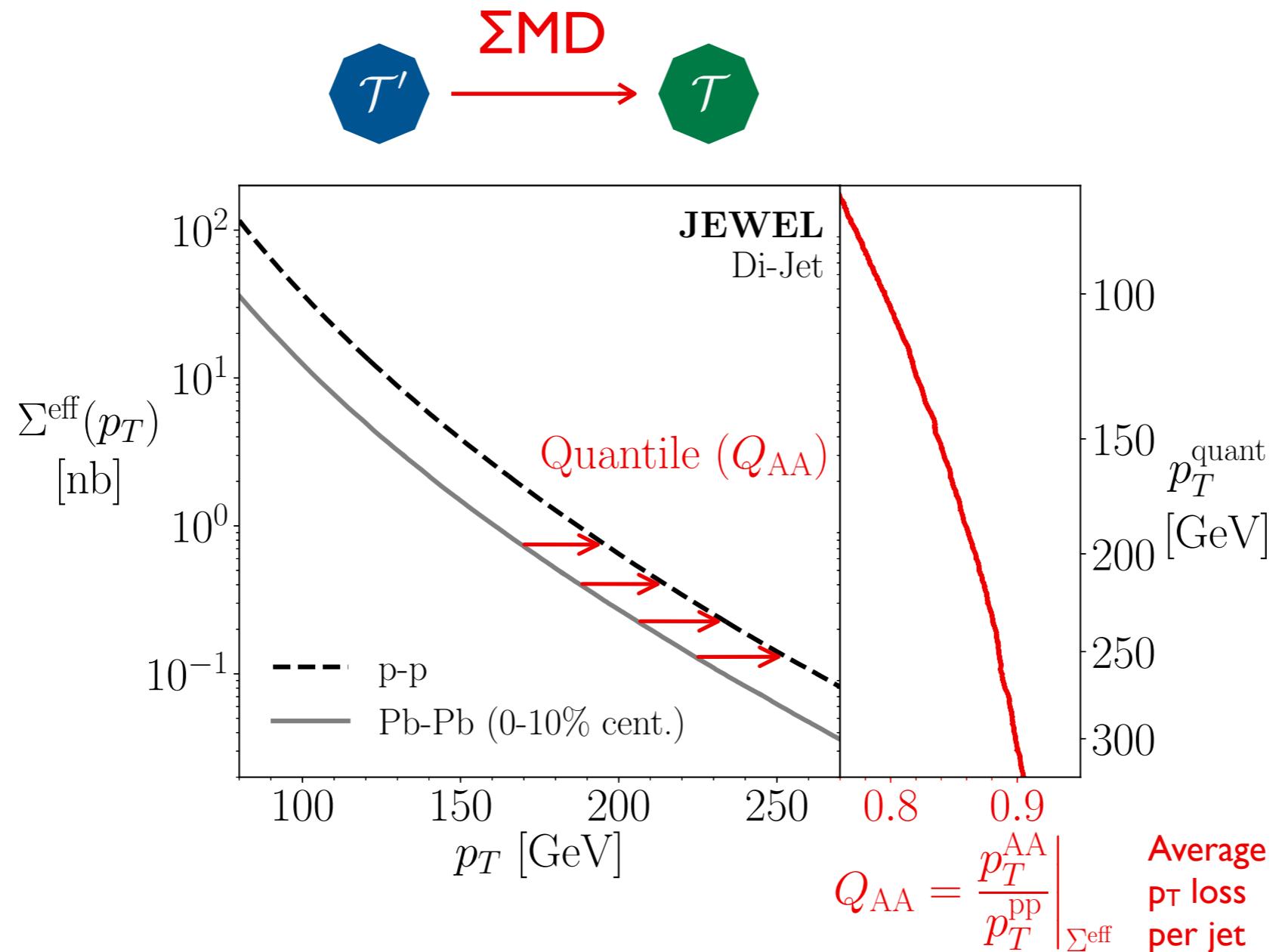
$\Sigma\text{MD}$   
 $\iff$

*Optimal transportation plan defines mapping  
between in-medium jets and vacuum jets!*



# Jet Quenching via Quantile Matching

*Equivalent to following a geodesic in theory space (!)*



[Brewer, Milhano, JDT, PRL 2019]



