

Discriminative Image Warping using Attribute Flow

Jianbo Shi

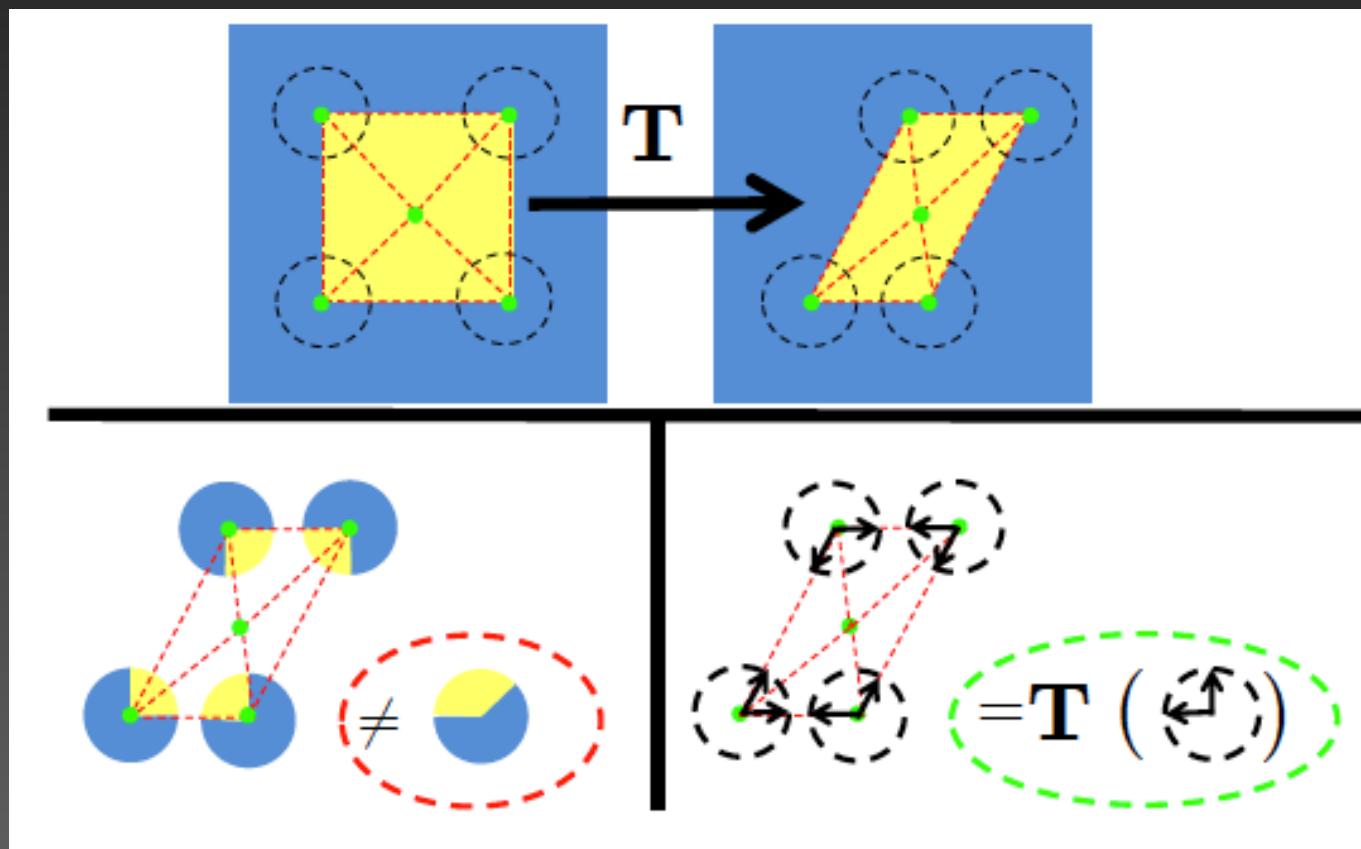
Joint work with Weiyu Zhang and Praveen Srinivasan

GRASP Laboratory



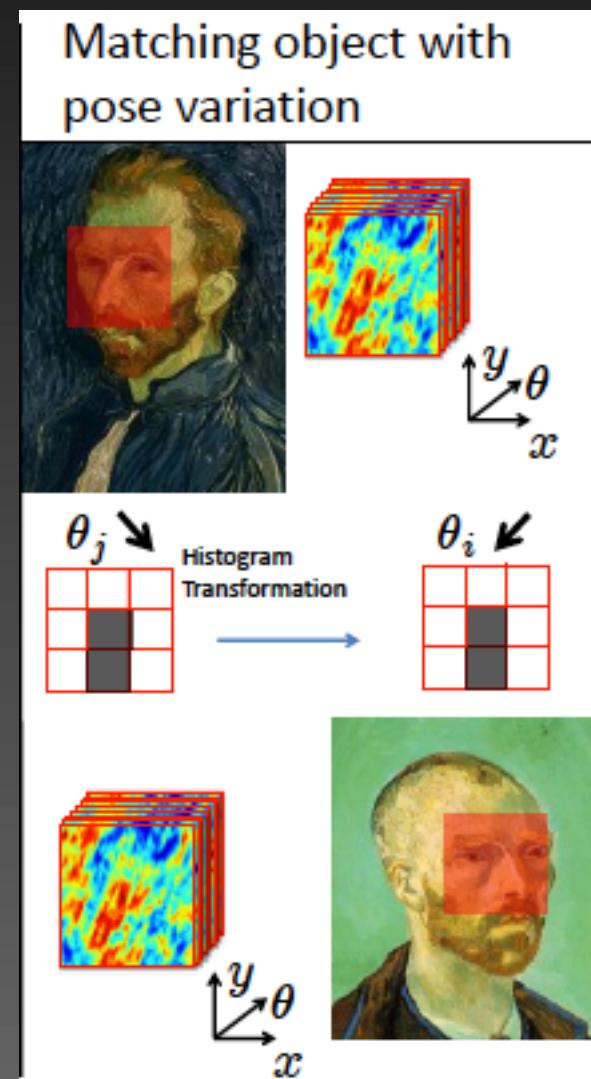
Attribute flow : idea 1

need to frame features right



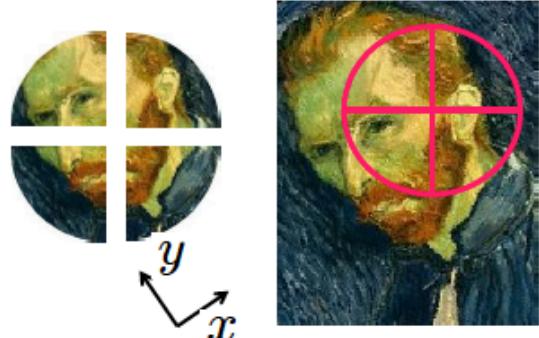
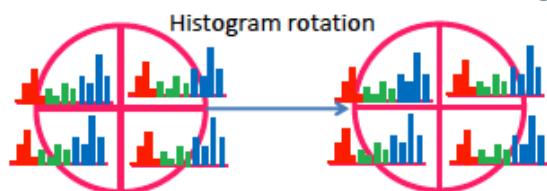
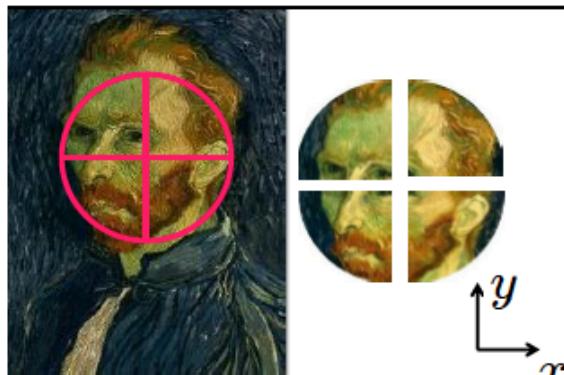
Attribute flow : idea 2

Transformation that explains
attribute change

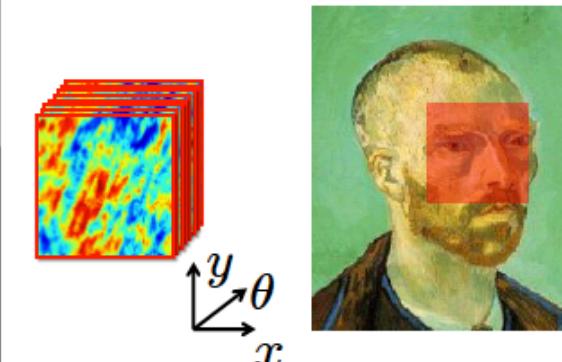
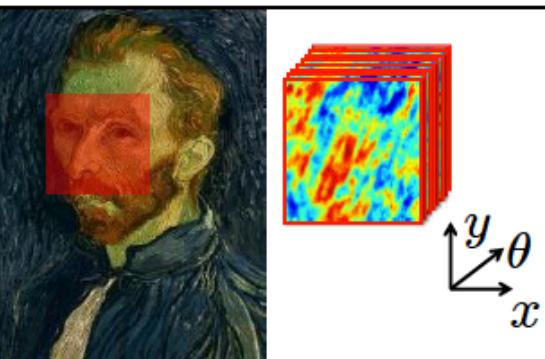


Attribute Flow

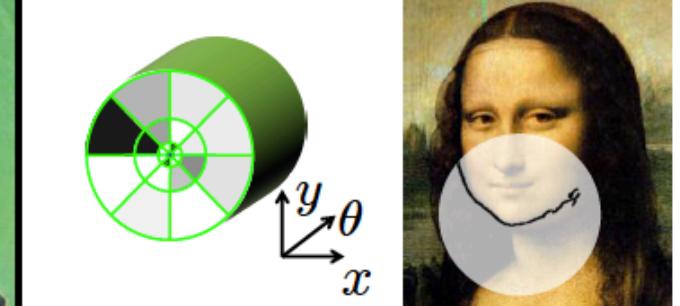
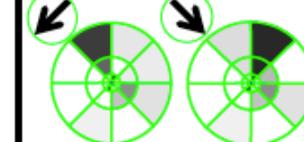
Matching same object
with deformation



Matching object with
pose variation



Matching instances in the
same object category



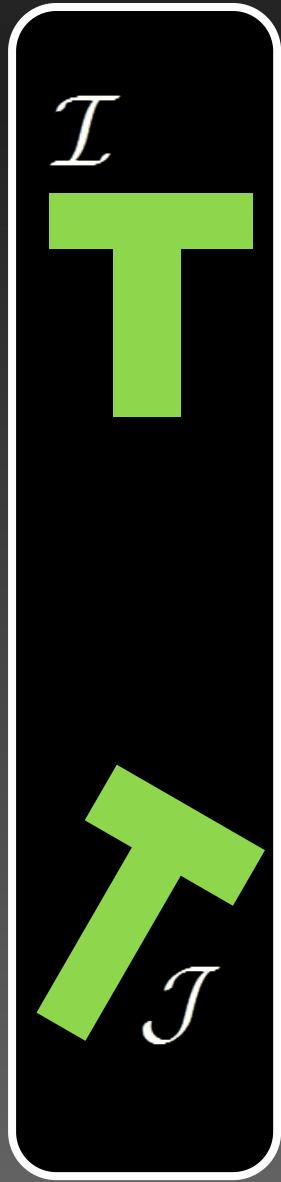
Challenges

Q1: How to compute the spatial transformation efficiently without explicit search?

Q2: How to constrain the computation such that it leads to valid spatial transformation?

Q3: How to deal with clutter environment?

