Data Science
Going further

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Beyond these models

Important topics

- ⋆ Non-linear dimension reduction, manifold learning

→ UMAP, t-SNE, LLE, IsoMap, Laplacian Eigenmaps, Diffusion maps, principal curves, ...

- ⋆ Social Network Analysis
- ⋆ Machine Learning for data modeling

→ VAE, GAN, flows, Transformers, ...

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