

Multi-Scale Kernel Bundles for LDDMM

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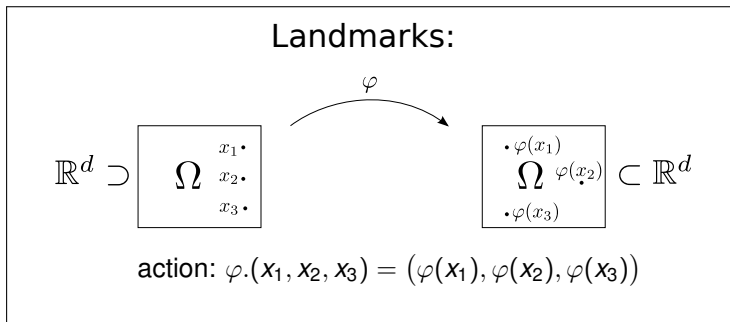


- Outline
 - Short Overview of LDDMM
 - Regularization and Kernels
 - LDDKBM, Kernel Bundles, and Scale
 - Some Experimental Results



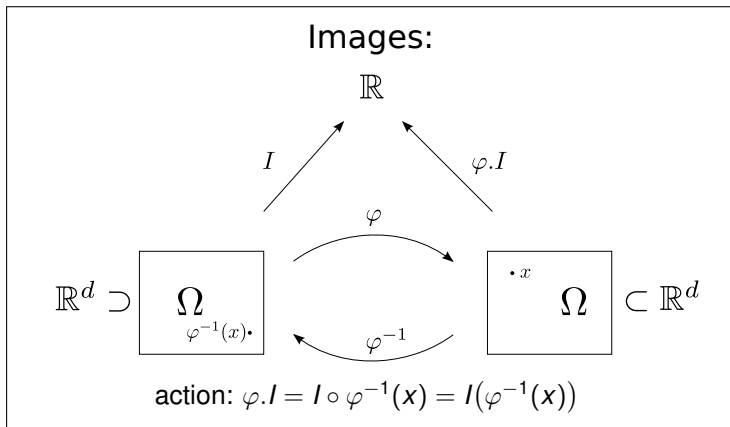
LDDMM: Large Deformation Diffeomorphic Metric Mapping

- diffeomorphisms act on landmarks, images, etc.

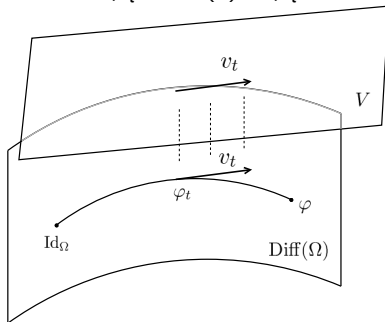


LDDMM: Large Deformation Diffeomorphic Metric Mapping

- diffeomorphisms act on landmarks, images, etc.



- manifold of diffeomorphisms $G_V \subset \text{Diff}(\Omega)$ with tangent space V (subset of vector fields on domain Ω)
- G_V parametrized by time-dependent vector fields $v_t : [0, T] \rightarrow V$
- associated flow $\partial_t \varphi_t^V = v(t) \circ \varphi_t^V$



- norm on V gives metric on G_V



Registration with LDDMM

- search for $\varphi \in G_V$ minimizing

$$E(\varphi) = E_1(\varphi) + \lambda U(\varphi) ,$$

$$E_1(\varphi) = \min_{v_t \in V, \varphi_{01}^v = \varphi} \int_0^1 \|v_s\|_V^2 ds$$

- optimal paths satisfy the *EPDiff* evolution equations
- the search can be phrased in terms of the initial velocity/momentum:

$$E(v_0) = E_1(v_0) + \lambda U(v_0) ,$$

$$E_1(v_0) = \min_{v_t \text{ satisfy EPDiff}} \int_0^1 \|v_s\|_V^2 ds$$

- necessitates a choice of norm $\|\cdot\|_V$



Regularization

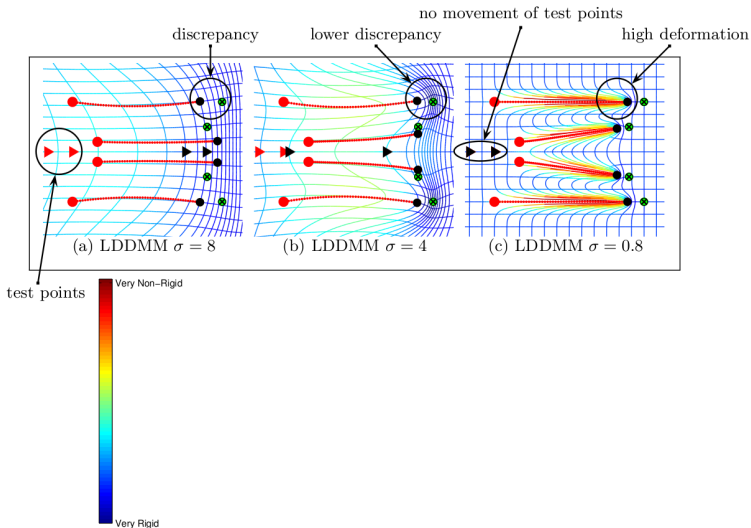
- common norms $\| \cdot \|_V$: Sobolev, induced from Gaussians
- under certain conditions, *RHKS* structure gives correspondence between kernels $K(\cdot, \cdot) \in \mathbb{R}^{d \times d}$ and norm
- the scale of the kernel affects the regularization
- the norm/kernel duality makes the choice of norm concrete, e.g. the EPDiff equations asserts that

$$v_t(\cdot) = \sum_{l=1}^N K(\cdot, x_{t,l}) a_{t,l}$$

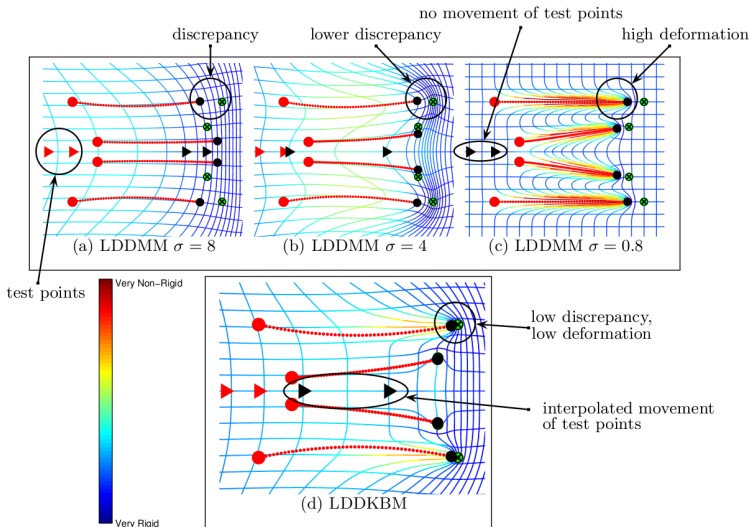
for landmarks



- a simple example: registration of four points with Gaussian kernels of different scales:



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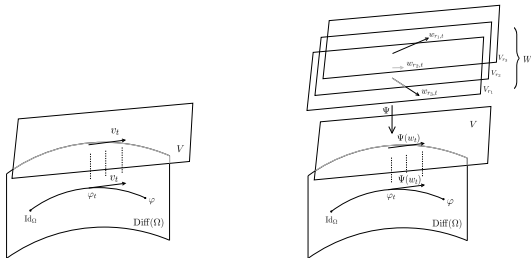


- sparse data provide good examples but the same phenomena arise for image data (consider e.g. areas of constant intensity)
- Modelling questions:
 - what is the right regularization? (if there is one)
 - does deformation occur at different scales?
- Performance improvements:
 - sparse descriptions of large deformations
 - can faster representations be compatible with norm? (sparse initial data must imply sparsity throughout evolution)



LDDKBM: Large Deformation Diffeomorphic Kernel Bundle Mapping

- the tangent space V is extended to a bundle W allowing different kernels/scales

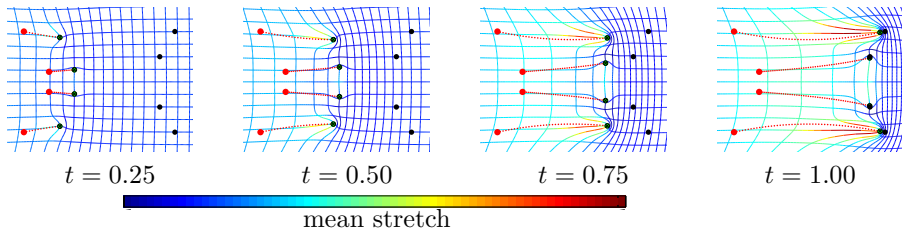


- we regularize using the norm $\|w\|_W^2 = \int_{I_W} \|w_r\|_r^2 dr$ on W (the contribution of each scale is split) and

