# A Closed-Form, Differentiable Metric on Events

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Metric on the Space of Collider Events



Natural Questions:

What is the dimensionality of this space?

What is its topology?

What is its volume?

Where are different events located?

Metric on the Space of Collider Events



Primitive Question:

What is the distance between events  $\mathcal{E}$  and  $\mathcal{E}'$ ?

Reminder: Three Properties of a Metic

"Identity of Indiscernibles"

$$d(\mathcal{E},\mathcal{E}) = 0$$

Shortest distance from an event to itself is 0

"Symmetry" (and non-negativity)

$$d(\mathcal{E}, \mathcal{E}') = d(\mathcal{E}', \mathcal{E}) \ge 0$$



Distance from one event to another equals opposite

"Triangle Inequality"

 $d(\mathcal{E}, \mathcal{E}') \le d(\mathcal{E}, \mathcal{E}'') + d(\mathcal{E}', \mathcal{E}'')$ 

$$\mathcal{E}''$$

The most direct route is the shortest

### Events as Point-Clouds on Detector/Celestial Sphere

$$\operatorname{EMD}(\mathcal{E}, \mathcal{E}') = \min_{\{f_{ij}\}} \sum_{ij} f_{ij} \frac{\theta_{ij}}{R} + \left| \sum_{i} E_i - \sum_{j} E'_j \right|$$
$$f_{ij} \ge 0, \quad \sum_{j} f_{ij} \le E_i, \quad \sum_{i} f_{ij} \le E'_j, \quad \sum_{ij} f_{ij} = E_{\min},$$

Metric is minimal transportation cost to rearrange one event into another

Monge's "Earth Mover's Distance" applied to colliders Monge, 1781

Cédric Villani won Fields Medal in 2010 for Optimal Transport

C.Villani, "Optimal Transport: Old and New" (2009)



## Three Issues with EMD and other metrics

Must Minimize Over All Labelings to Ensure Particle Permutation Invariance



No Closed-Form Expression for Two-Dimensional Optimal Transport



# **POT: Python Optimal Transport**

Non-Zero Distance for Events Related by Isometries





### New Metric Proposal

Particle Permutation Invariance Ensured by Not Labeling Particles



Represent Event as a One-Dimensional Distribution to Use Exact Transport Results

$$d_p(\mathcal{E}, \mathcal{E}') = \int dE^2 \left| S_{\mathcal{E}}^{-1}(E^2) - S_{\mathcal{E}'}^{-1}(E^2) \right|^p$$

Only Use Relative Angles to Ensure Invariance to Isometries





Introducing The Spectral EMD

Fundamental Object: "Spectral Function"

name: hep-ph/9601308 first use: Basham, Brown, Ellis, Love 1977

$$s_{\mathcal{E}}(\omega) = \sum_{\substack{i,j \in \mathcal{E} \\ (including \ i = j)}} E_i E_j \, \delta(\omega - \omega_{ij})$$
angle between particles *i* and *j* weighted by product of energies

Previously Used in Jet Substructure for Identification of Dominant Structure



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$$s_{\mathcal{E}}(\omega) = \sum_{\substack{i,j \in \mathcal{E} \\ (\text{including } i = j)}} E_i E_j \, \delta(\omega - \omega_{ij})$$
sum runs over all *i,j* angle between particles *i* and *j* weighted by product of energies

Distribution of Particles on Celestial Sphere can be Reconstructed (with probability 1, up to isometries)



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#### **One-Dimensional Optimal Transport**



**One-Dimensional Optimal Transport** 

"p-Wasserstein" Metric

$$d_{p}(\mathcal{E}, \mathcal{E}') = \int dE^{2} \left| S_{\mathcal{E}}^{-1}(E^{2}) - S_{\mathcal{E}'}^{-1}(E^{2}) \right|^{p}$$