# *Backup Slides*

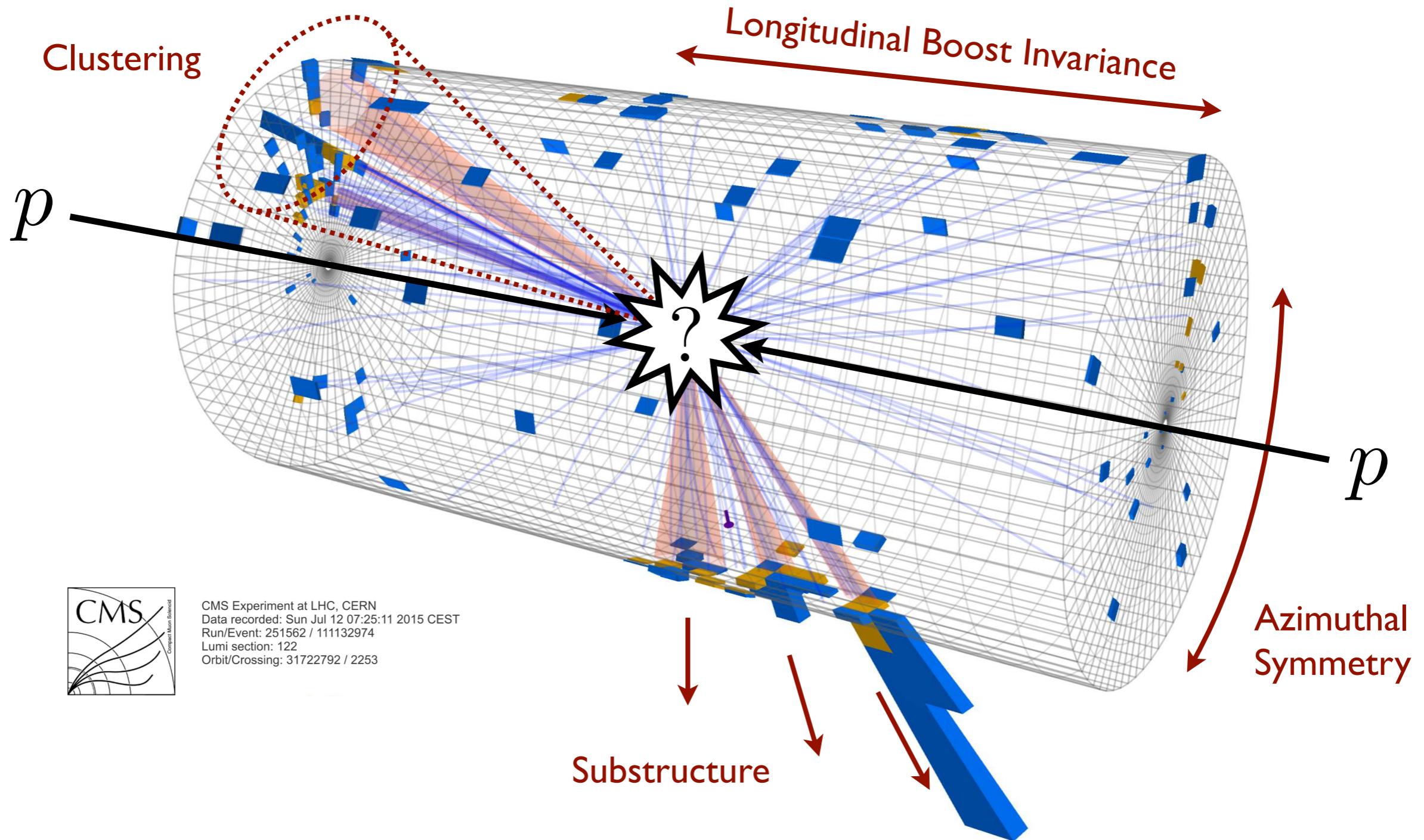
# AI<sup>2</sup>: Ab Initio Artificial Intelligence



