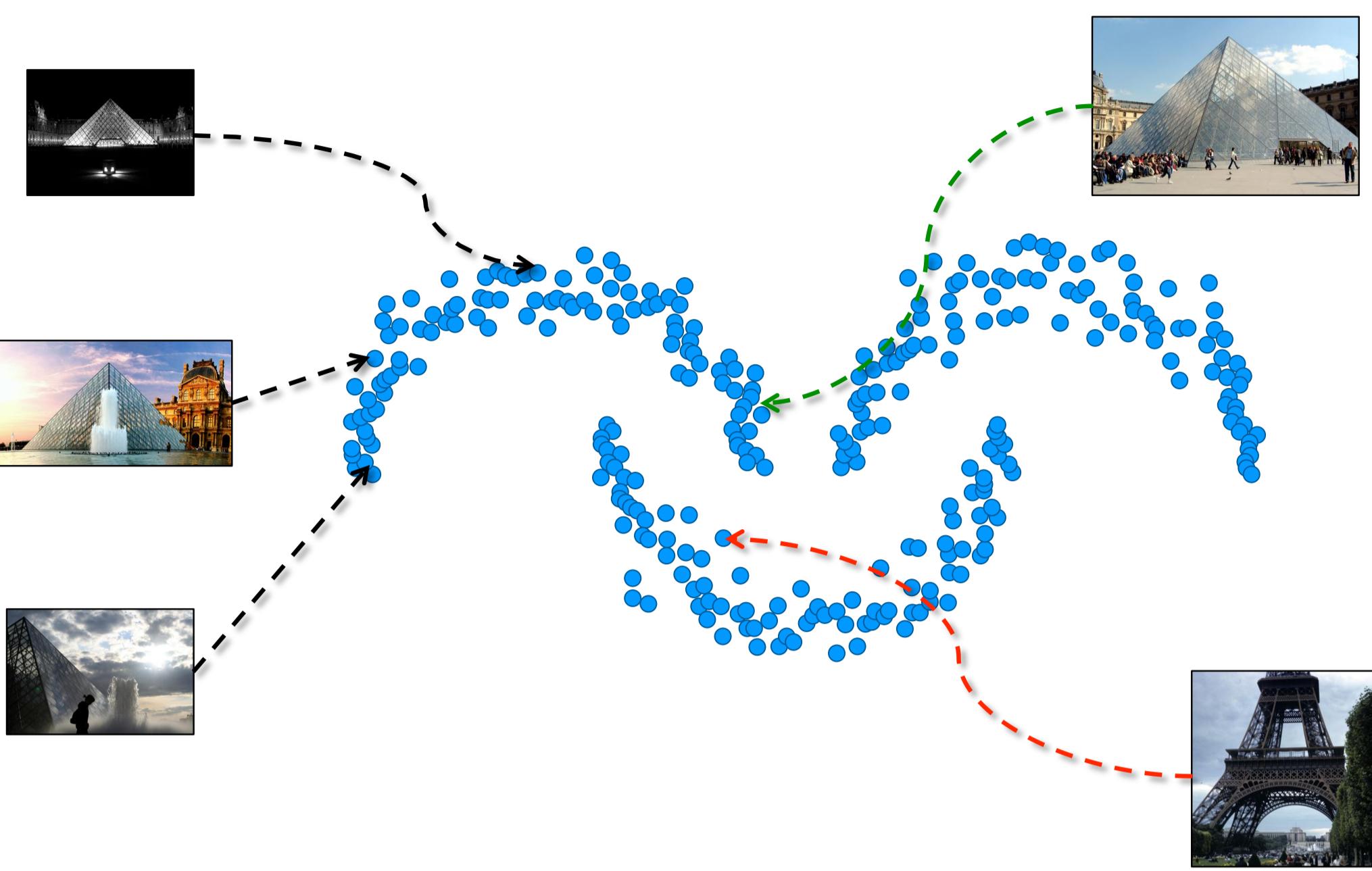


Efficient Diffusion on Region Manifolds: Recovering Small Objects with Compact CNN Representations

