

Stochastic Optimization for Large-scale Optimal Transport

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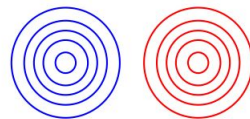
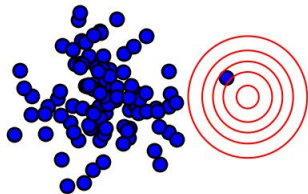
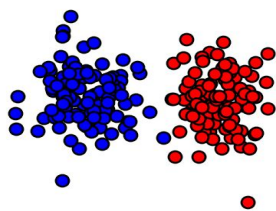
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MVA - Computational Optimal Transport
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Presentation of the problem

- How to compute OT in different settings ?

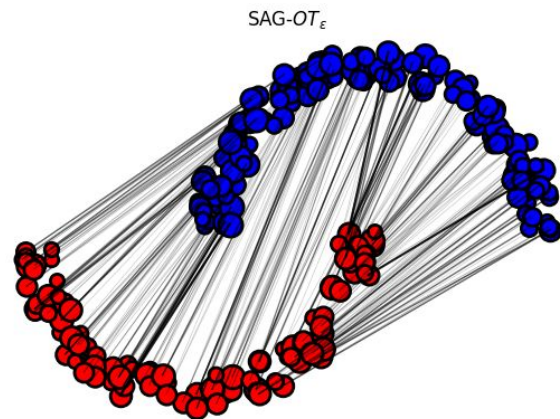
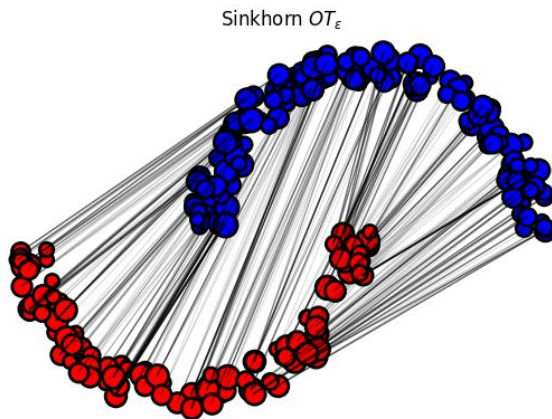
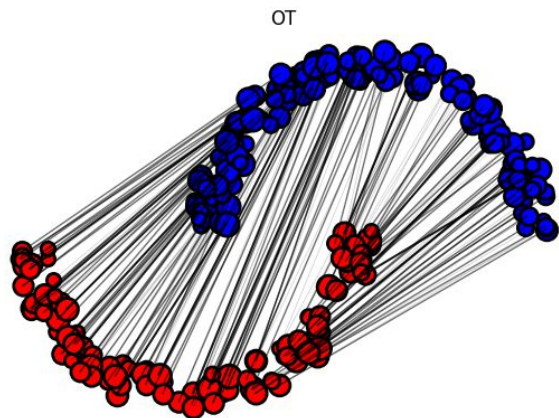


- Using stochastic algorithms !

$$\begin{aligned} OT_{\epsilon}(\alpha, \beta) &= \max_{f, g \in \mathcal{C}(\mathcal{X}) \times \mathcal{C}(\mathcal{Y})} \mathbb{E}_{X, Y} \left[f(x) + g(y) - \epsilon \exp \left(\frac{f(x) + g(y) - C(x, y)}{\epsilon} \right) \right] \\ &= \max_{f \in \mathcal{C}(\mathcal{X})} \mathbb{E}_Y \left[\int_{\mathcal{X}} f d\alpha + f^{c, \epsilon}(y) - \epsilon \right] \end{aligned}$$