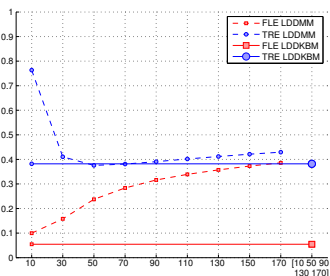
$$E_1^W(\varphi) = \min_{w_t \in W, \varphi_{01}^{\Psi(w)} = \varphi} \int_0^1 \|w_s\|_W^2 ds$$



- Evolution of optimal 3-scale diffeomorphism path:



- in essence: we decouple the contribution of each scale in the registration
- advantages of kernel bundle extension:
 - models deformation at multiple scales/kernel shapes
 - many of the mathematical properties of LDDMM are preserved: momentum conservation, EPDiff evolution equations, well-posedness, etc.
 - statistics using scale information
 - possibility for faster registration: sparsity and smoothness at different scales
 - removes the need for scale selection:



KB-EPDiff equations

- the EPDiff equations

$$v_t = \int_{\Omega} K(\cdot, x) a_t(x) dx ,$$

$$\partial_t a_t = -Da_t v_t - a_t \nabla \cdot v_t - (Dv_t)^T a_t$$

extend to the KB-EPDiff equations (scale: r)

$$w_{r,t} = \int_{\Omega} K_r(\cdot, x) a_{r,t}(x) dx ,$$

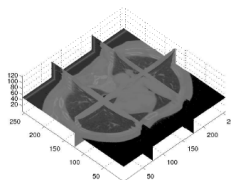
$$\partial_t a_{r,t} = \int_{I_W} -Da_{r,t} w_{s,t} - a_{r,t} \nabla \cdot w_{s,t} - (Dw_{s,t})^T a_{r,t} ds$$

- in a more general form:

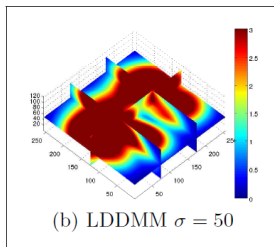
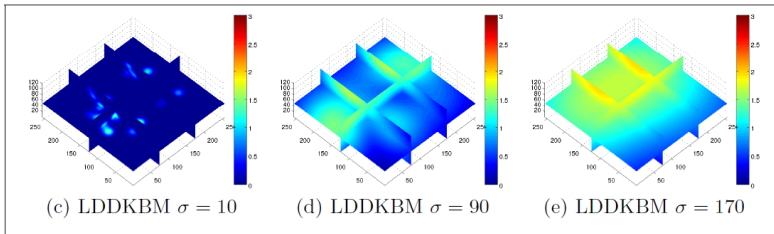
$$w_{t,r} = \text{Ad}_{\varphi_{t0}^{\Psi(w)}}^{T,r} w_{0,r} \quad \text{and} \quad \partial_t a_{t,r} = -\text{ad}_{\Psi(w_t)}^* a_{t,r}$$



- towards sparsity: initial vector fields of lung registration, LDDMM and LDDKBM



(a) Slices of 3D lung image

(b) LDDMM $\sigma = 50$ (c) LDDKBM $\sigma = 10$ (d) LDDKBM $\sigma = 90$ (e) LDDKBM $\sigma = 170$ 

- References:

- *A Multi-Scale Kernel Bundle for LDDMM: Towards Sparse Deformation Description Across Space and Scales*

S. Sommer, M. Nielsen, F. Lauze, X. Pennec, IPMI 2011

- *Kernel Bundle EPDiff: Evolution Equations for Multi-Scale Diffeomorphic Image Registration*

S. Sommer, F. Lauze, M. Nielsen, X. Pennec, SSVM 2011

- *Accelerating Multi-Scale Flows for LDDKBM Diffeomorphic Registration*

S. Sommer, GPU CV workshop at ICCV 2011