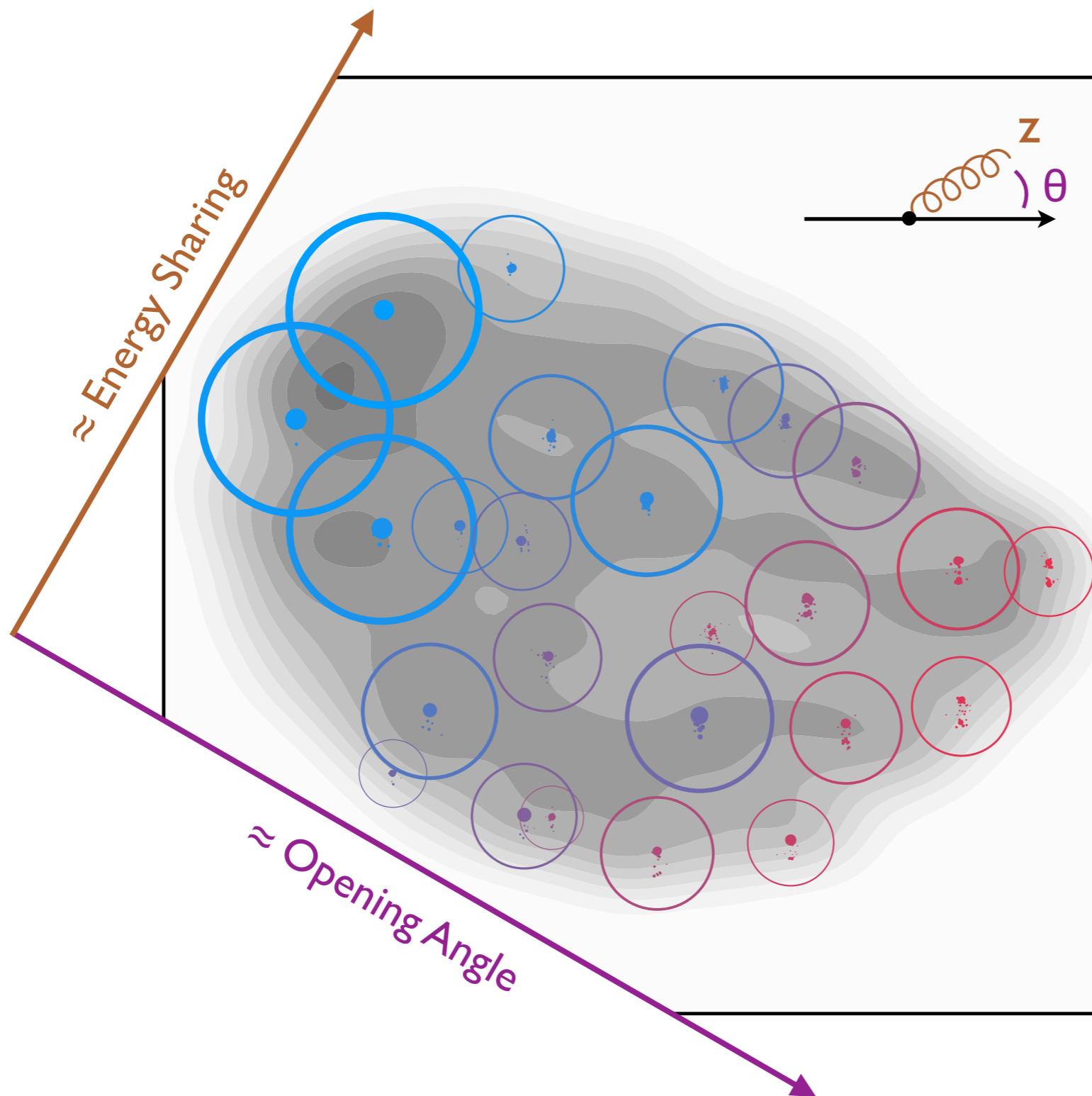
*Machine learning that incorporates  
first principles, best practices, and domain knowledge  
from fundamental physics*

*Symmetries, conservation laws, scaling relations, limiting behaviors, locality, causality,  
unitarity, gauge invariance, entropy, least action, factorization, unit tests,  
exactness, systematic uncertainties, reproducibility, verifiability, ...*