Image Deformation as Attribute Flow



Input Model

Input Image



Transformation that explains attribute change

$$\min_{\mathbf{T}} \int_{\mathbb{R}^2} \|\mathcal{A}_{\mathcal{I}}(x) - \mathcal{A}_{\mathcal{J}}(\mathbf{T}(x))\|_p dx$$

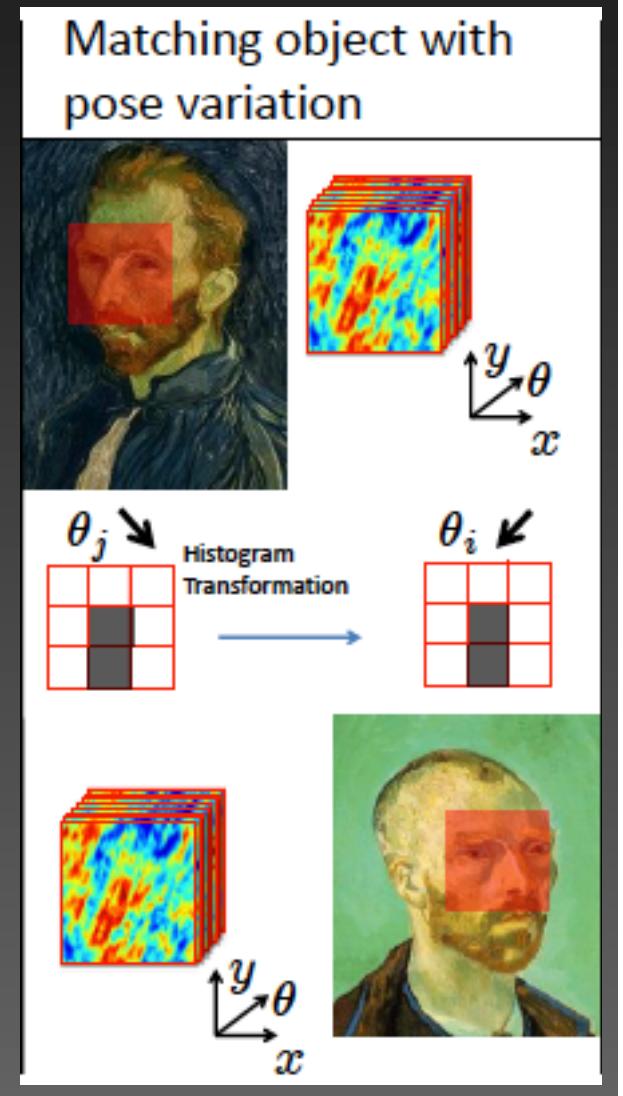
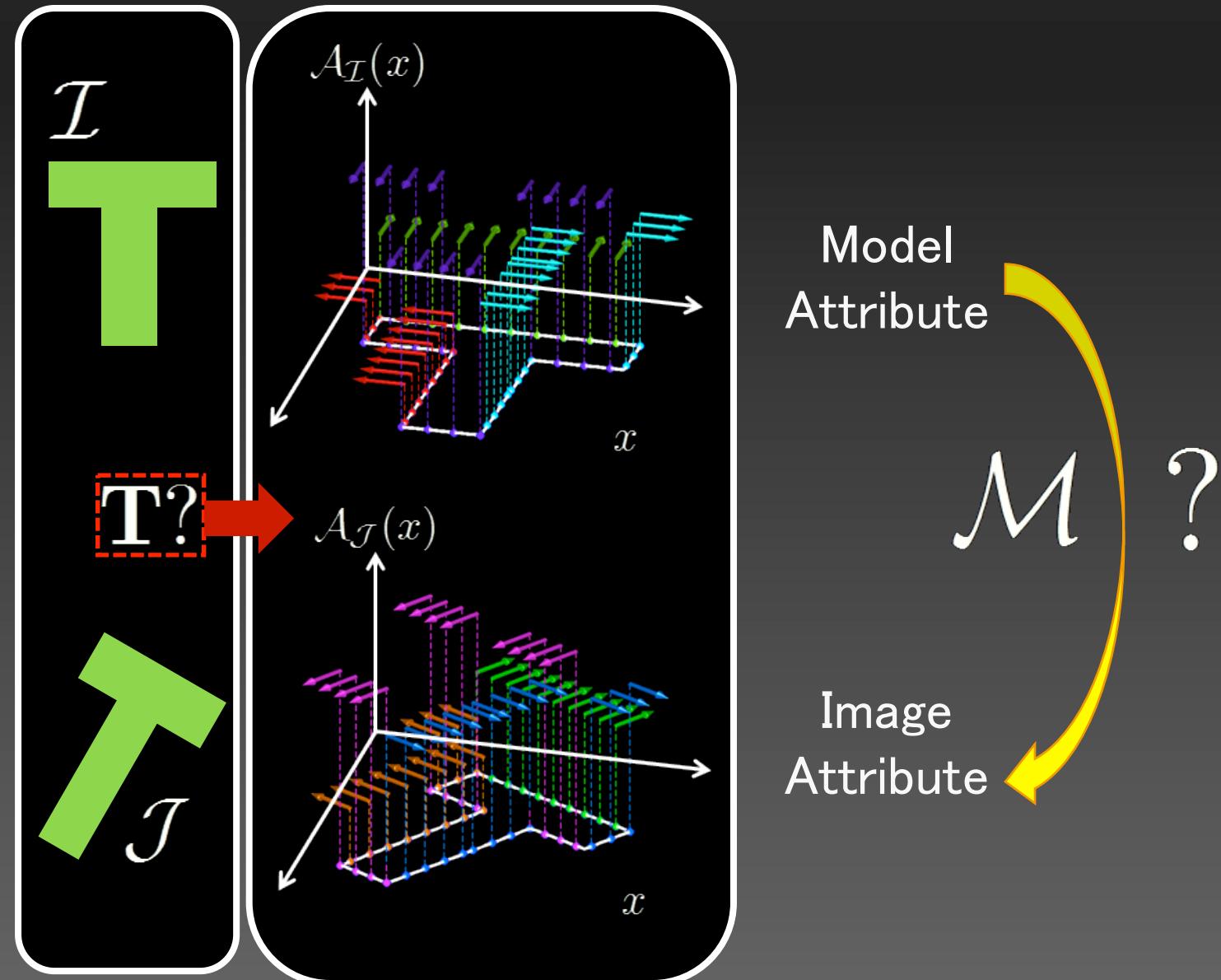


Image Deformation as Attribute Flow

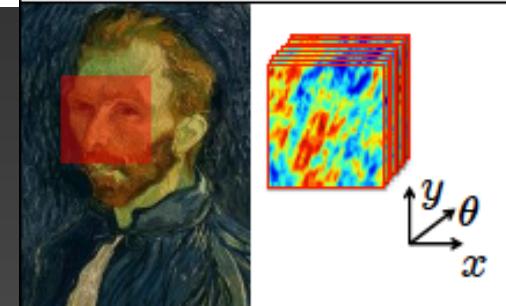


$$\min_{\mathcal{M}} \int_{\mathbf{R}^{n+2}} |\delta(y - \mathcal{M}(\mathcal{A}_{\mathcal{I}}^+)) - \delta(y - \mathcal{A}_{\mathcal{J}}^+)|_1 dy$$

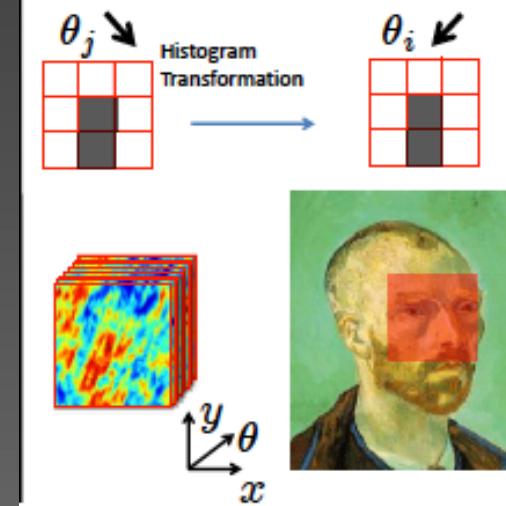
Assuming the first elements of $\mathcal{A}(\mathcal{I}(x))$ are the spatial location x

$$\mathcal{M}(\mathcal{A}(\mathcal{I}(x))) = (\mathbf{T}(x), m(a_2), \dots, m(a_n)).$$

Matching object with pose variation



Attribute: edge orientation, or oriented filter



$$\min_{\mathcal{M}} \int_{\mathbf{R}^{n+2}} |\delta(y - \mathcal{M}(\mathcal{A}_{\mathcal{I}}^+)) - \delta(y - \mathcal{A}_{\mathcal{J}}^+)|_1 dy$$

Assuming the first elements of $\mathcal{A}(\mathcal{I}(x))$ are the spatial location x

$$\mathcal{M}(\mathcal{A}(\mathcal{I}(x))) = (\mathbf{T}(x), m(a_2), \dots, m(a_n)).$$

Matching instances in the same object category

Attribute: **edge orientation**, or oriented filter

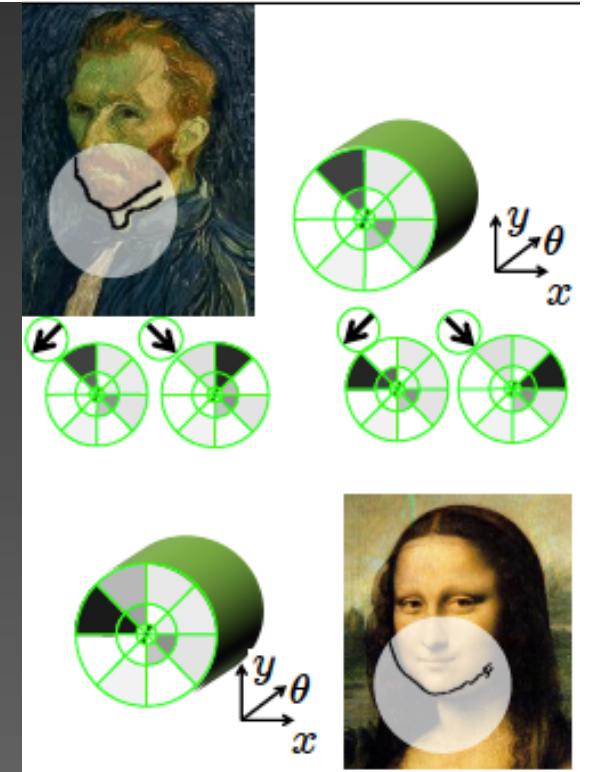
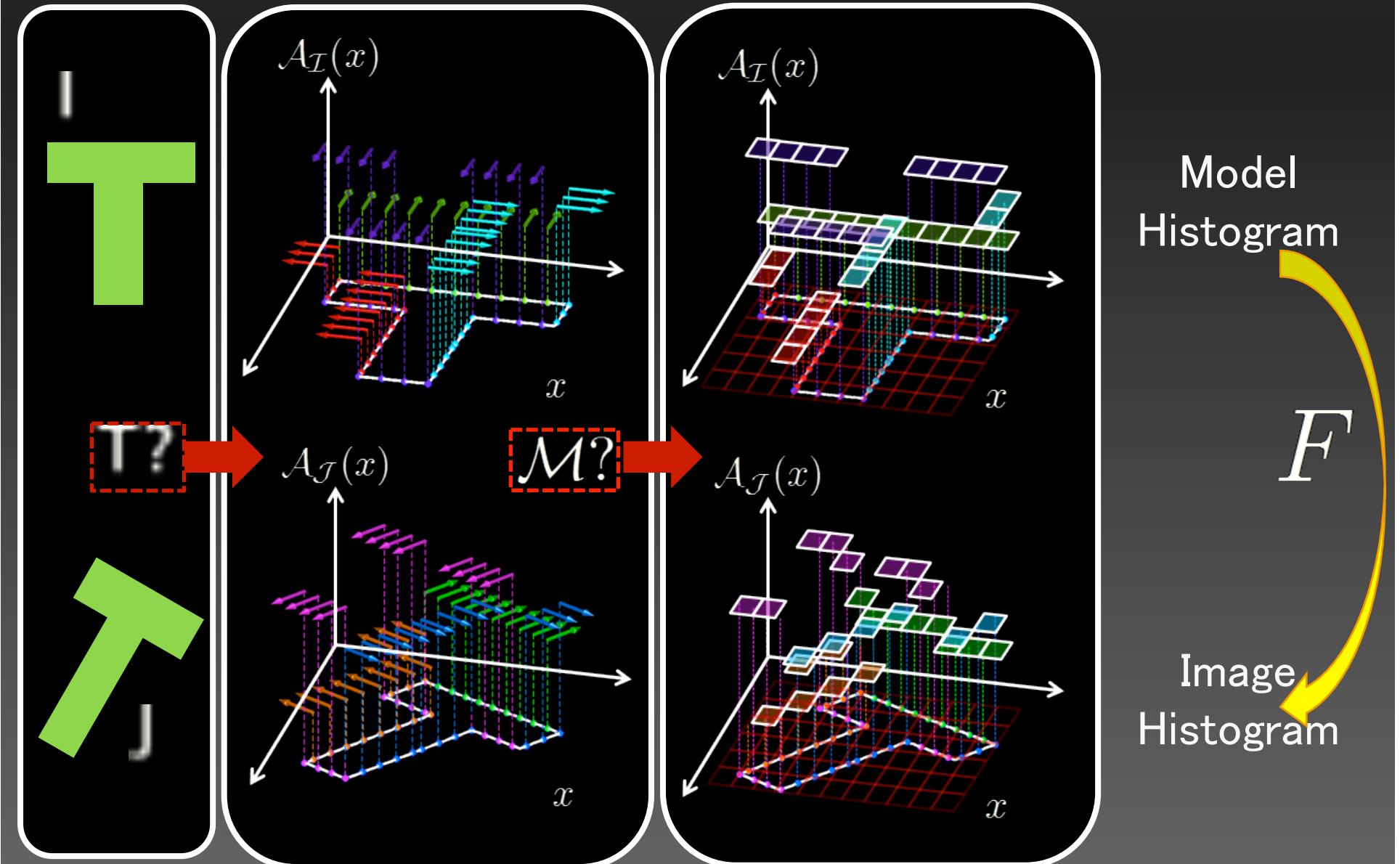