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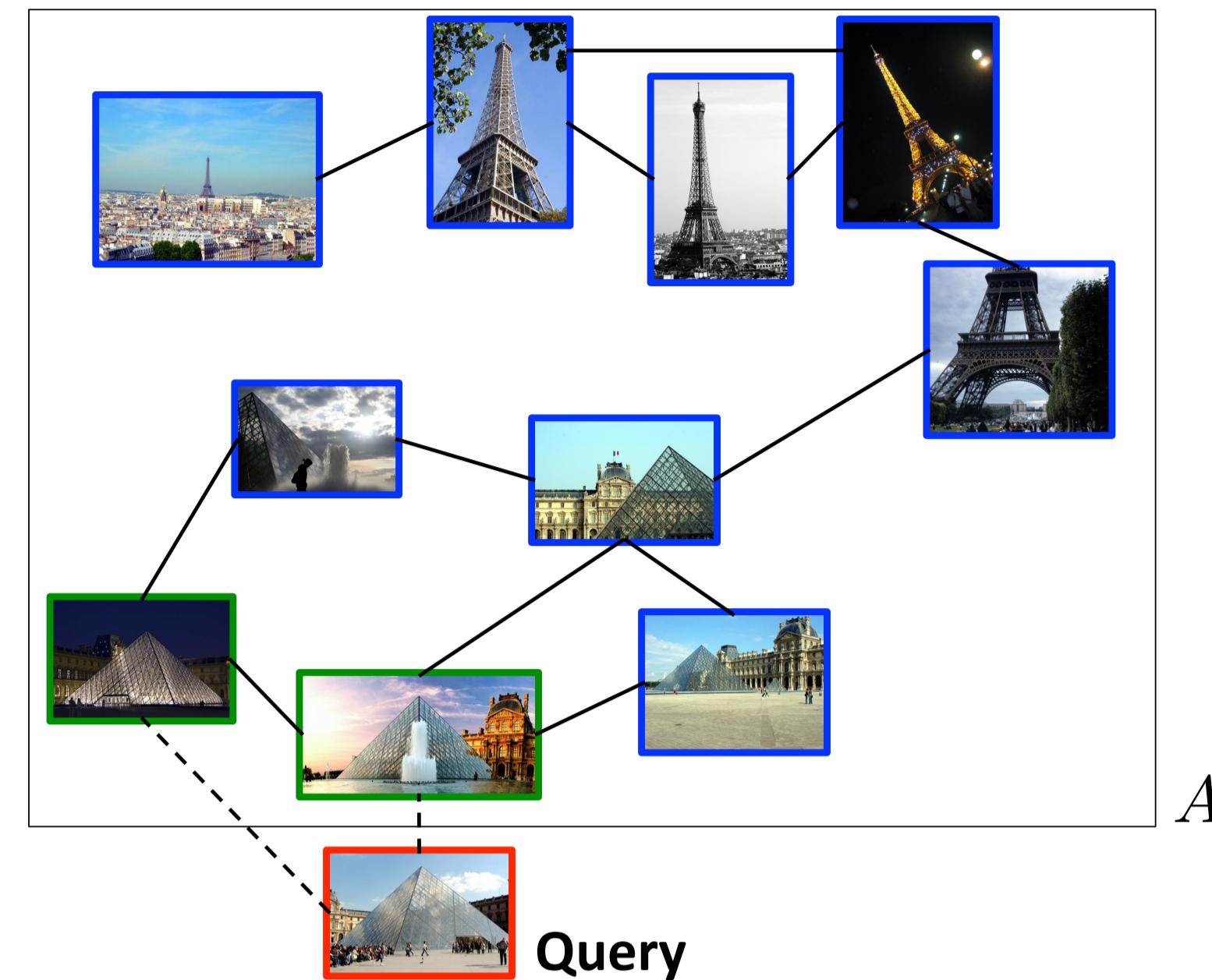
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Motivation