# The Manifest Geometry of One Collision



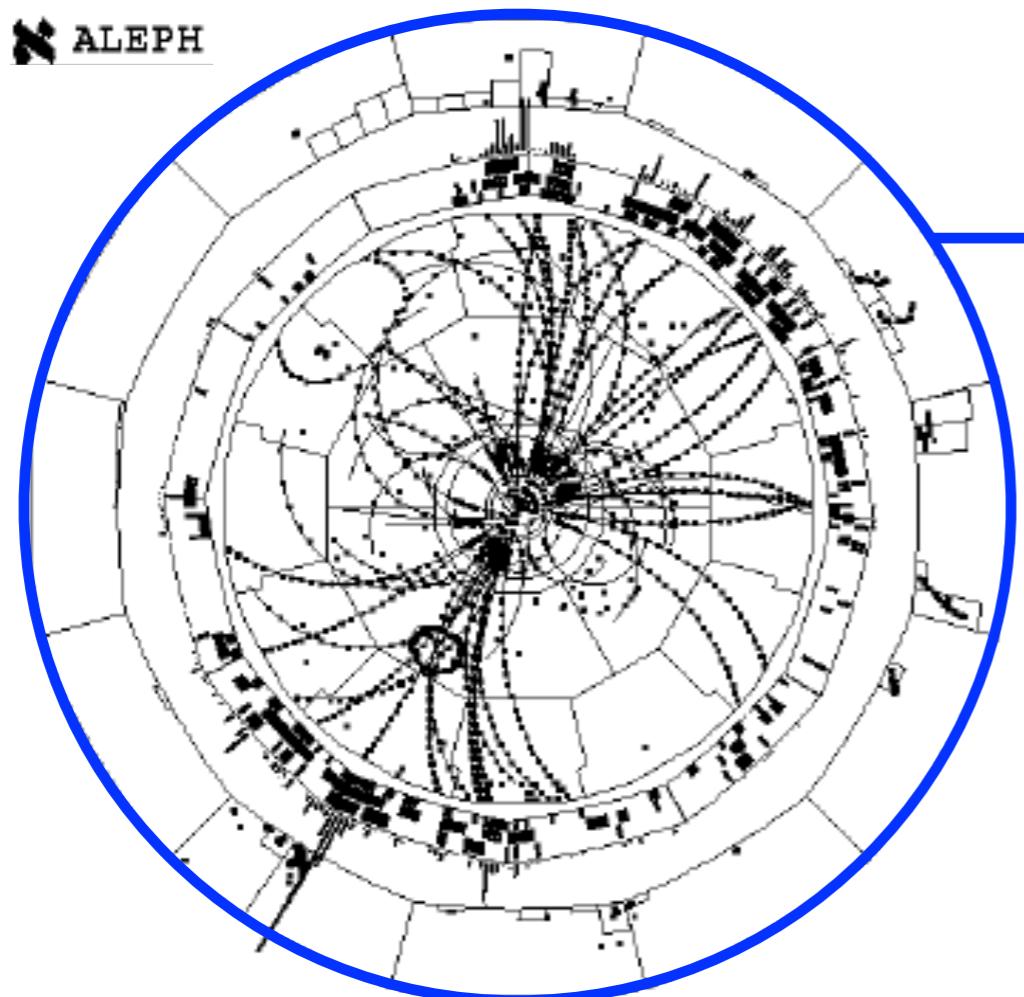
# The Emergent Geometry of Many Collisions