Image Deformation as Attribute Flow



Solve Attribute Flow via Histogram Matching

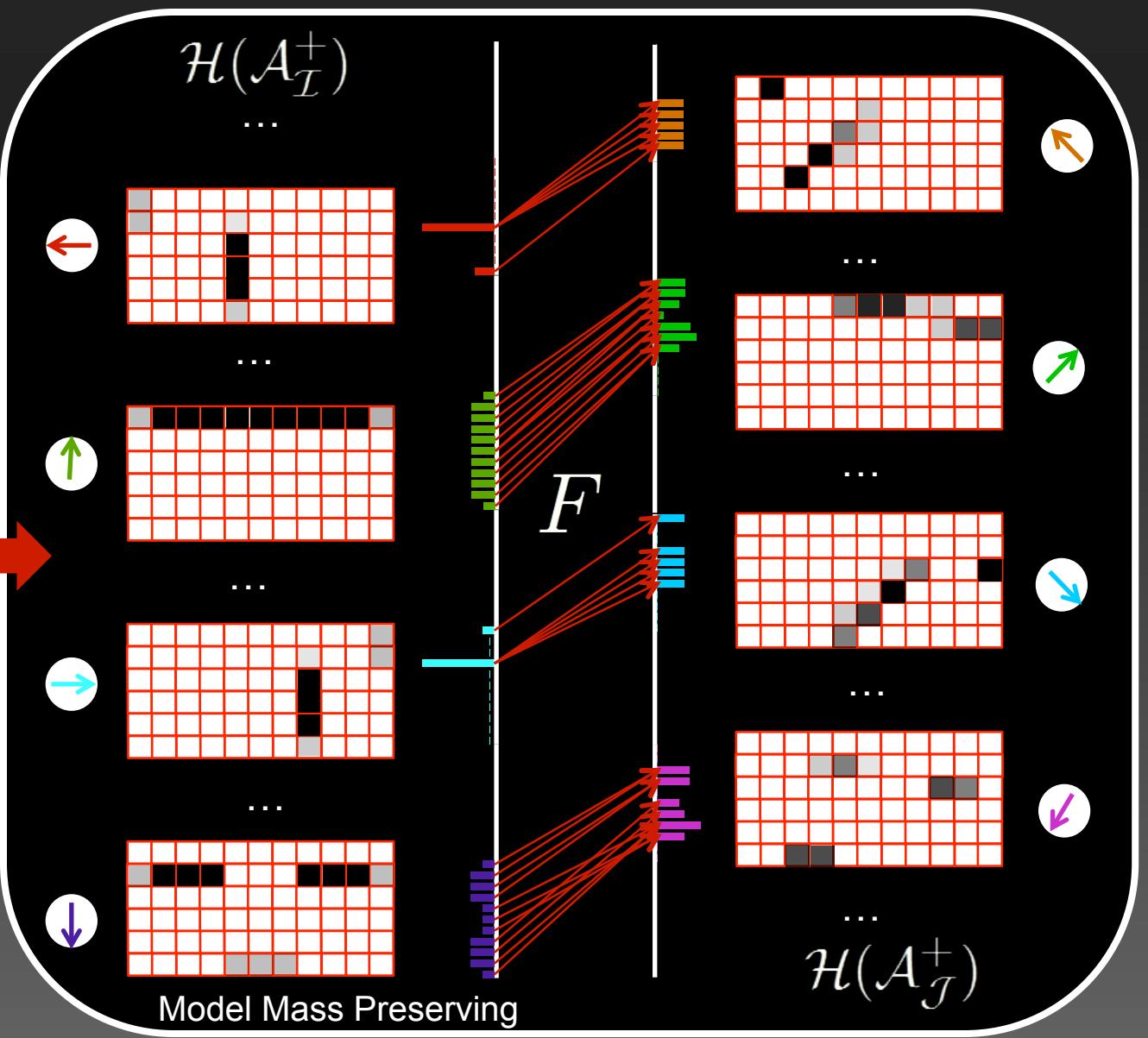
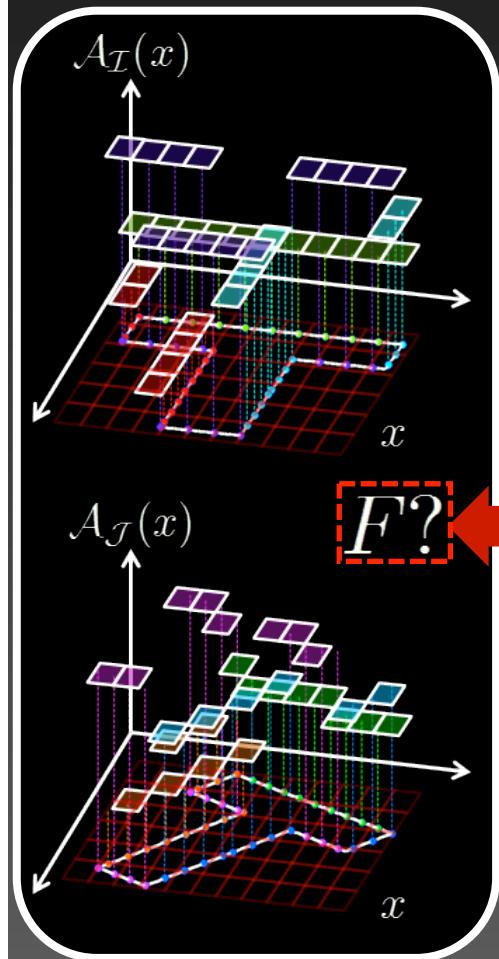
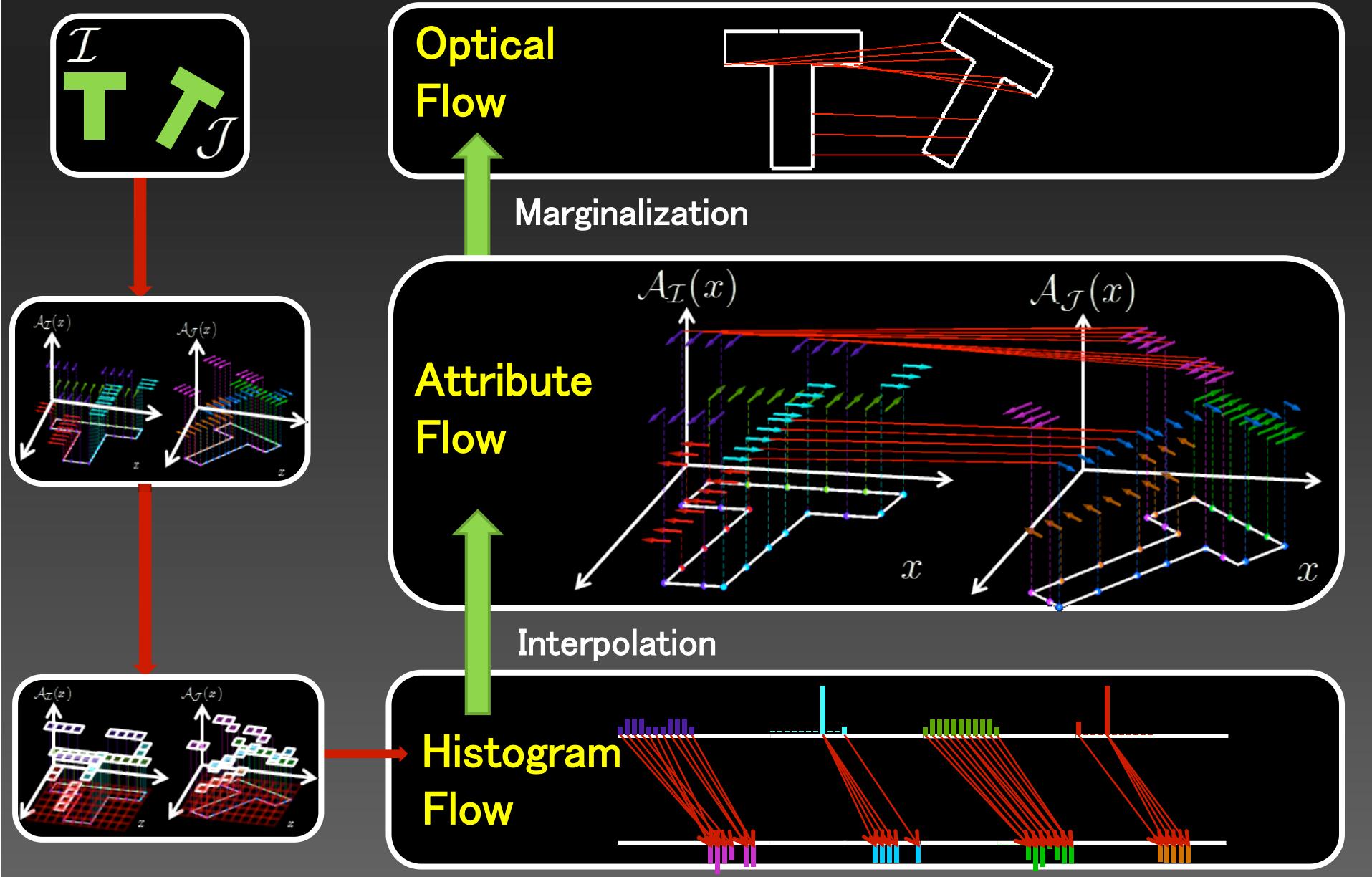
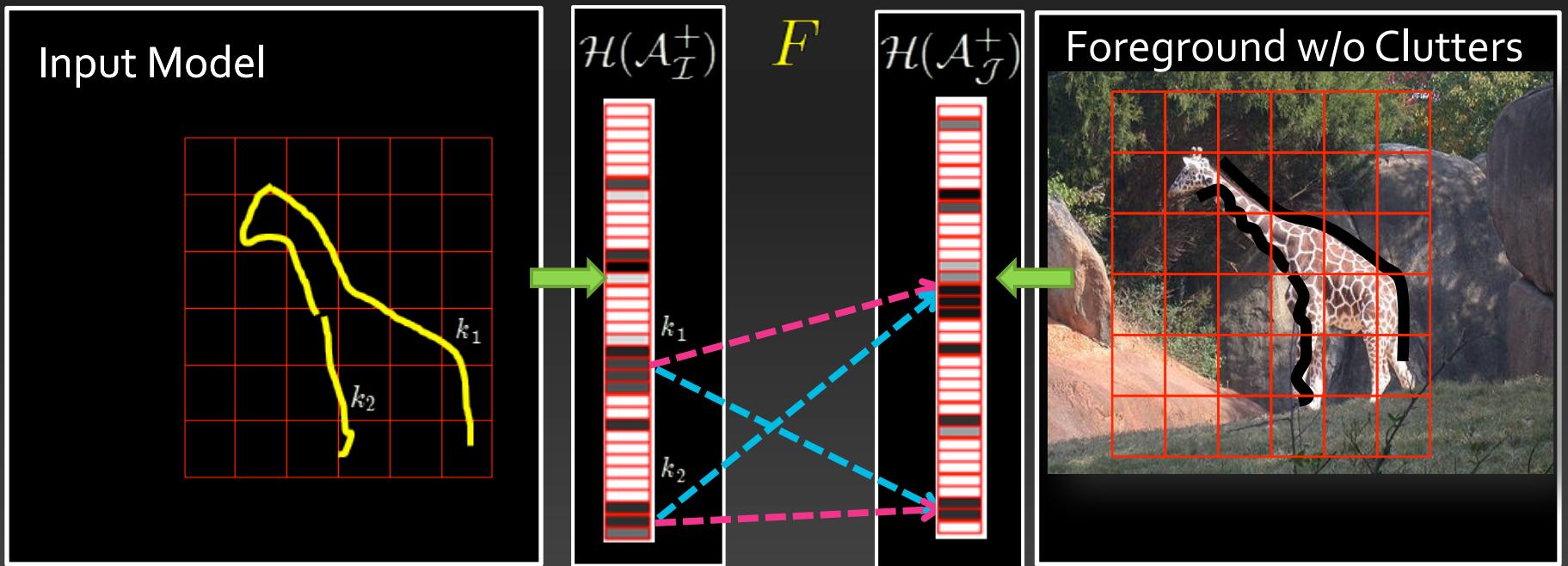