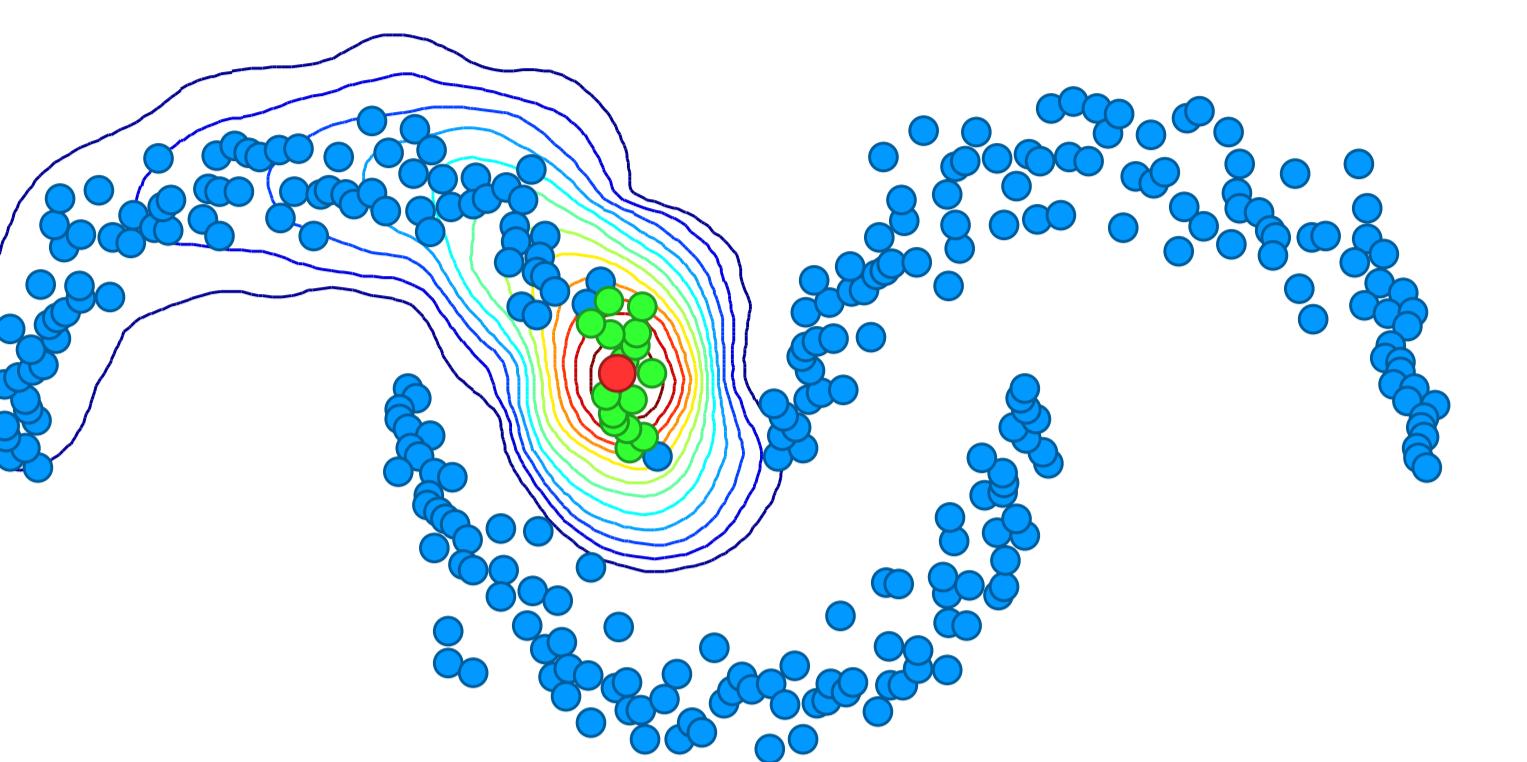


- ▶ Euclidean distance not appropriate for severe visual variations
- ▶ Solution: Ranking on manifolds via graph-based approach, i.e. diffusion [1]

Diffusion for unseen queries



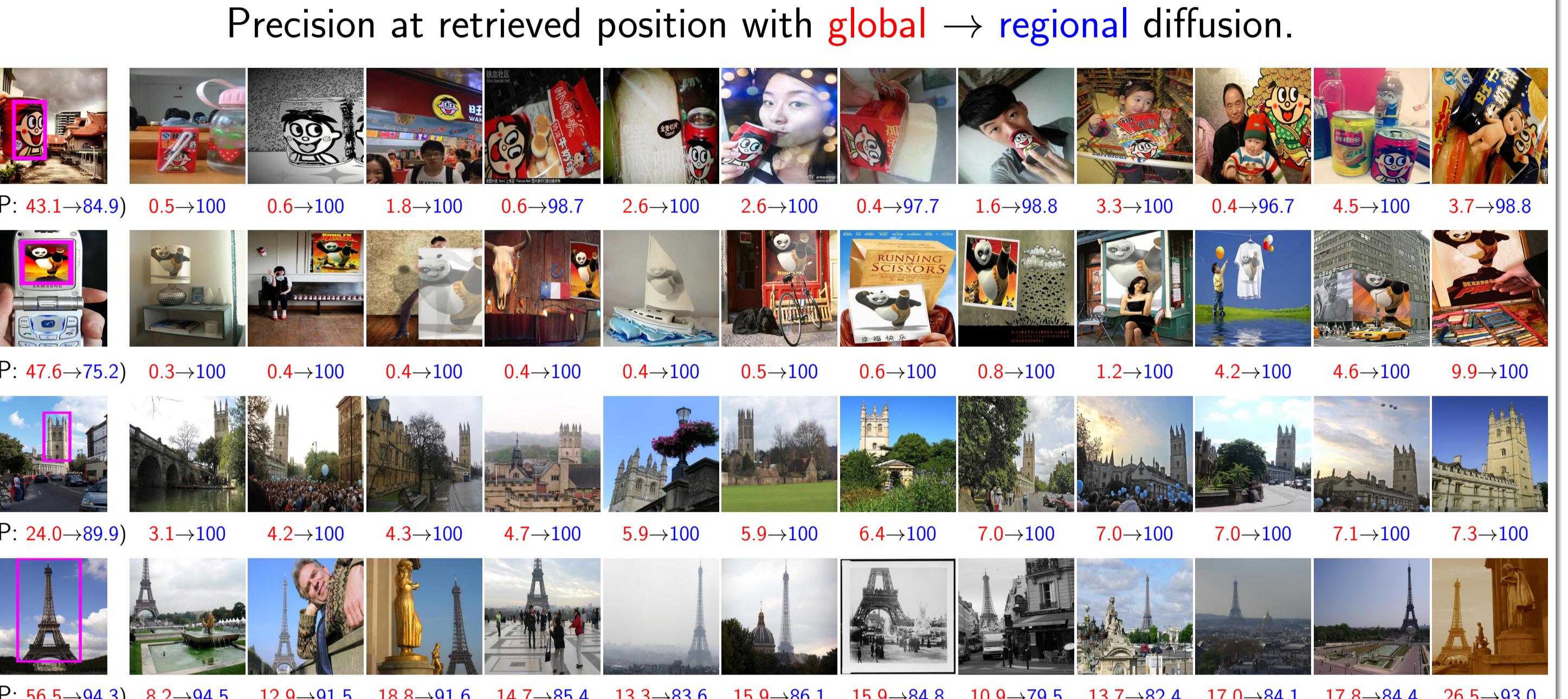
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Toy with 2D database points, query point, kNN of the query, and iso-contours for manifold similarity