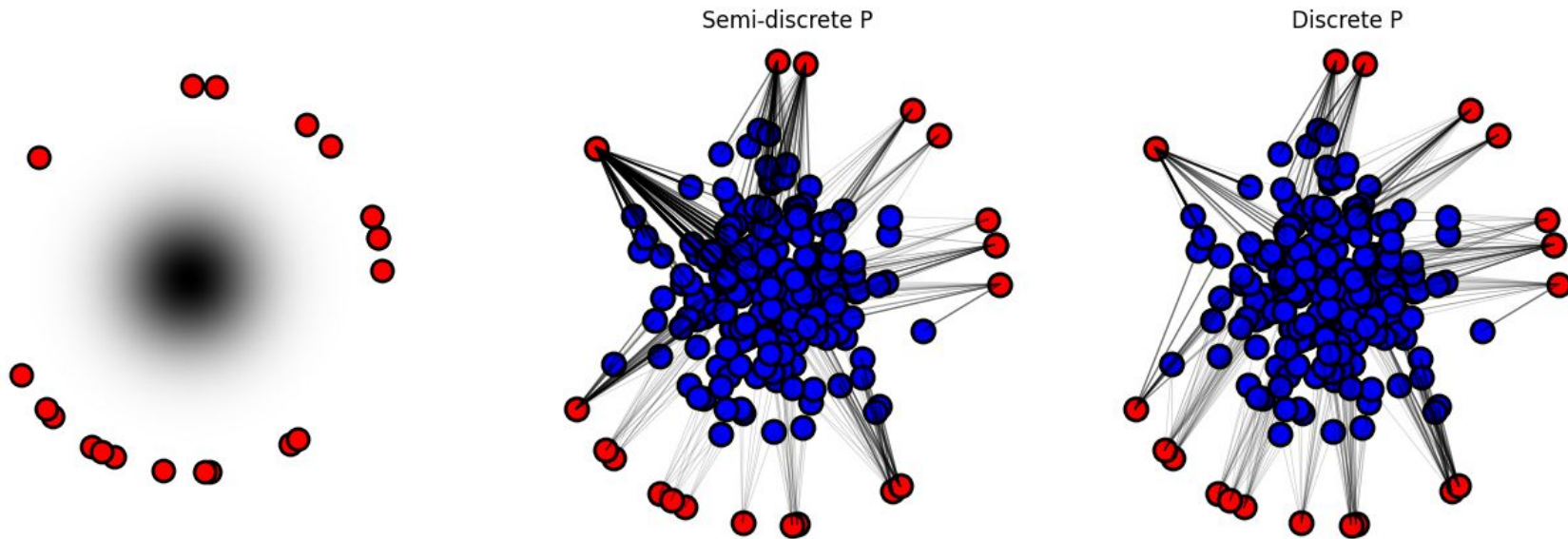
Discrete Optimal Transport

$$OT_{\epsilon}(\alpha, \beta) = \max_{f \in \mathbb{R}^n} \mathbb{E}_Y \left[\sum_{i=1}^n f_i \alpha_i - \epsilon \log \left(\sum_{i=1}^n \exp \left(\frac{f_i - C(x_i, y_j)}{\epsilon} \right) \alpha_i \right) - \epsilon \right]$$

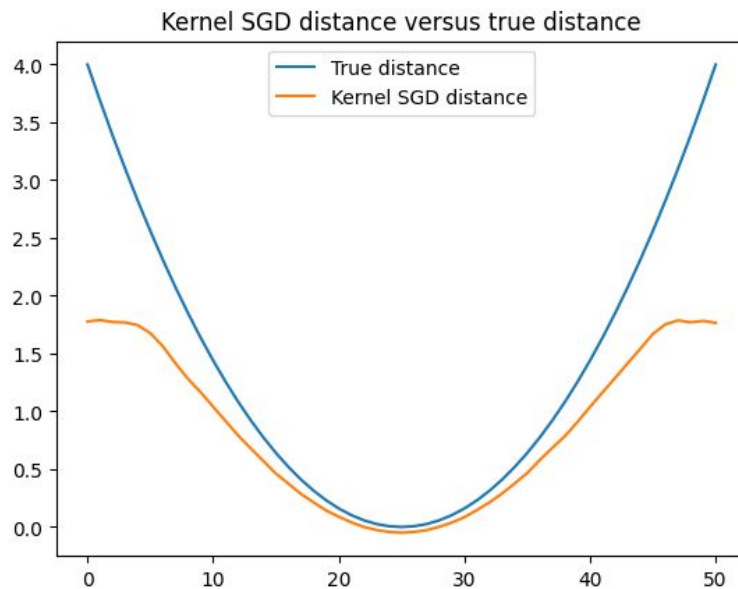
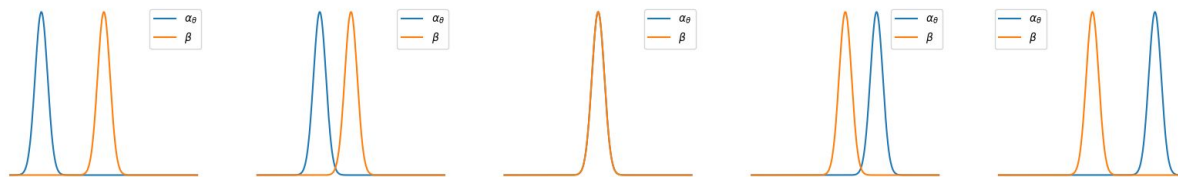


Semi discrete Optimal Transport

$$OT_{\epsilon}(\alpha, \beta) = \max_{f \in \mathbb{R}^n} \mathbb{E}_Y \left[\sum_{i=1}^n f_i \alpha_i - \epsilon \log \left(\sum_{i=1}^n \exp \left(\frac{f_i - C(x_i, y_j)}{\epsilon} \right) \alpha_i \right) - \epsilon \right]$$

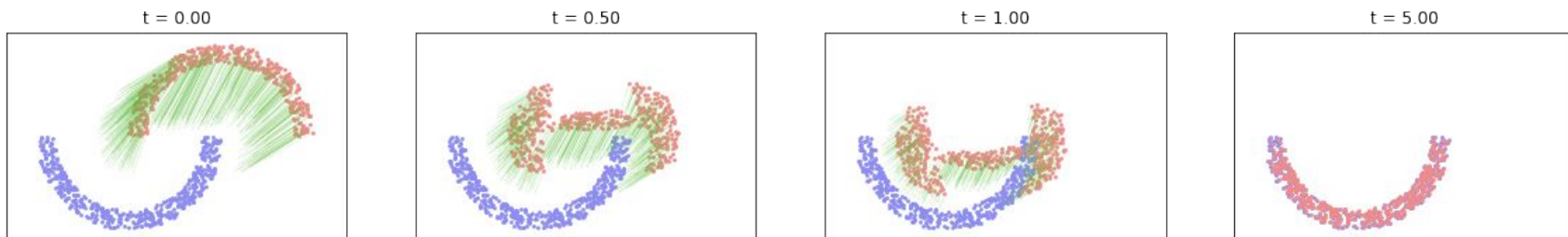


Continuous Optimal Transport



Critics

- The **discrete SAG** is the most important contribution of the paper.
- **Lack of applications** for semi-discrete and continuous cases.
- The paper targets OT applications in Data Science, but it disregards the **entropic bias** and **topology preservation issues** when fitting distributions with OT.



Conclusion

- Big step in **democratizing OT** by expressing it as a **stochastic optimisation problem** → Useful for large scale solvers.
- Many possible improvements for the continuous case in application specific research ?
- Nowadays : OT at scale is almost solved, mostly a discrete problem, continuous “to the limits”.