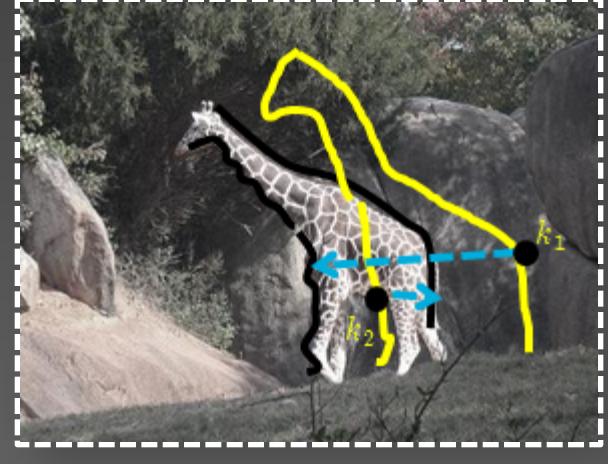
Image Deformation as Attribute Flow



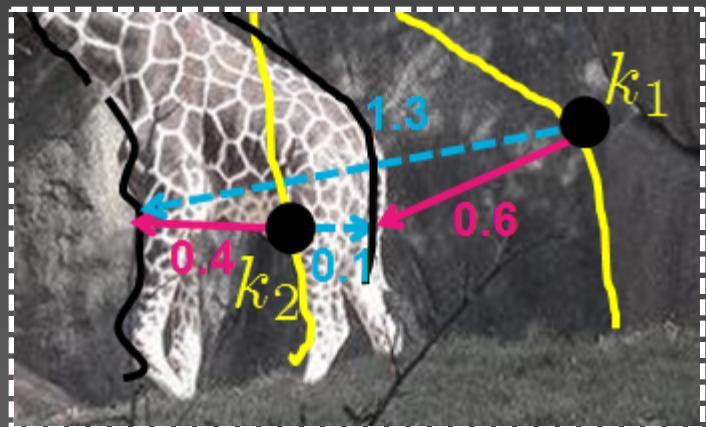
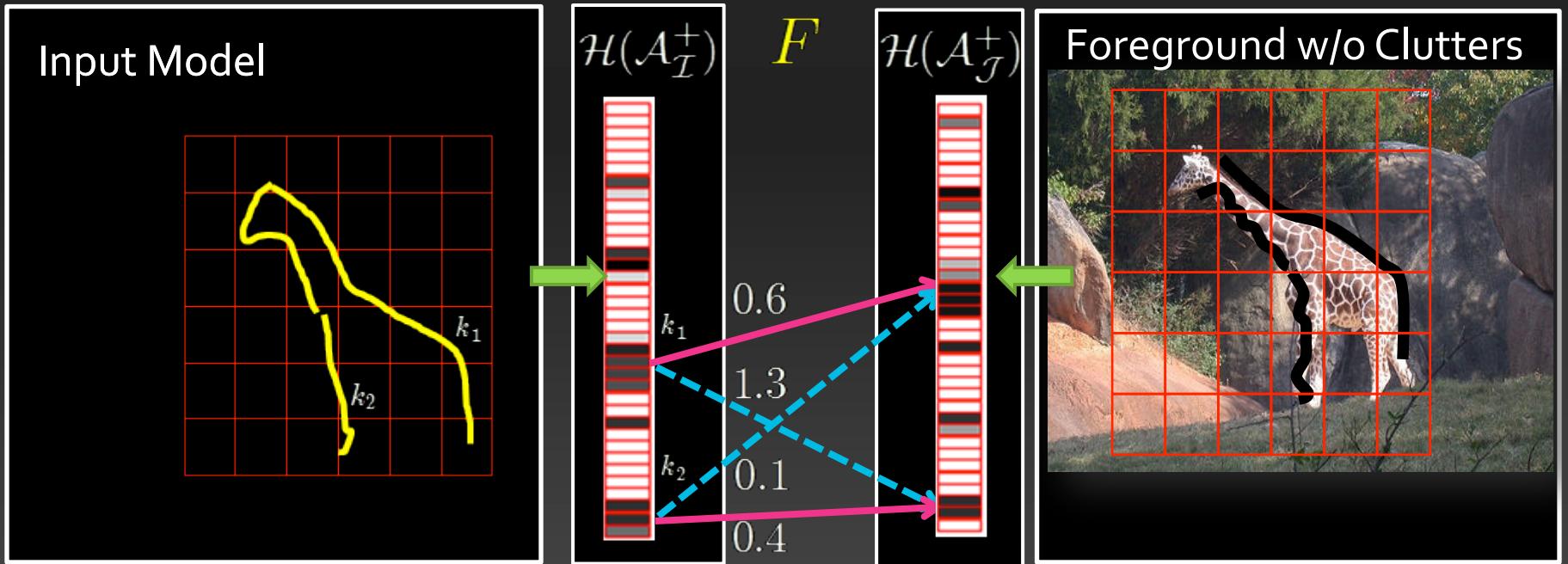
Histogram Flow



How to differentiate
two flows?



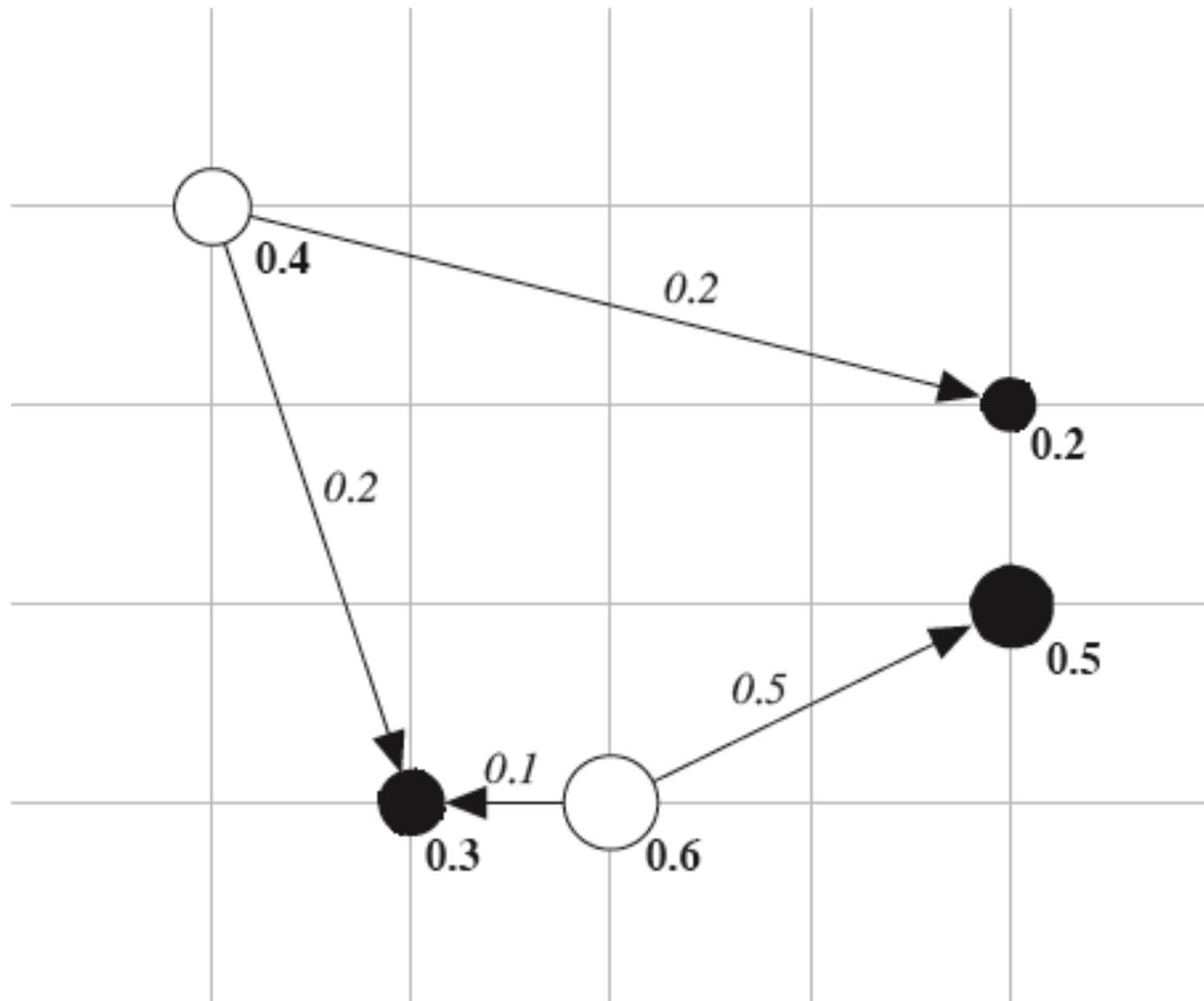
Histogram Flow Ground Distance Cost



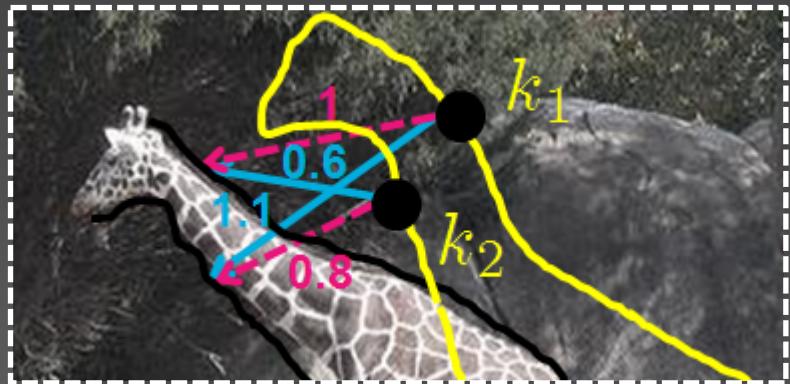
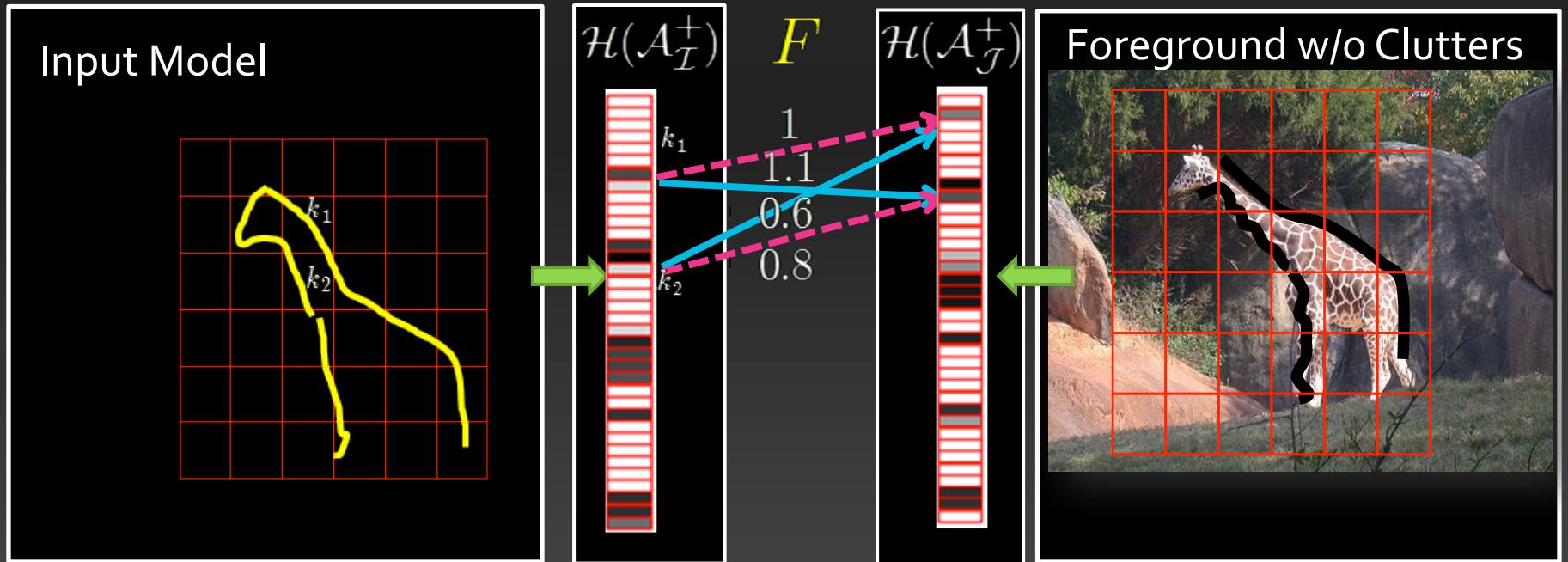
Earth Mover's Distance:

$$GD(F) = \sum_{k,l} F_{k,l} d_{k,l}$$

Linear programming for solving EMD

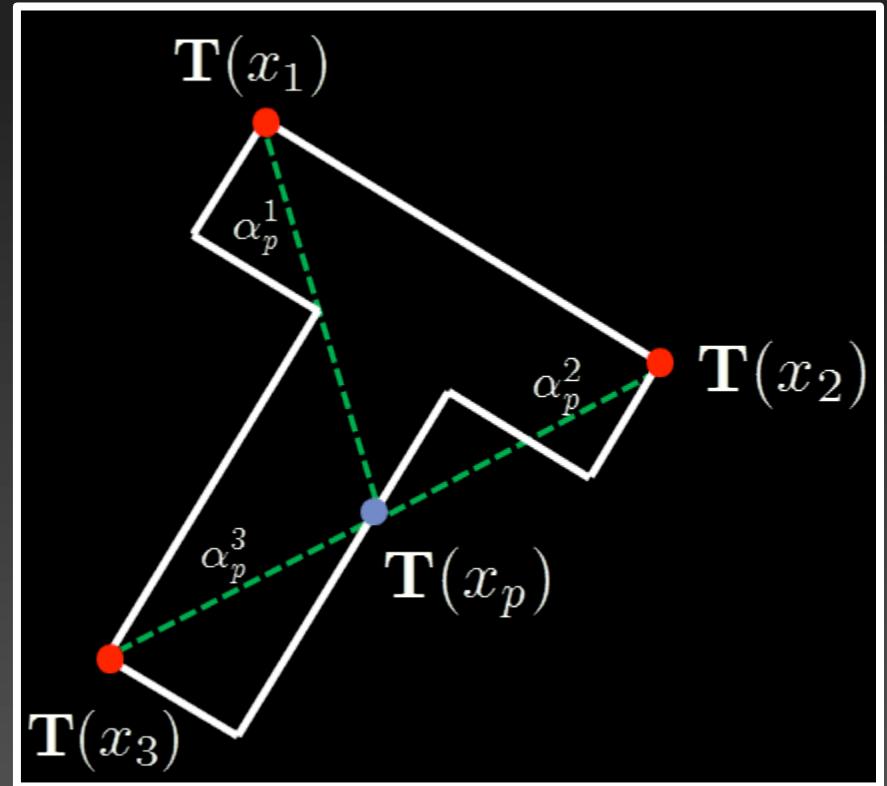
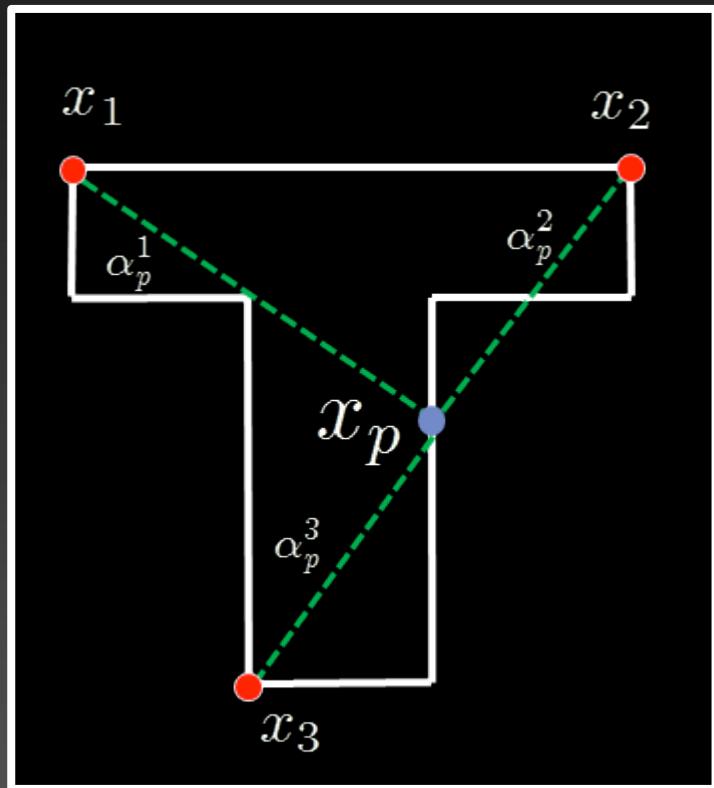


But, Ground Distance fails in rotation!



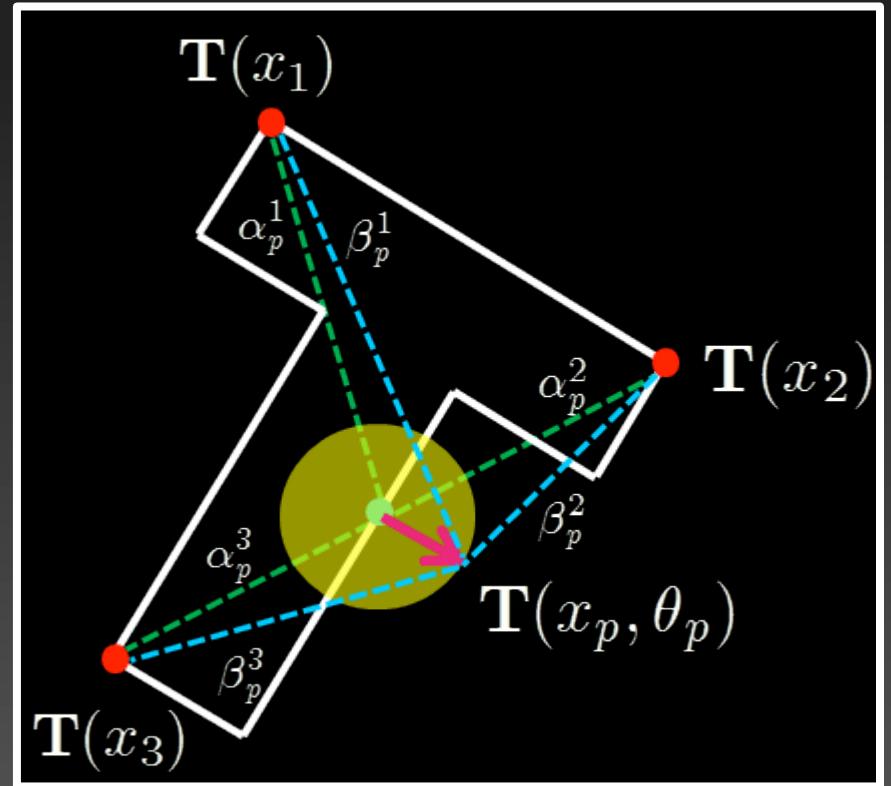
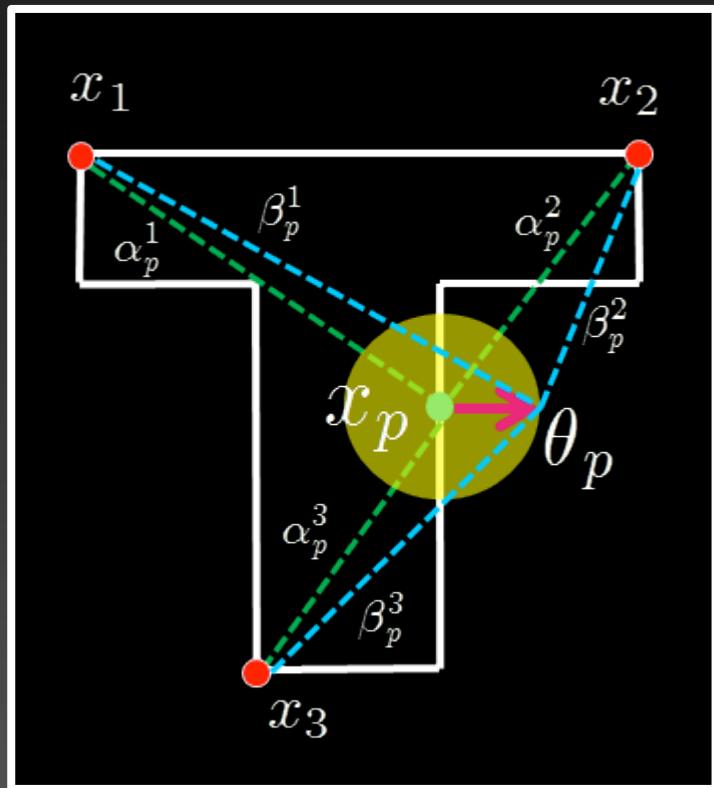
Minimizing ground distance fails when the object rotates.

Affine Constraint on Spatial Transformation



Affine Constraint 1: $T(x_p) = \sum_{i=1}^3 \alpha_i^p T(x_i)$

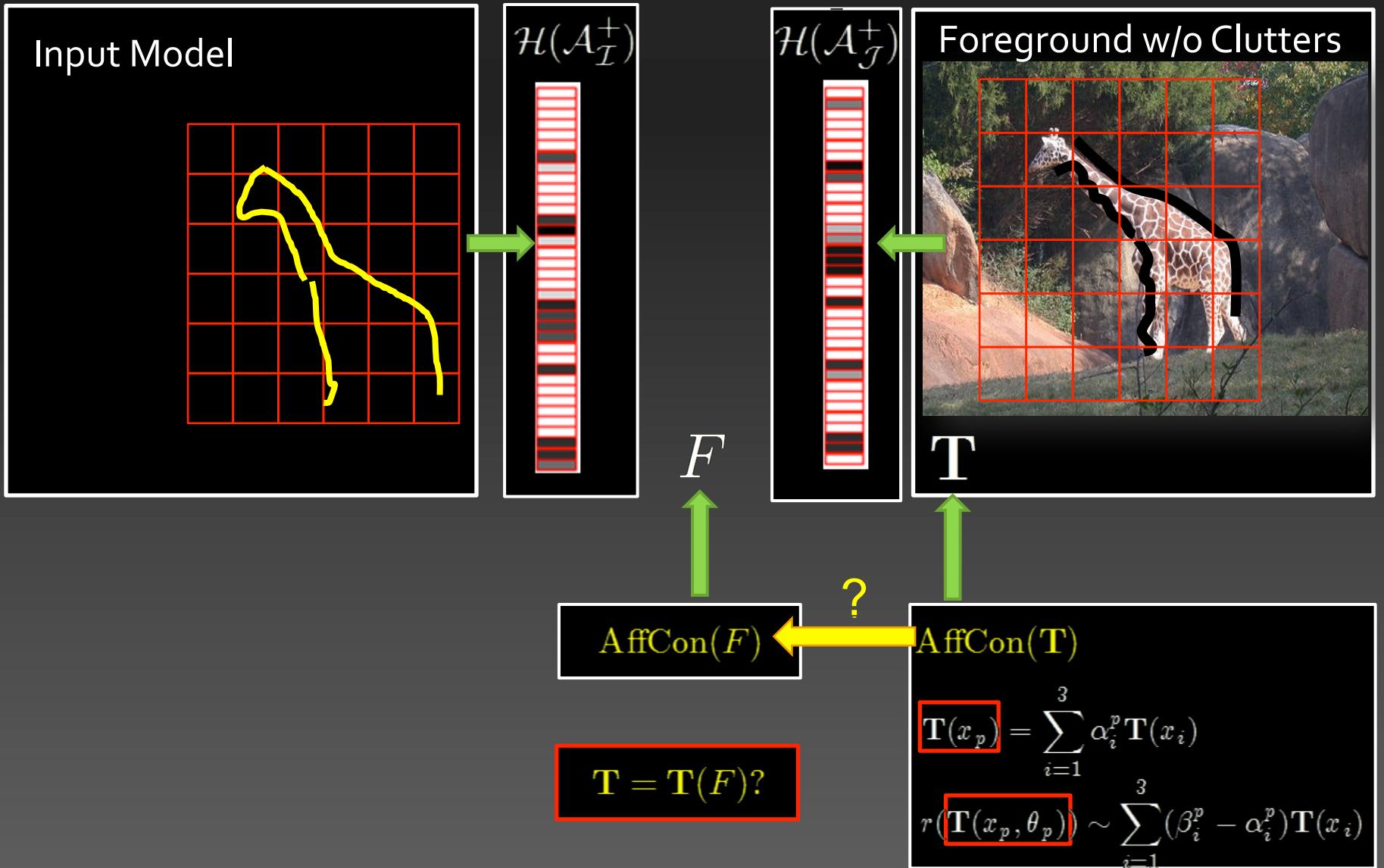
Affine Constraint on Spatial Transformation