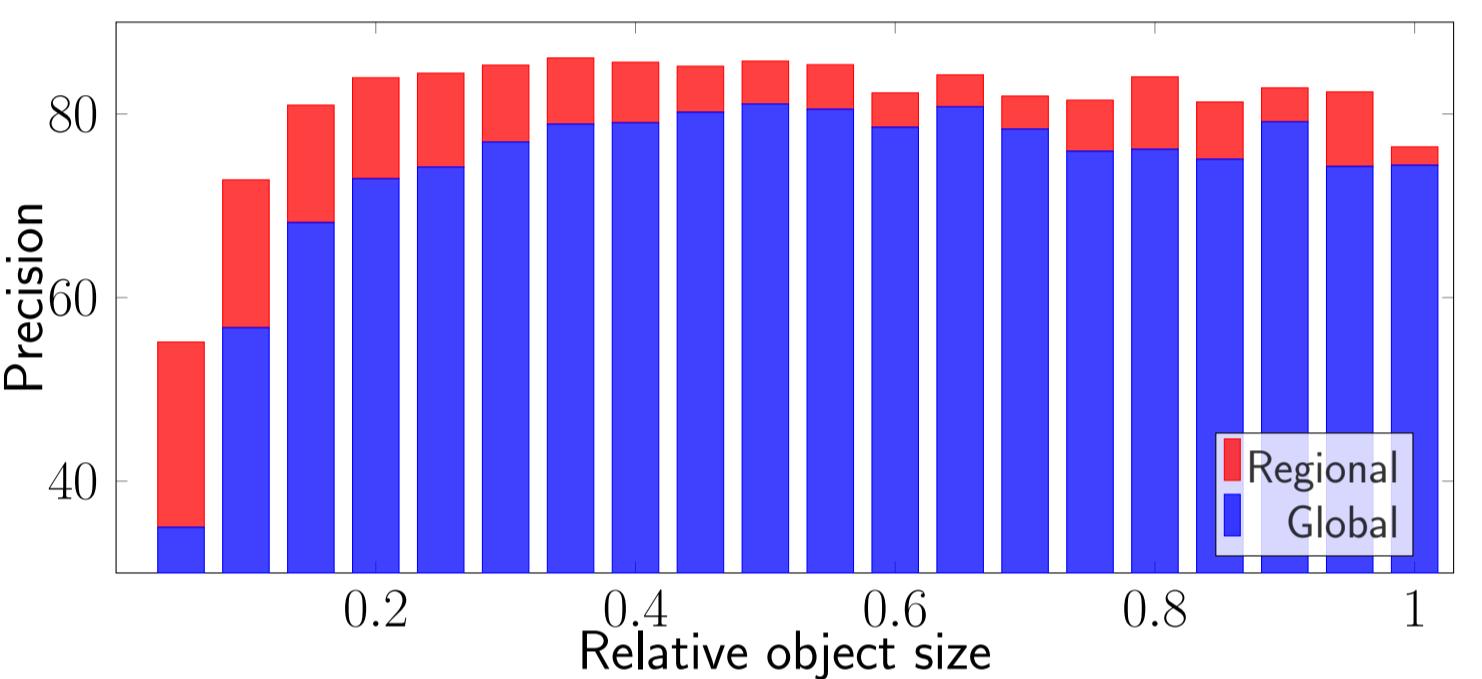
- ▶ Image retrieval with unseen queries: not part of the graph
- ▶ **Contribution:** Instead of searching for the query, search for its neighbors:
 $y_i = 1$ (or equal to similarity) if i -th node is a kNN of the query, $y_i = 0$ otherwise

Retrieval of Small Objects



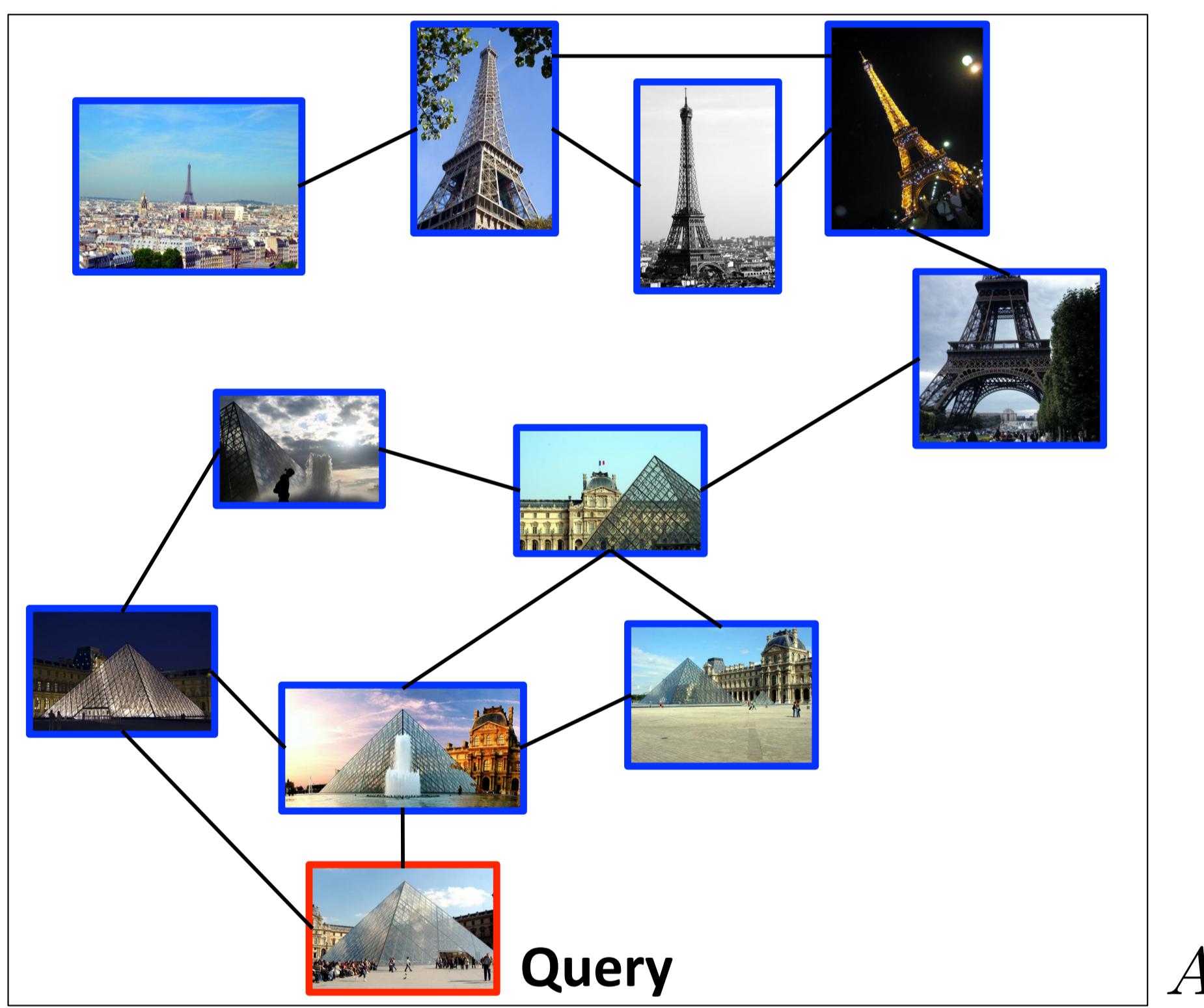
Precision at retrieved position with global → regional diffusion.