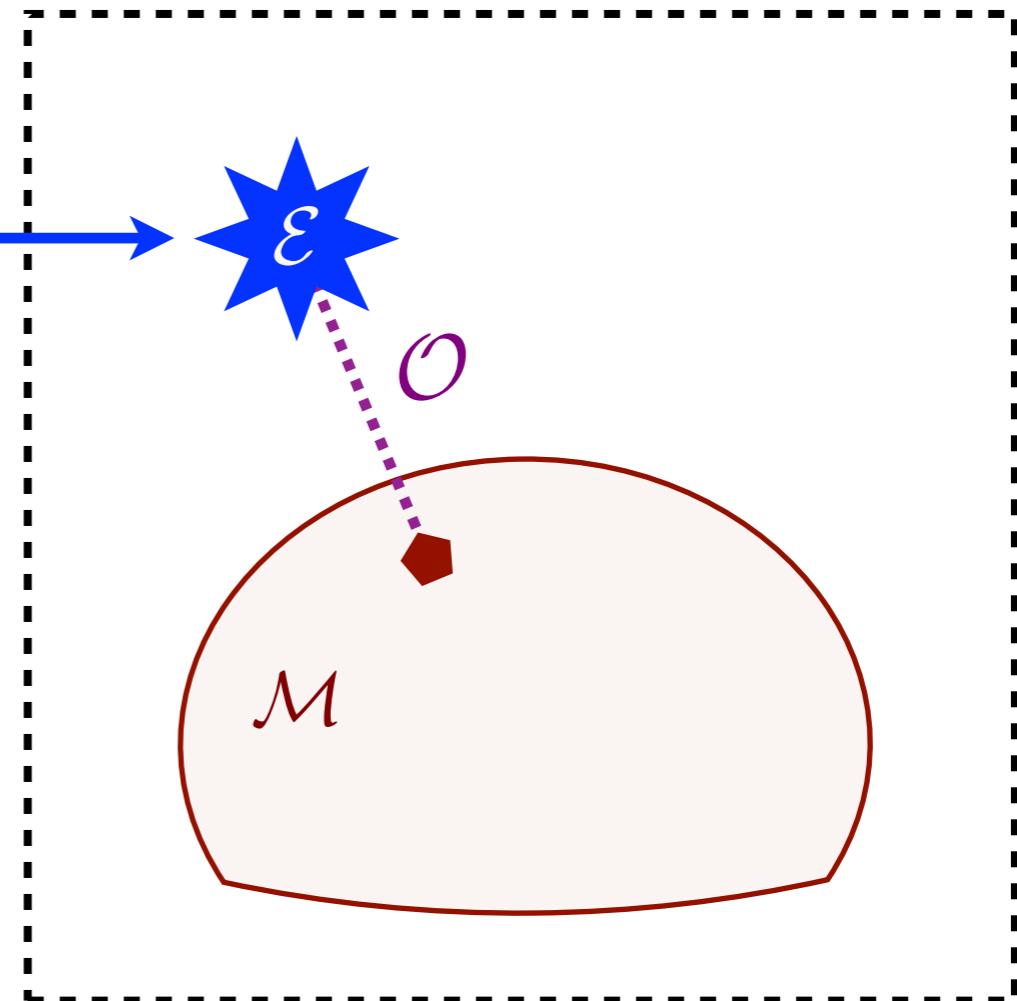
[Komiske, Mastandrea, Metodiev, Naik, JDT, [PRD 2020](#);  
based on Komiske, Metodiev, JDT, [PRL 2019](#); using [EnergyFlow](#) and [CMS Open Data](#)]

# The Hidden Geometry of Particle Collisions

E.g. Classic QCD Event Shapes



One Electron-Positron Event



Distance to a Manifold in Event Space

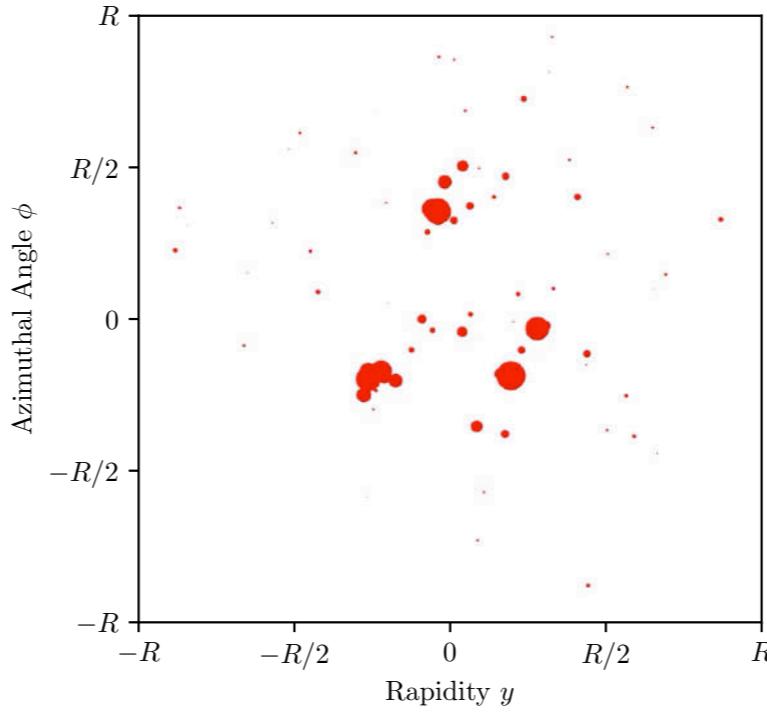
[Komiske, Metodiev, JDT, [JHEP 2020](#)]

[Brandt, Peyrou, Sosnowski, Wroblewski, [PL 1964](#); Farhi, [PRL 1977](#)]