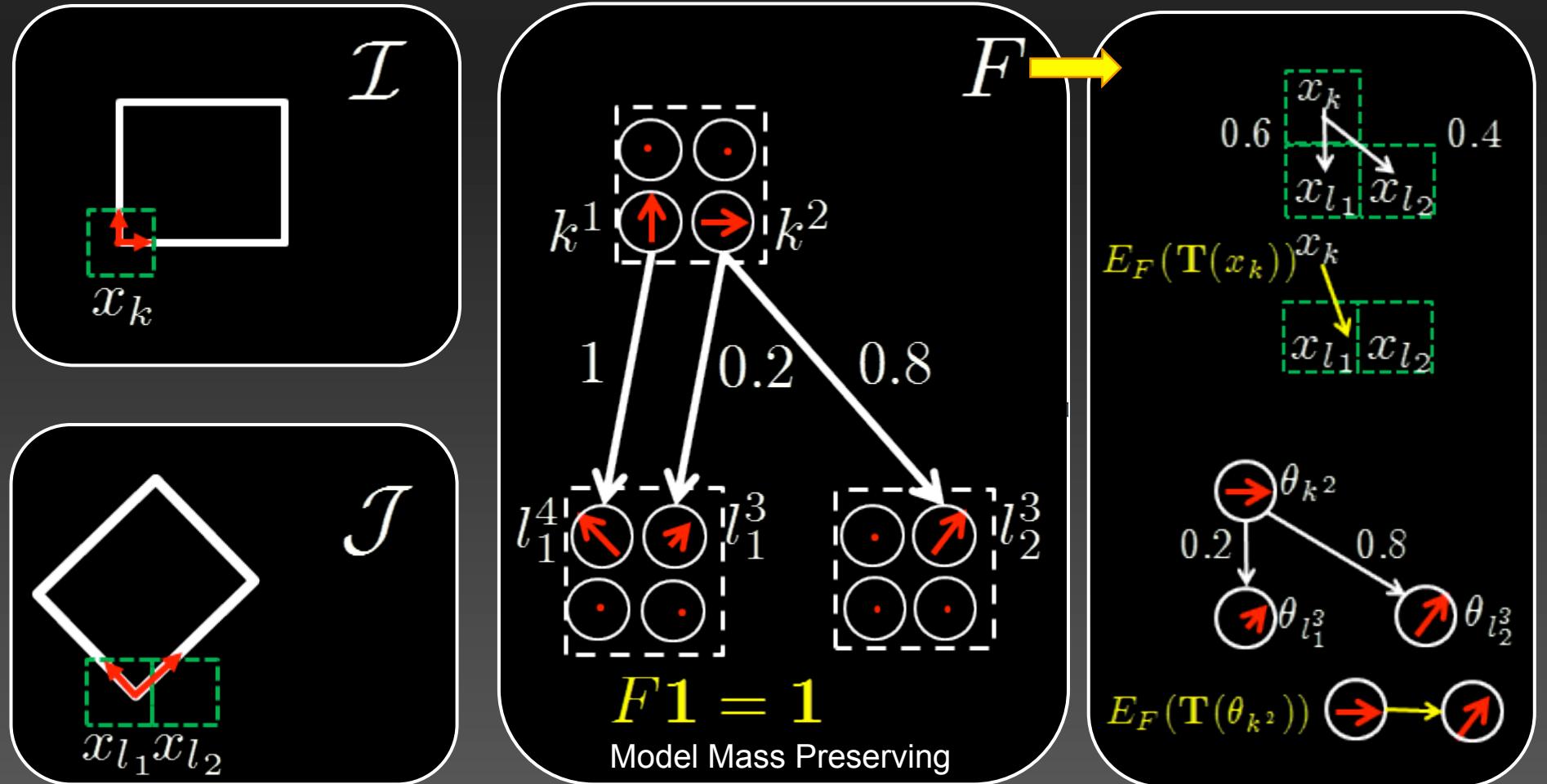
Affine Constraint 2: $r(\mathbf{T}(x_p, \theta_p)) \sim \sum_{i=1}^3 (\beta_i^p - \alpha_i^p) \mathbf{T}(x_i)$

$$r(\theta) = [\cos(\theta), \sin(\theta)]^\top$$

Affine Constraint on Histogram Flow



Spatial Transformation from Histogram Flow



F can be viewed as a probability encoding of T

Geometrical constraint 1

$$\mathbf{T}(x_z) = \sum_{i=1}^3 \alpha_i^z \mathbf{T}(x_i)$$



$$\text{AffCon}_{\mathbf{x}}(F, k) : ||E_F(\mathbf{T}(p_k)) - \sum_{i=1}^k \alpha_i^k E_F(\mathbf{T}(p_{k_i}))||_1$$

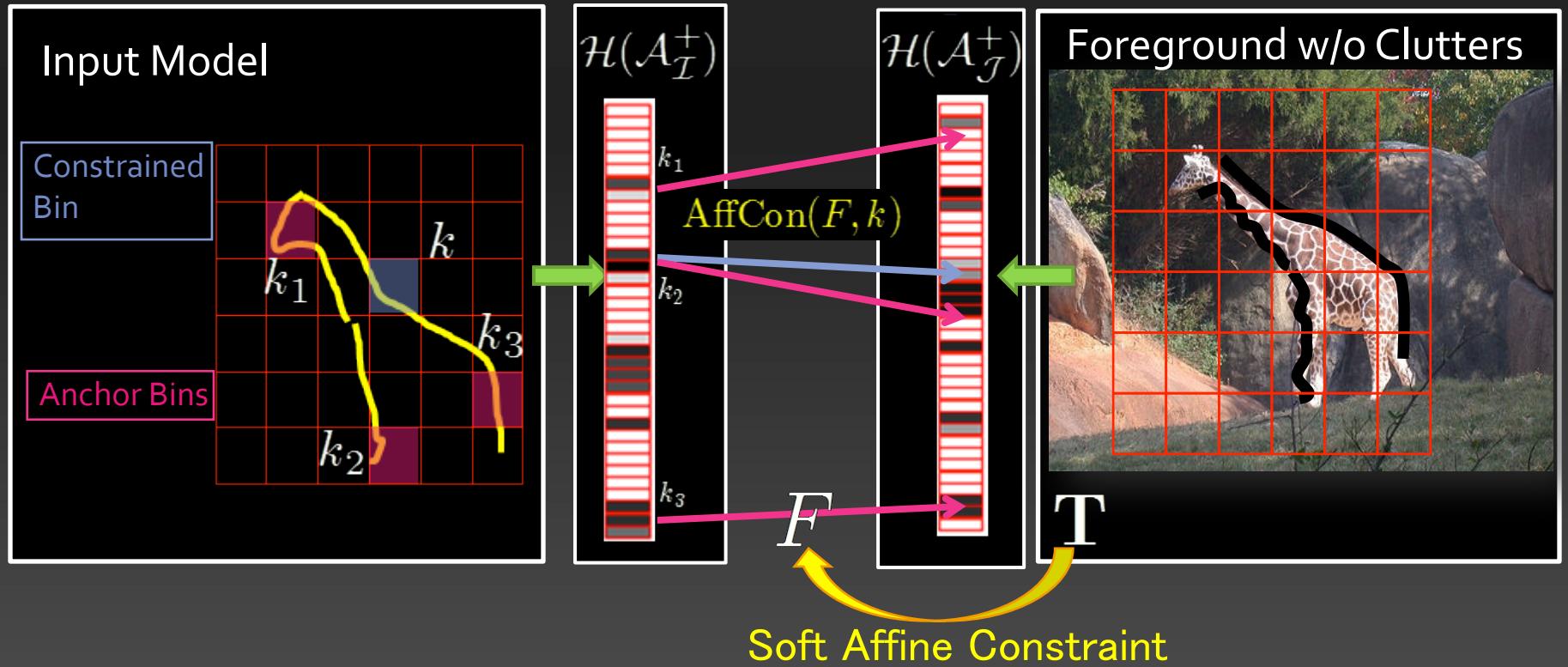
Geometrical constraint 2

$$\mathbf{T}([\cos(\theta_z) \sin(\theta_z)]^T) \sim \sum_{i=1}^3 (\beta_i^z - \alpha_i^z) \mathbf{T}(x_i),$$



$$E_F(\mathbf{T}([\cos(\theta_k) \sin(\theta_k)]^T)) \sim \sum_{i=1}^3 (\beta_i^k - \alpha_i^k) E_F(\mathbf{T}(p_{k_i})).$$

Constrained Attribute Histogram Flow

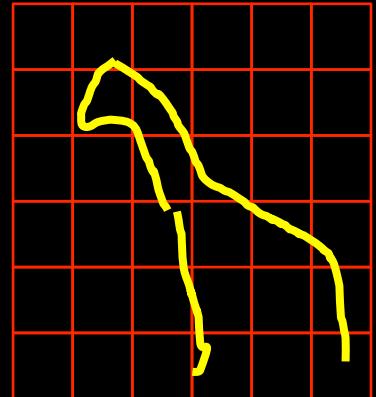


$$\text{AffCon}_x(F, k) : |E_F(\mathbf{T}(x_k)) - \sum_{i=1}^3 \alpha_i^k E_F(\mathbf{T}(x_{k_i}))|_1$$

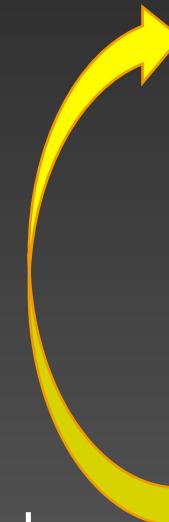
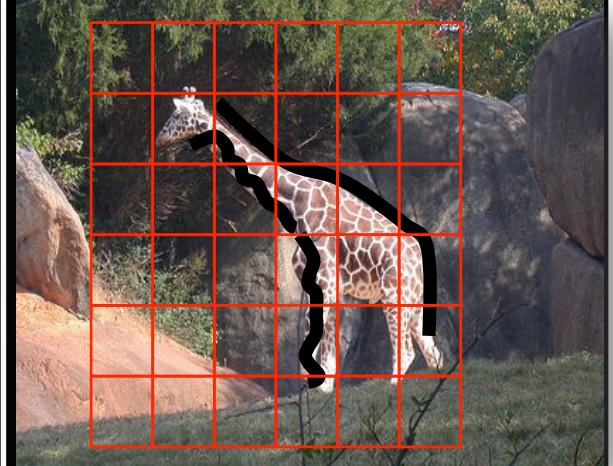
$$\text{AffCon}(F, k) = \text{AffCon}_\theta(F, k) + \min_s |s \cdot r(E_F(\mathbf{T}(\theta_k))) - \sum_{i=1}^3 (\beta_i^k - \alpha_i^k) E_F(\mathbf{T}(x_{k_i}))|_1$$

Match with Selected Region

Input "Clean" Model

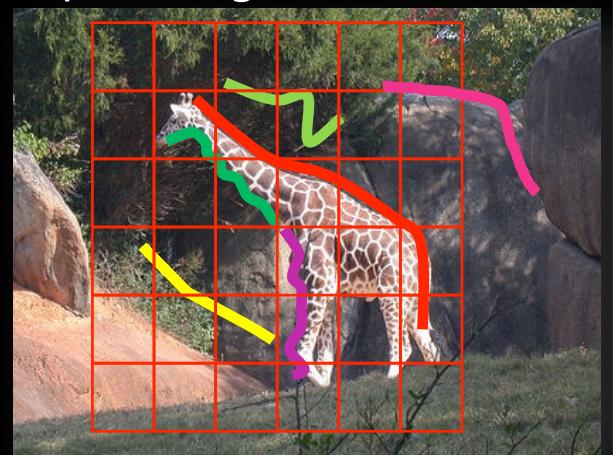


Foreground w/o Clutters



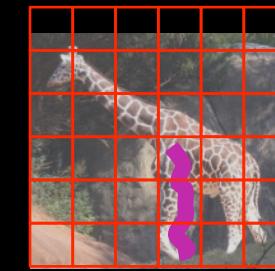
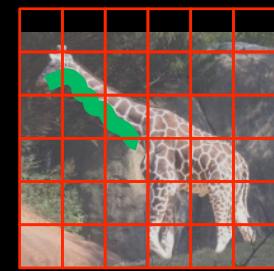
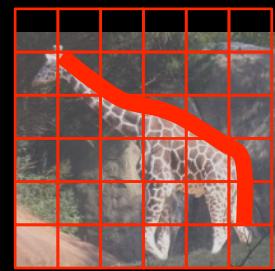
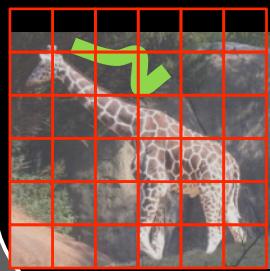
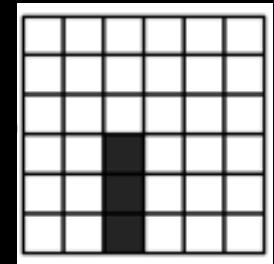
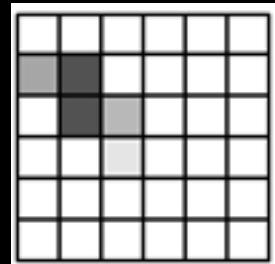
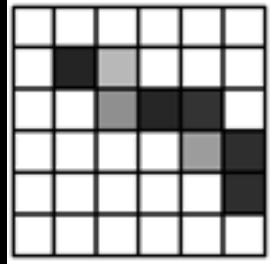
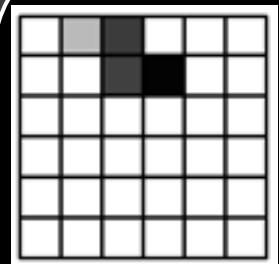
Q: How to efficiently select foreground region to avoid "cherry-picking"?