Query images (left: bounding box) and retrieved images with the largest improvement by regional diffusion.



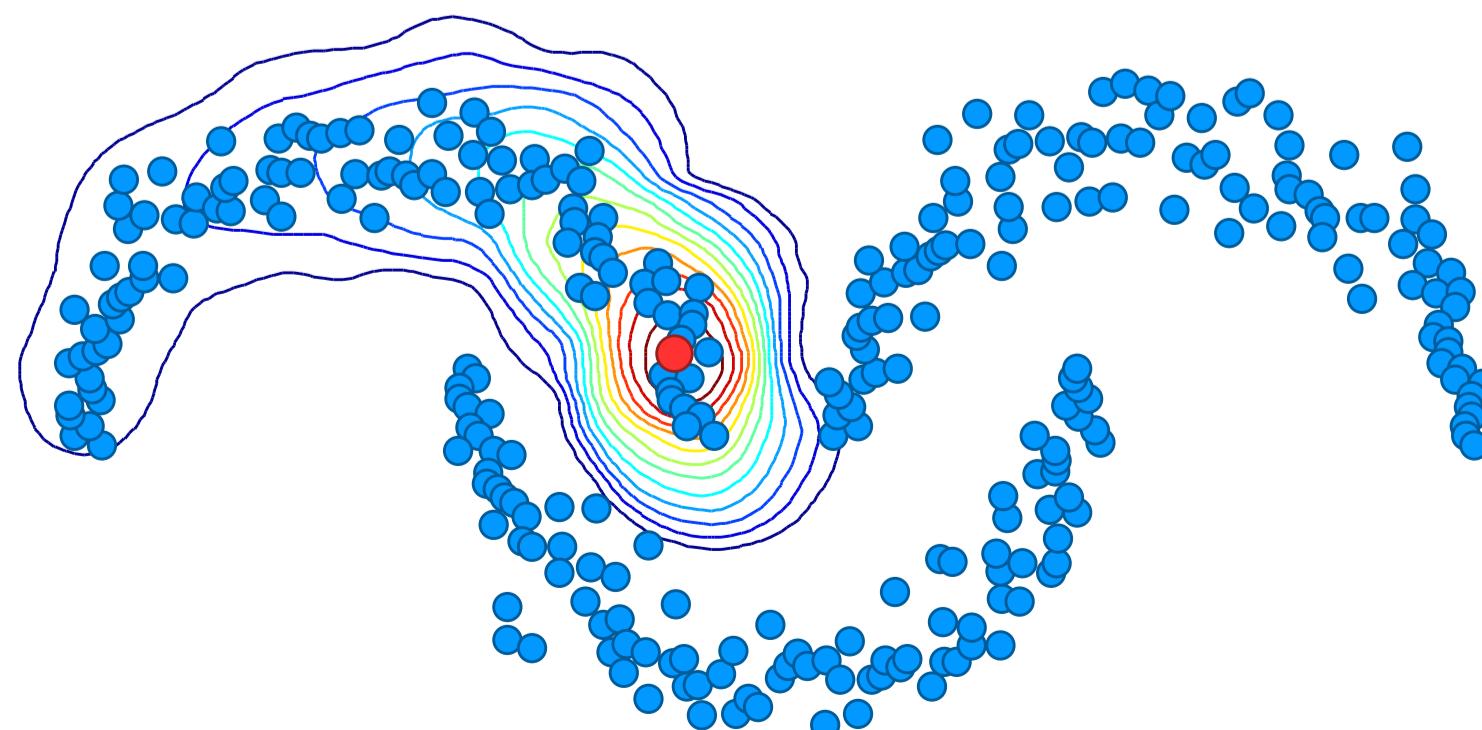
Precision at retrieved position on INSTRE dataset (averaged over positive images according relative object size)

Standard Diffusion [3]