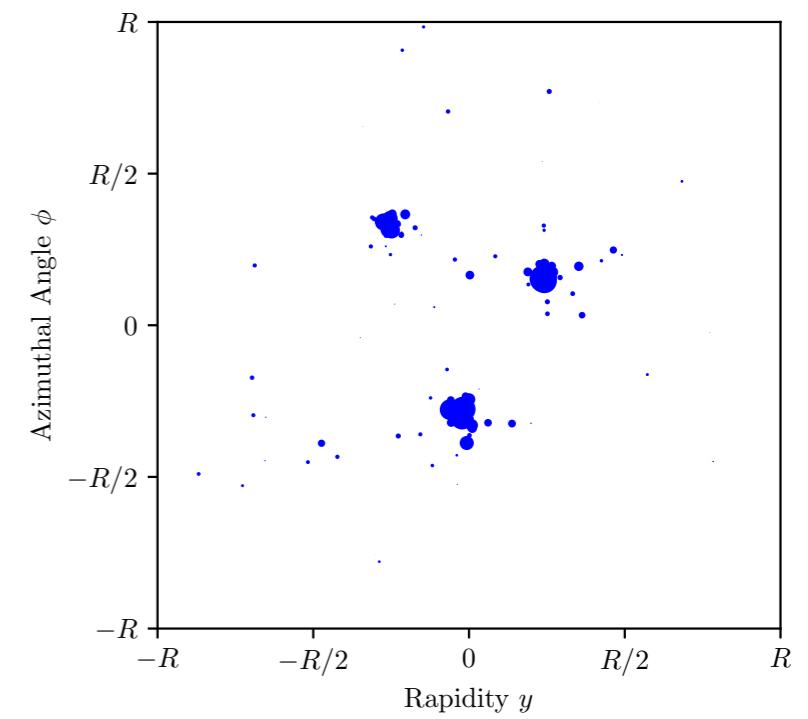
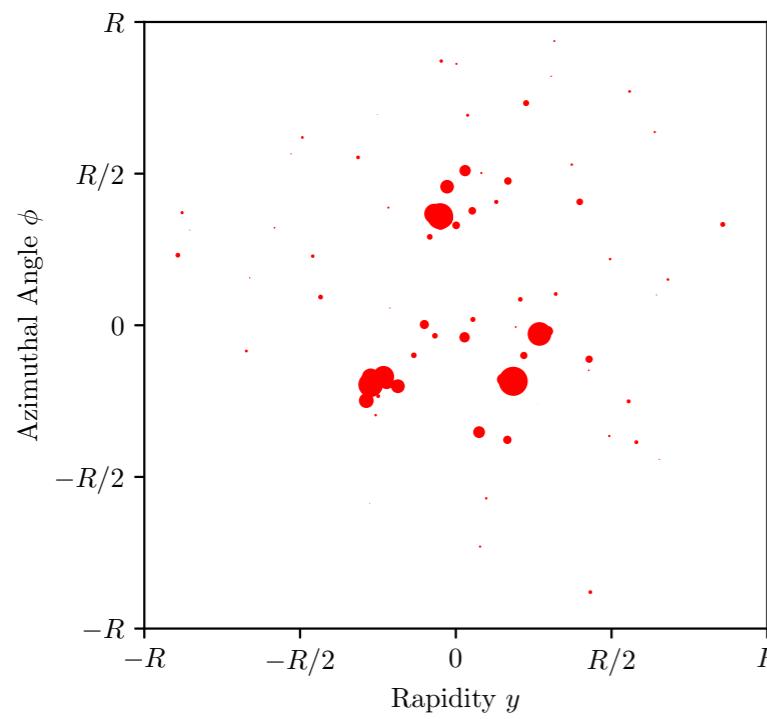