Input Image w/ Clutters

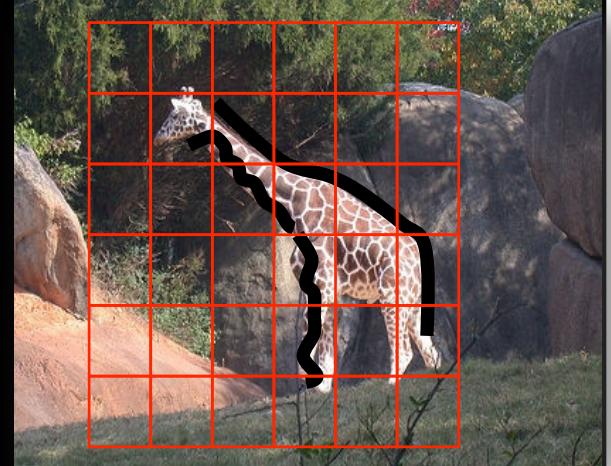


Match with Selected Region

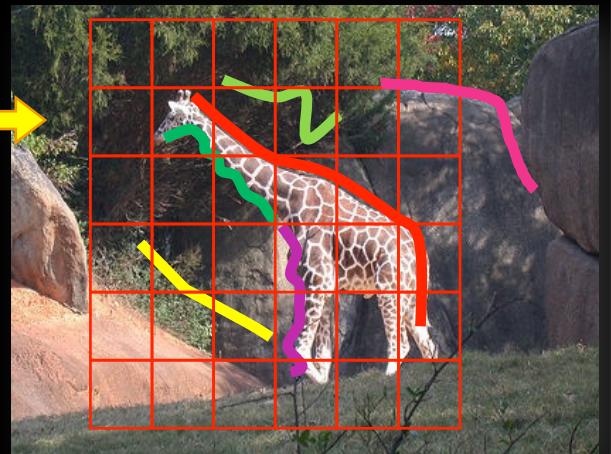
“Shape Basis”



Foreground w/o Clutters

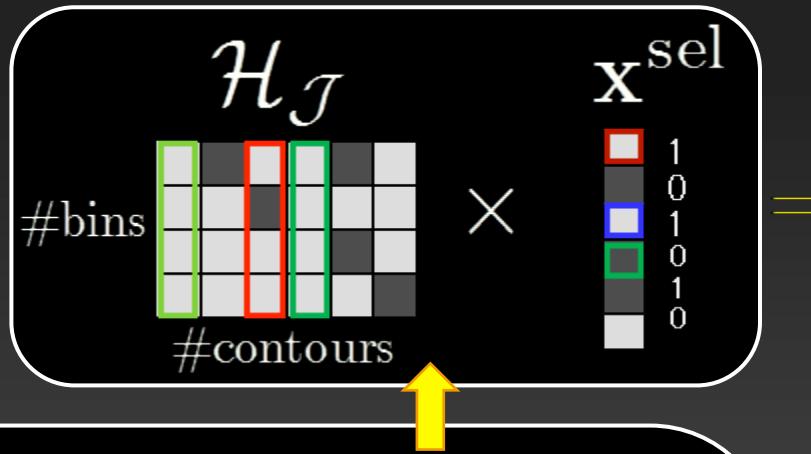


Input Image w/ Clutters

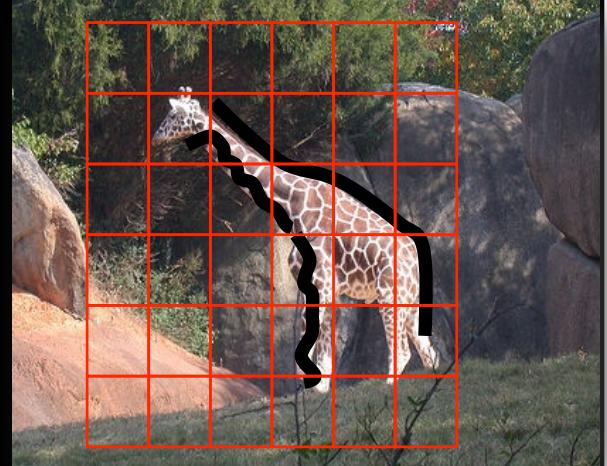


Match with Selected Region

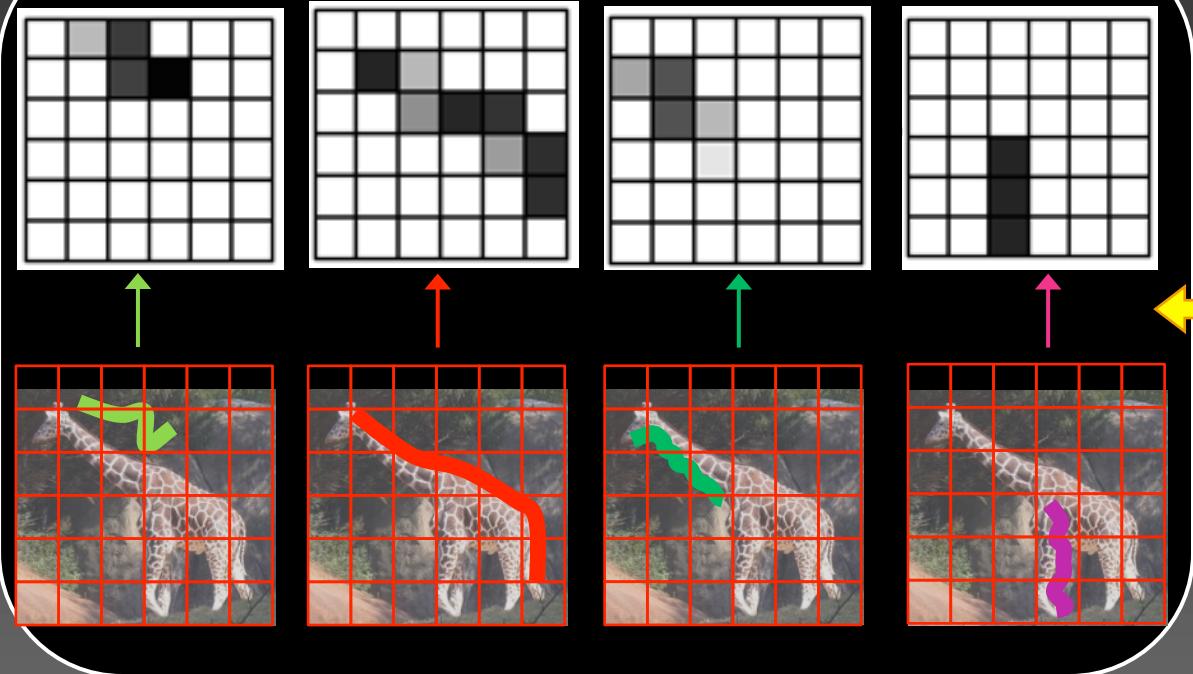
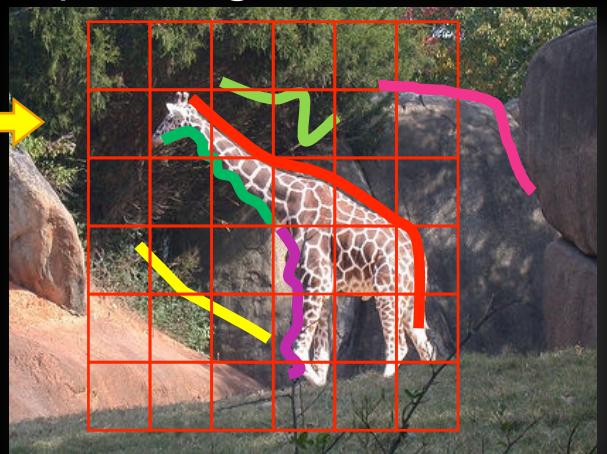
Contour Selection



Foreground w/o Clutters



Input Image w/ Clutters



Application: Simultaneous Detection and Alignment

Input

Shape Model

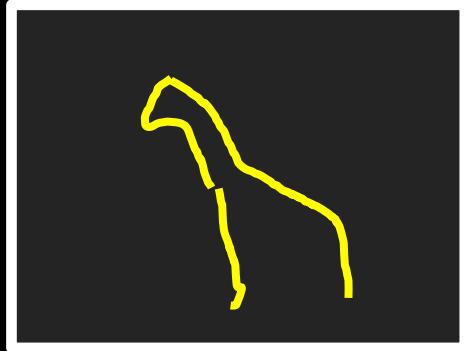
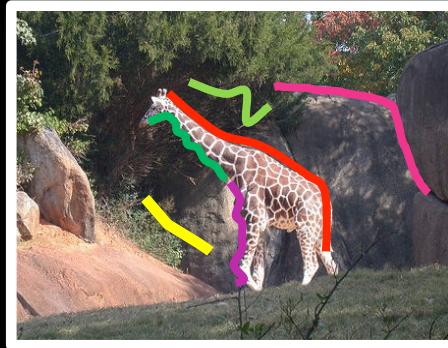
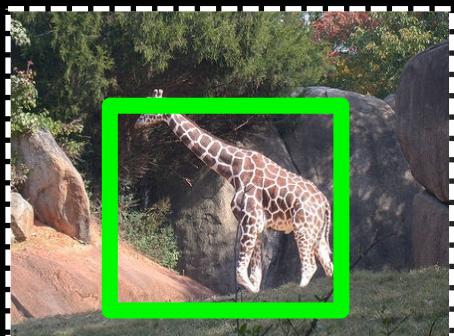