**One-Dimensional Optimal Transport** 

"p-Wasserstein" Metric

$$d_{p}(\mathcal{E}, \mathcal{E}') = \int dE^{2} \left| S_{\mathcal{E}}^{-1}(E^{2}) - S_{\mathcal{E}'}^{-1}(E^{2}) \right|^{p}$$

$$Inverse Cumulative Spectral Function in Gory Detail:
$$S^{-1}(E^{2}) = \sum_{\substack{n \in \mathcal{E}^{2} \\ \omega_{n} < \omega_{n+1}}} \omega_{n} \Theta \left( \sum_{i \in \mathcal{E}} E_{i}^{2} + \sum_{\substack{m \leq n \in \mathcal{E}^{2}}} (2EE)_{m} - E^{2} \right) \Theta \left( E^{2} - \sum_{i \in \mathcal{E}} E_{i}^{2} - \sum_{m < n \in \mathcal{E}^{2}} (2EE)_{m} \right)$$$$

$$\sigma \sim \int_{\text{Explicitly}} \frac{d^2 p_i}{d^2 m} \frac{2\pi\delta(p_i^2 - m_i^2)}{d^2 m} \int_{\text{tail}} \frac{\delta^{(4)}}{\delta^{(4)}} \left( \begin{array}{c} Q - \sum_{i=1}^{M} p_i \end{array} \right) |\mathcal{M}|^2$$

Overcomplete event information: order-N<sup>2</sup> pairwise angles, but only need order-N to triangulate the plane (Euler characteristic)



#### Closed-Form Expression for 2-Wasserstein Spectral Metric

$$d_{2}(\mathcal{E}_{1}, \mathcal{E}_{2}) = \sum_{i < j \in \mathcal{E}_{1}} 2E_{i}E_{j}\omega_{ij}^{2} + \sum_{i < j \in \mathcal{E}_{2}} 2E_{i}E_{j}\omega_{ij}^{2}$$
$$- 2\sum_{\substack{n \in \mathcal{E}_{1}^{2}, l \in \mathcal{E}_{2}^{2} \\ \omega_{n} < \omega_{n+1} \\ \omega_{l} < \omega_{l+1}}} \omega_{n}\omega_{l} \left( \min\left[\sum_{n \le m \in \mathcal{E}_{1}^{2}} (2EE)_{m}, \sum_{l \le k \in \mathcal{E}_{2}^{2}} (2EE)_{k}\right] - \max\left[\sum_{n < m \in \mathcal{E}_{1}^{2}} (2EE)_{m}, \sum_{l < k \in \mathcal{E}_{2}^{2}} (2EE)_{k}\right] \right)$$
$$\times \Theta\left(\min\left[\sum_{n \le m \in \mathcal{E}_{1}^{2}} (2EE)_{m}, \sum_{l \le k \in \mathcal{E}_{2}^{2}} (2EE)_{k}\right] - \max\left[\sum_{n < m \in \mathcal{E}_{1}^{2}} (2EE)_{m}, \sum_{l < k \in \mathcal{E}_{2}^{2}} (2EE)_{k}\right] \right)$$

Computational cost is due to ordering ~N<sup>2</sup> angles

Memory costs may be significant for high-multiplicities: ~40 kB to record 4-vectors of 1000 particles ~10 MB to record spectral function of 1000 particles

May be ways to compress spectral function information c.f., spectral function versus Fourier transform

## Comparison with Other Differentiable Metric Proposals Based on EMD



SHAPER: 2302.12266

Model complex shapes by large numbers of points and then use POT



NEEMo: 2209.15624 EMD CNN: 2306.04712

Construct a new neural network for geometric fitting of particles in event

#### Comparison with Other Differentiable Metric Proposals Based on EMD



Can easily analytically optimize for ring radius R

Summary

By exploiting the information in the spectral function, we can construct a metric on events that is invariant to isometries and expressed in closed form

Preliminary studies also indicate that the EMD is significantly more computationally intensive that the spectral metric (at least a factor of N slower)

Simple, differentiable, closed form expressions for the spectral metric enable first-principles theory calculations that are impossible with the EMD

Look forward to more applications and public code in the near future!

# Bonus



 $d_2(\mathcal{E}_1, \mathcal{E}_2) = 2E_1 E_3 \omega_{13}^2 + 2E_2 E_4 \omega_{24}^2 - 4\min[E_1 E_3, E_2 E_4] \omega_{13} \omega_{24}$ 



# $d_2(\mathcal{E}_1, \mathcal{E}_2) = 2E_1 E_3 \omega_{13}^2 + 2E_2 E_4 \omega_{24}^2 - 4\min[E_1 E_3, E_2 E_4] \omega_{13} \omega_{24}$

sum of jets' two-point energy correlation functions

Differentiable in all pairwise angles: extremization is easy

Non-analytic in energies: extremization is a discrete partitioning problem