# Similarity of Two Energy Flows?

$$\mathcal{E}(\hat{n}) = \sum_i E_i \delta(\hat{n} - \hat{n}_i)$$

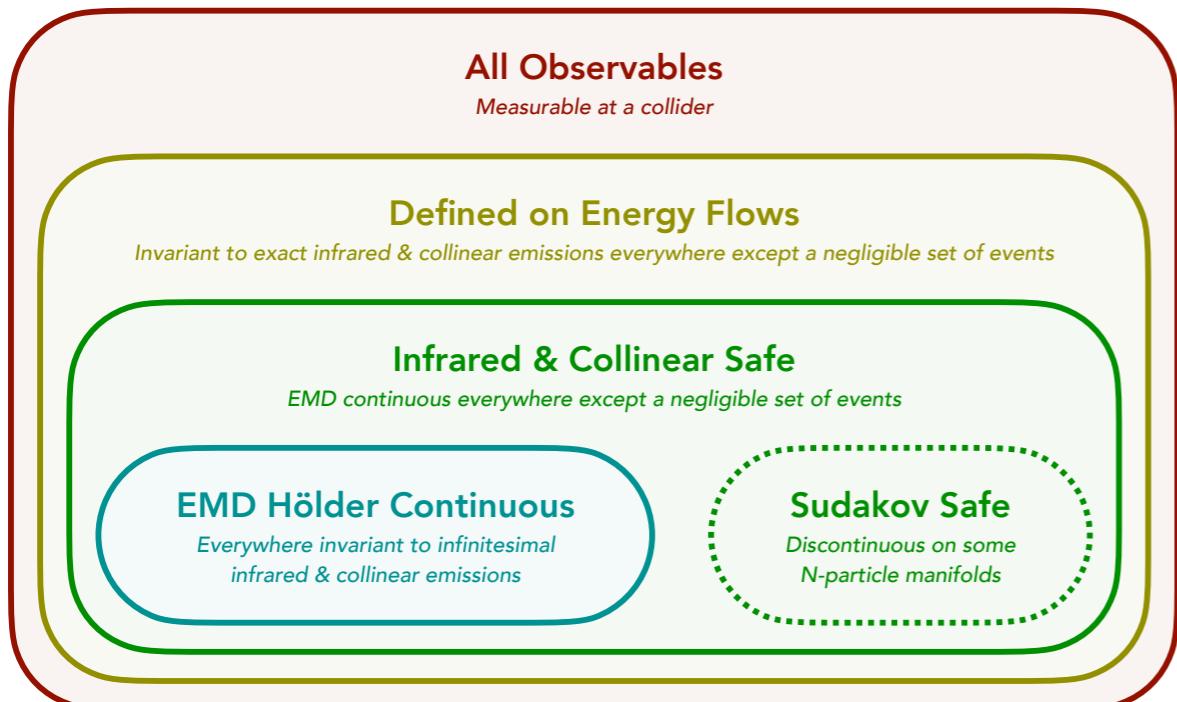


Optimal Transport:  
*Earth Mover's Distance*  
a.k.a.  $l$ -Wasserstein metric



[Komiske, Metodiev, JDT, PRL 2019; code at Komiske, Metodiev, JDT, [energyflow.network](#)]

# Observable Taxonomy



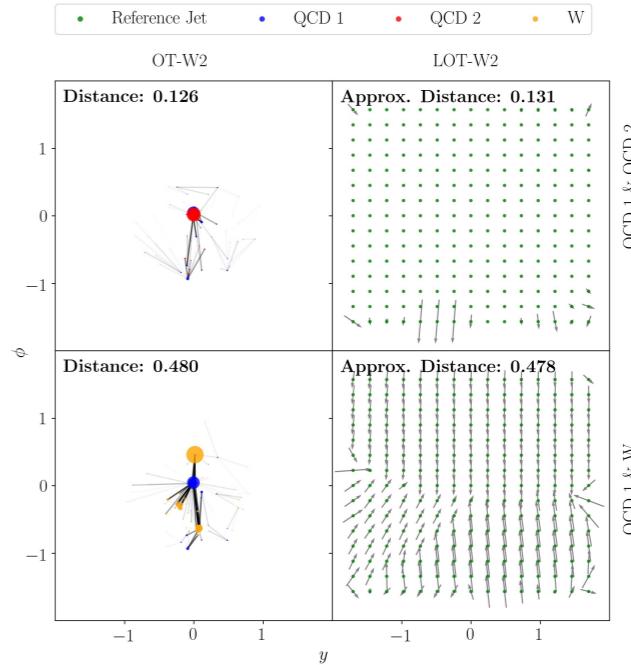
All Observables	Comments
Multiplicity ( $\sum_i 1$ )	IR unsafe and C unsafe
Momentum Dispersion [65] ( $\sum_i E_i^2$ )	IR safe but C unsafe
Sphericity Tensor [66] ( $\sum_i p_i^\mu p_i^\nu$ )	IR safe but C unsafe
Number of Non-Zero Calorimeter Deposits	C safe but IR unsafe
Defined on Energy Flows	
Pseudo-Multiplicity ( $\min\{N \mid \mathcal{T}_N = 0\}$ )	Robust to exact IR or C emissions
Infrared & Collinear Safe	
Jet Energy ( $\sum_i E_i$ )	Disc. at jet boundary
Heavy Jet Mass [67]	Disc. at hemisphere boundary
Soft-Dropped Jet Mass [38, 68]	Disc. at grooming threshold
Calorimeter Activity [69] ( $N_{95}$ )	Disc. at cell boundary
Sudakov Safe	
Groomed Momentum Fraction [39] ( $z_g$ )	Disc. on 1-particle manifold
Jet Angularity Ratios [37]	Disc. on 1-particle manifold
$N$ -subjettiness Ratios [47, 48] ( $\tau_{N+1}/\tau_N$ )	Disc. on $N$ -particle manifold
$V$ parameter [36] (Eq. (2.11))	Hölder disc. on 3-particle manifold
EMD Hölder Continuous Everywhere	
Thrust [40, 41]	
Spherocity [42]	
Angularities [70]	
$N$ -jettiness [44] ( $\mathcal{T}_N$ )	
$C$ parameter [71–74]	Resummation beneficial at $C = \frac{3}{4}$
Linear Sphericity [72] ( $\sum_i E_i n_i^\mu n_i^\nu$ )	
Energy Correlators [36, 75–77]	
Energy Flow Polynomials [15, 17]	