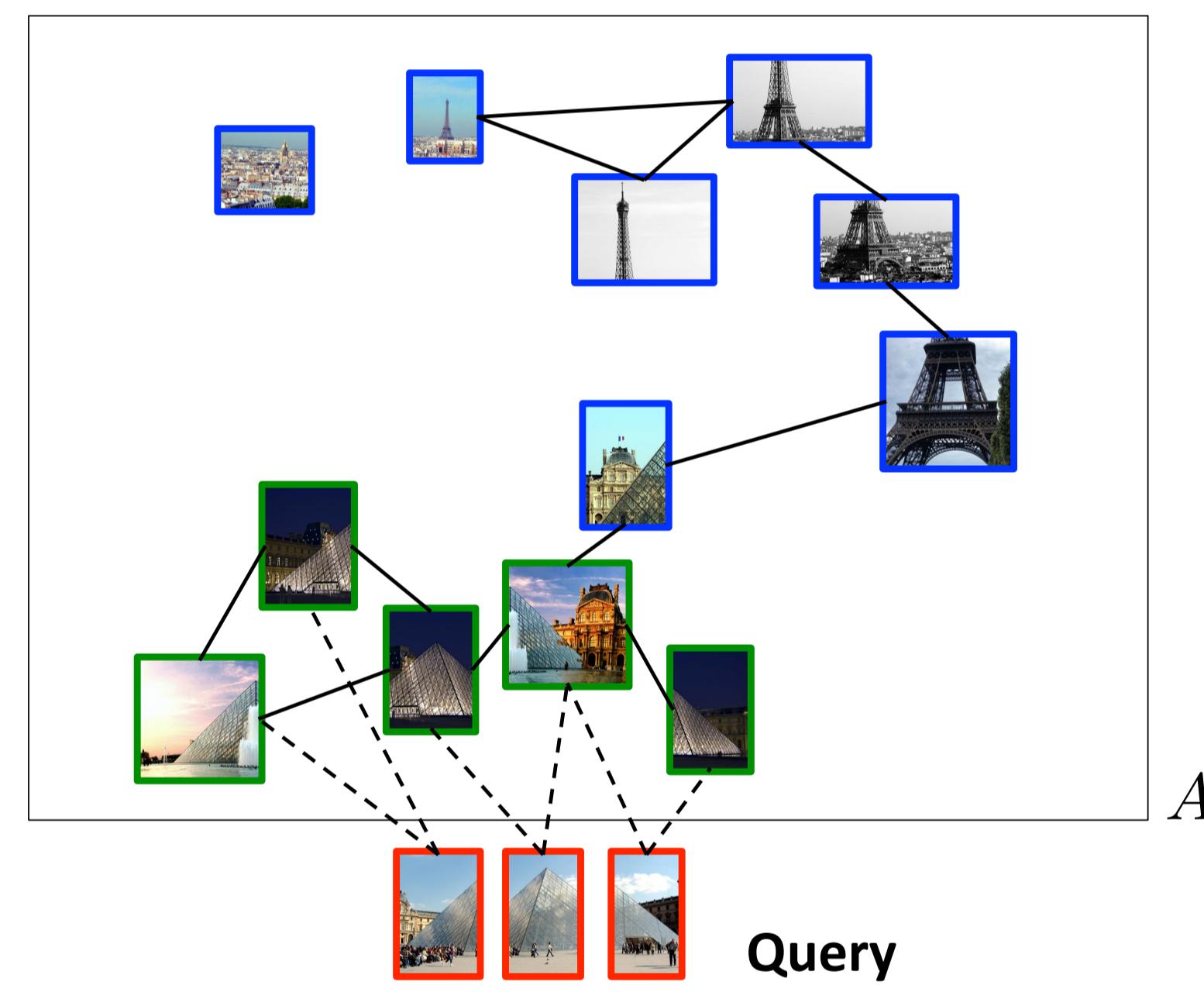
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- ▶ Normalized affinity (reciprocal kNN) matrix: $S := D^{-1/2}AD^{-1/2}$
- ▶ The query is part of the graph
- ▶ $\mathbf{y} = (y_i) \in \mathbb{R}^n$, $y_i = 1$ if i -th node is a query, $y_i = 0$ otherwise
- ▶ Iterative solution preferred in prior work [1]
- ▶ $\mathbf{f}^t = \alpha S \mathbf{f}^{t-1} + (1 - \alpha) \mathbf{y}$
- ▶ Closed-form solution [3] commonly avoided
- ▶ $\mathbf{f}^* = (1 - \alpha) \mathcal{L}_\alpha^{-1} \mathbf{y}$, where $\mathcal{L}_\alpha := I_n - \alpha S$

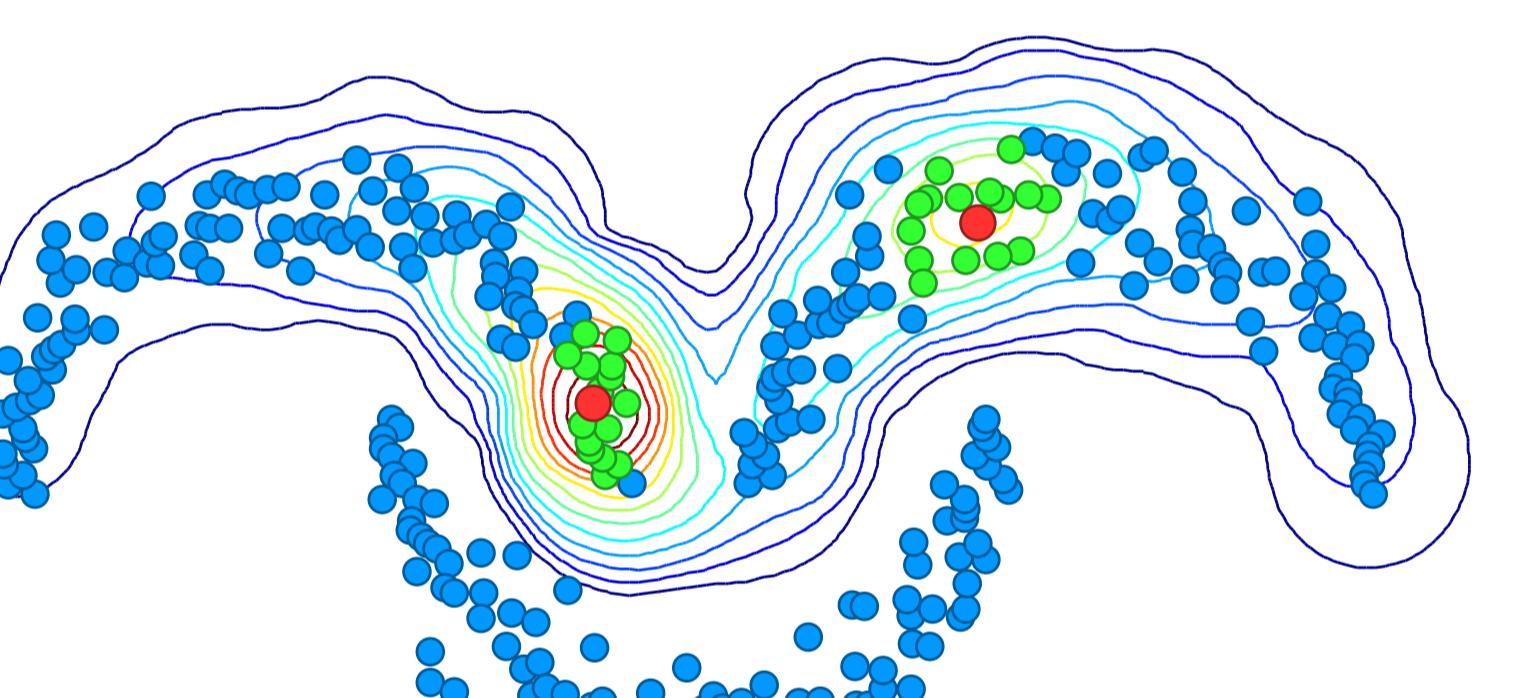


Toy with 2D database points, query point, and iso-contours for manifold similarity