[Komiske, Metodiev, JDT, [JHEP 2020](#); cf. Sterman, [PRD 1979](#); Banfi, Salam, Zanderighi, [JHEP 2005](#); Larkoski, Marzani, JDT, [PRD 2015](#)]

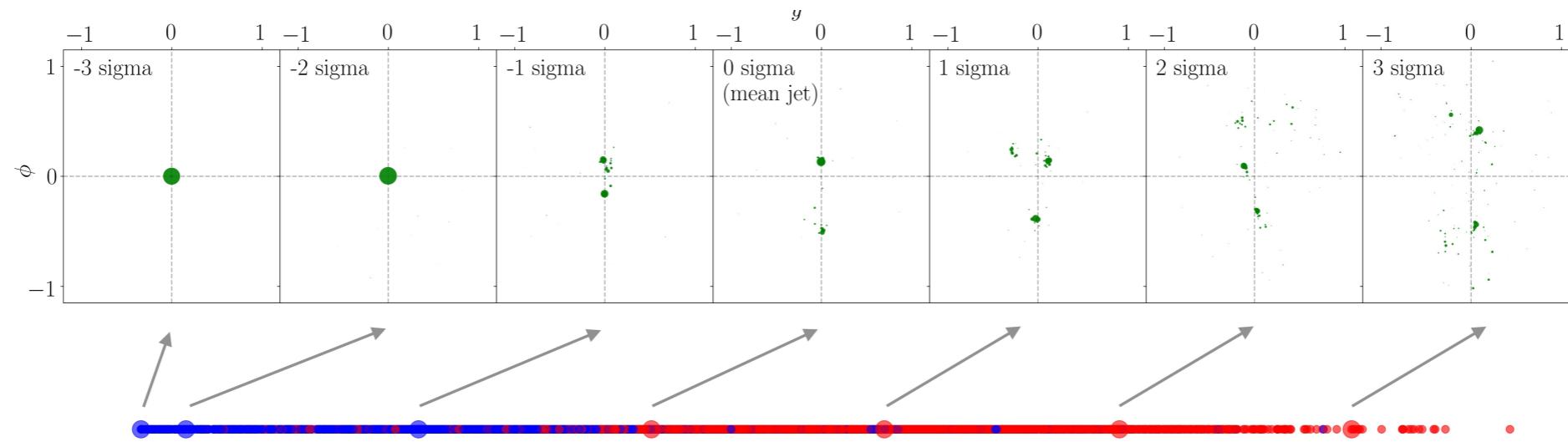
# Linearized Optimal Transport

With the help of a reference event, transportation distances\* can be efficiently mapped to Euclidean distances

\* assuming the 2-Wasserstein measure



Enables coordinate-based techniques like Linear Discriminate Analysis



[Cai, Cheng, Craig, Craig, PRD 2020]