Regional diffusion



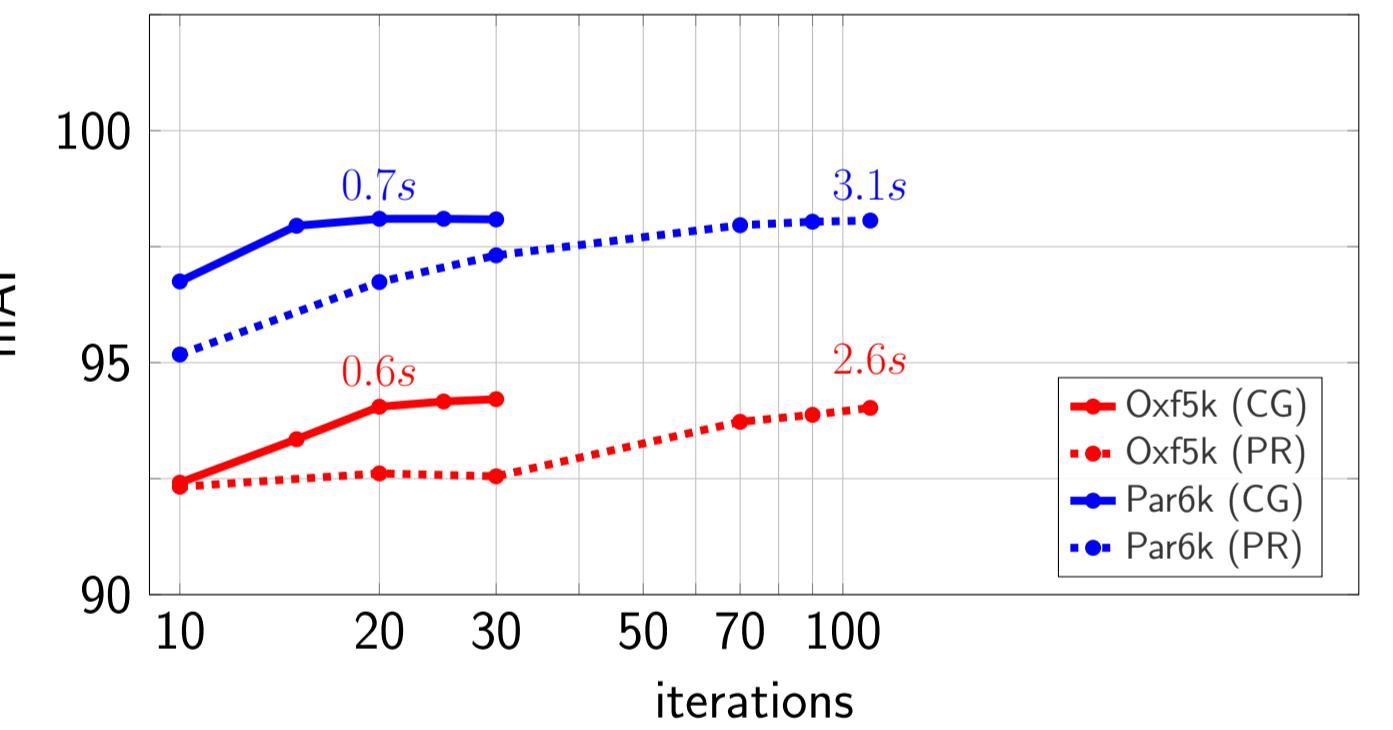
A



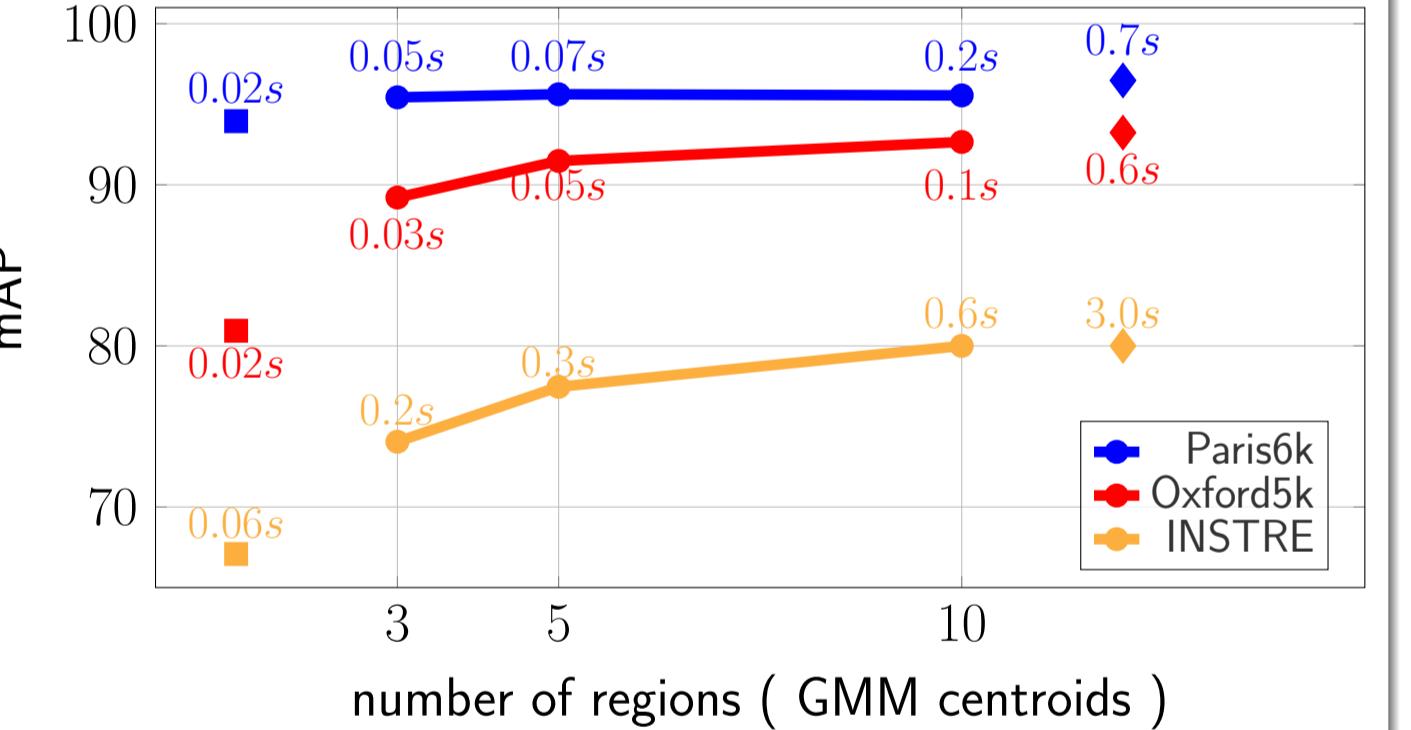
Toy with multiple query points

- ▶ Global descriptors not effective for small objects, occlusion.
- ▶ Represent images by uniformly sampled overlapping regions [2]: each image represented by m vectors
- ▶ **Contribution:** Diffusion with regions as nodes, multiple regional queries issued with the cost of one
- ▶ $y_i = 1$ (or equal to similarity) if i -th node is a kNN of any query region, $y_i = 0$ otherwise

Experiments



Speed and convergence comparison for regional diffusion between the iterative solution (PR) and ours with conjugate gradient (CG)



Performance and speed comparison vs number of vectors/image.

□: global diffusion, ◇: default grid with 21 regions per image

Method	$m \times d$	INSTRE	Oxf5k	Oxf105k	Par6k	Par106k
Regional descriptors - nearest neighbor search						
R-match [2]	21×512	55.5	81.5	76.5	86.1	79.9
R-match [2]	21×2,048	71.0	88.1	85.7	94.9	91.3
Regional descriptors - query expansion						
Hamming Query Expansion	2.4k×128	74.7	89.4 [†]	84.0 [†]	82.8 [†]	-
R-match [2]+AQE	21×512	60.4	83.6	78.6	87.0	81.0
Regional diffusion*	5×512	77.5	91.5	84.7	95.6	93.0
Regional diffusion*	21×512	80.0	93.2	90.3	96.5	92.6
R-match [2]+AQE	21×2,048	77.1	91.0	89.6	95.5	92.5
Regional diffusion*	5×2,048	88.4	95.0	90.0	96.4	95.8
Regional diffusion*	21×2,048	89.6	95.8	94.2	96.9	95.3

References:

- [1] M. Donoser and H. Bischof. Diffusion processes for retrieval revisited. In CVPR, 2013.
- [2] A. S. Razavian, J. Sullivan, S. Carlsson, and A. Maki. Visual instance retrieval with deep convolutional networks. *ITE Transactions on Media Technology and Applications*, 4:251–258, 2016.
- [3] D. Zhou, J. Weston, A. Gretton, O. Bousquet, and B. Schölkopf. Ranking on data manifolds. In NIPS, 2003.