Image with contours, or segments



Output

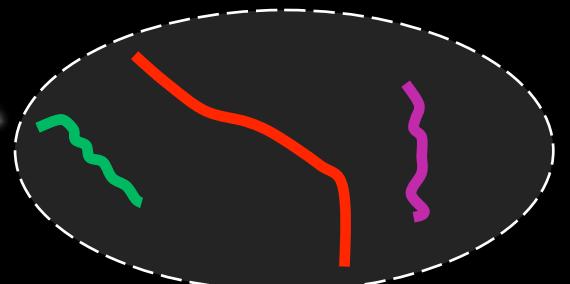
Detection



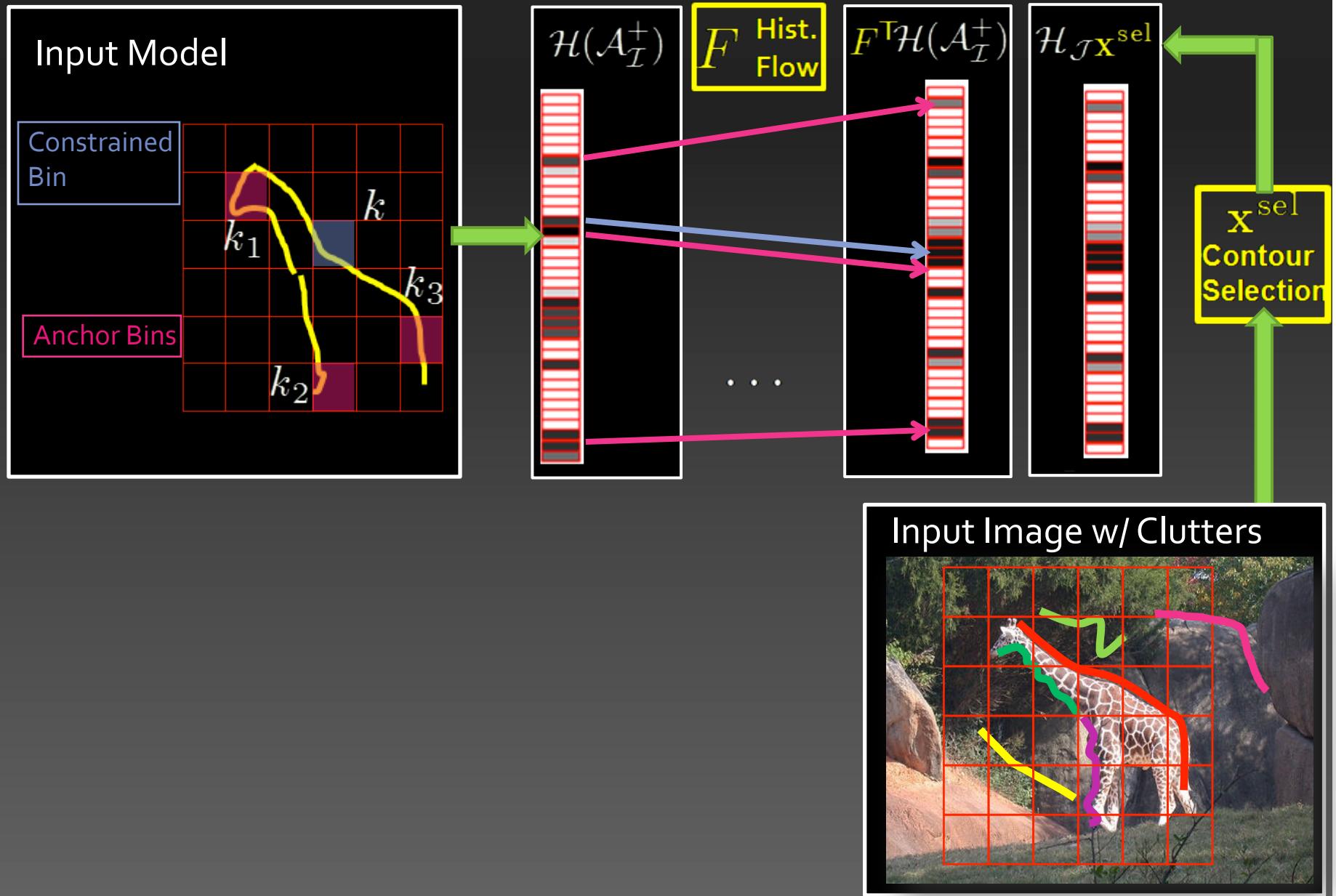
Shape Alignment



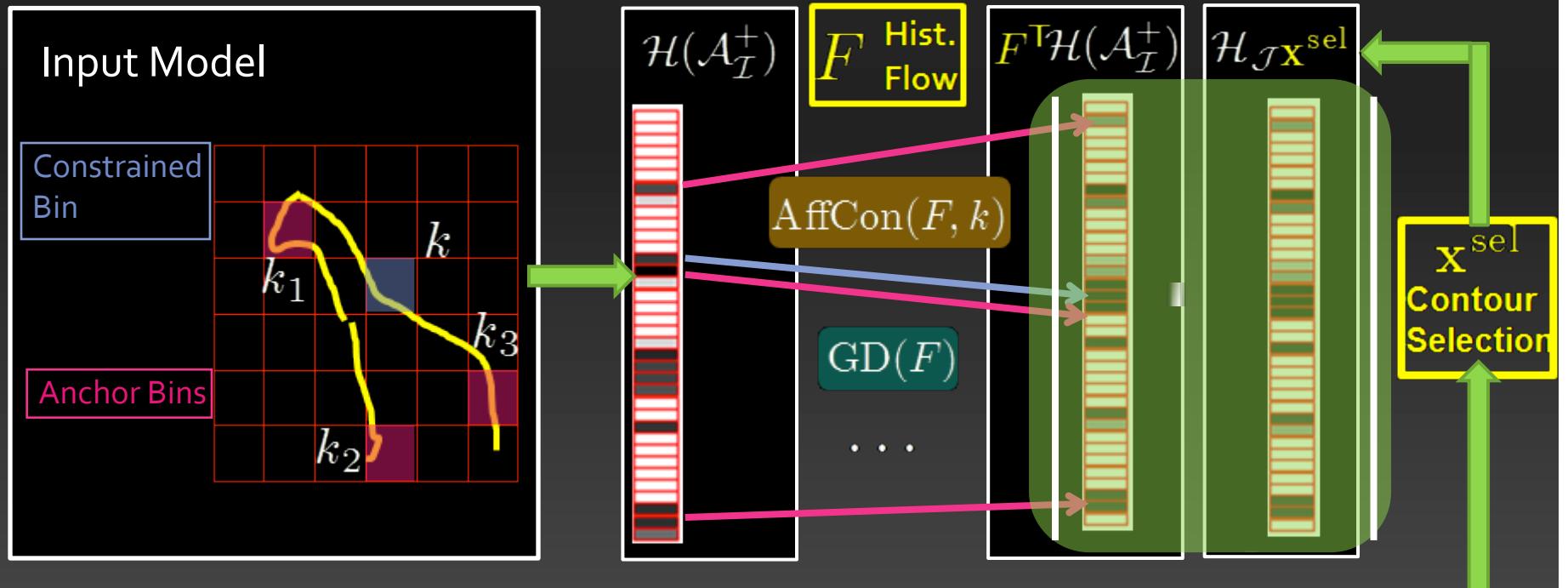
Foreground
Segmentation



Algorithm Overview



Algorithm Overview



Constrained Histogram Flow

$$\begin{aligned}
 & \min_{F, \mathbf{x}^{\text{sel}}} \quad \text{GD}(F) + \lambda \sum_{k=1}^m \text{AffCon}(F, k) \\
 & \quad + \gamma \|F^\top \mathcal{H}(\mathcal{A}_I^+) - \mathcal{H}_{\mathcal{J}} \mathbf{x}^{\text{sel}}\|_1 \\
 \text{s.t.} \quad & F \mathbf{1} = \mathbf{1}, \quad F \geq 0, \quad \mathbf{x}^{\text{sel}} \in [0, 1]^{|\mathcal{C}|}
 \end{aligned}$$

Input Image w/ Clutters

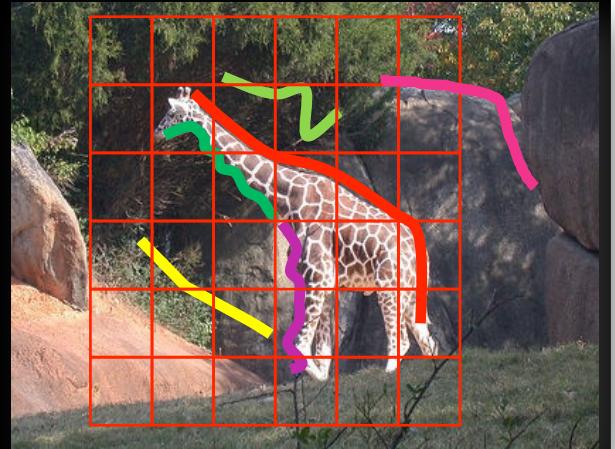
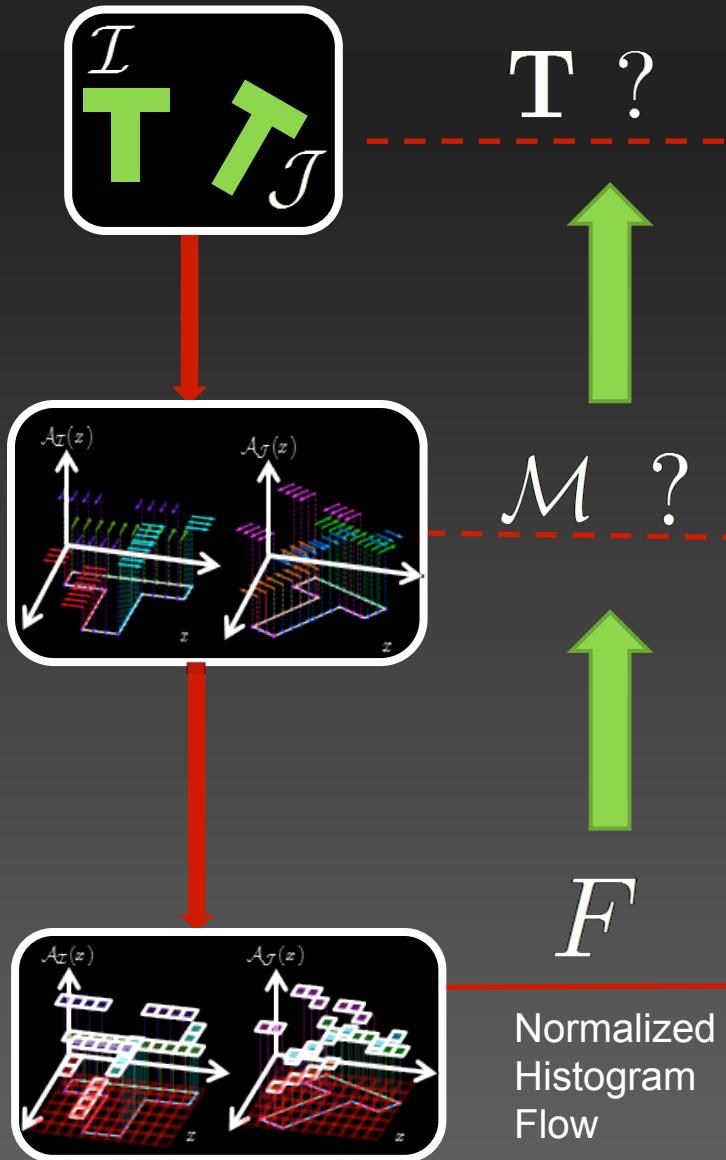


Image Deformation as Attribute Flow



Attribute Optical Flow

$$\min_{\mathbf{T}} \int_{\mathbf{R}^2} \|\mathcal{A}_{\mathcal{I}}(x) - \mathcal{A}_{\mathcal{J}}(\mathbf{T}(x))\|_p dx$$

Attribute Flow

$$\min_{\mathcal{M}} \int_{\mathbf{R}^{n+2}} |\delta(y - \mathcal{M}(\mathcal{A}_{\mathcal{I}}^+)) - \delta(y - \mathcal{A}_{\mathcal{J}}^+)|_1 dy$$

Attribute Histogram Flow (LP)

$$\begin{aligned} \min_F \quad & |F^T \mathcal{H}(\mathcal{A}_{\mathcal{I}}^+) - \mathcal{H}(\mathcal{A}_{\mathcal{J}}^+)|_1 \\ \text{s.t.} \quad & F\mathbf{1} = \mathbf{1} \quad F \geq 0 \\ & \text{Model Mass Preserving} \end{aligned}$$

Challenges

Q1: How to compute the spatial transformation efficiently without explicit search?

A : Attribute Histogram Flow.

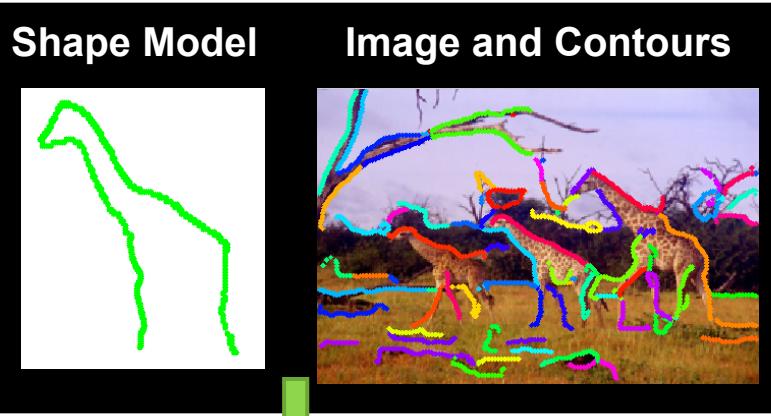
Q2: How to constrain the computation such that it leads to valid spatial transformation?

A : Soft Affine Constraint on expectation.

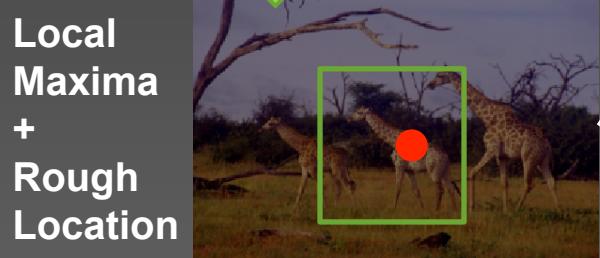
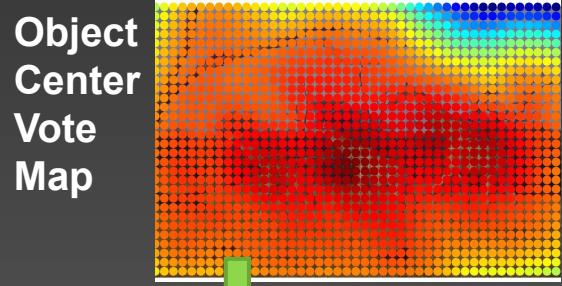
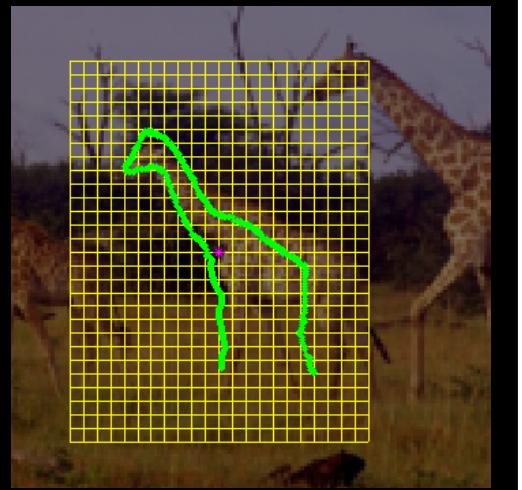
Q3: How to deal with clutter environment?

A : Contour Selection.

Detection Example

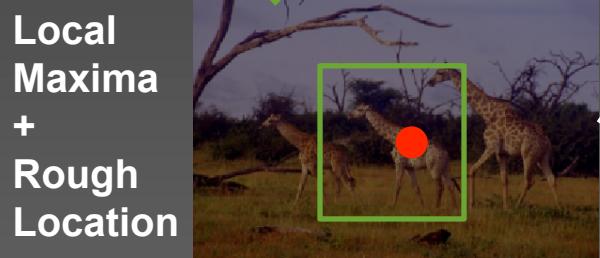
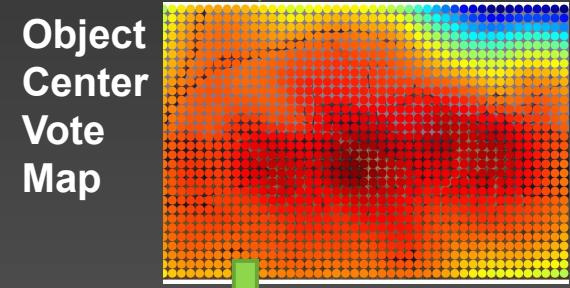
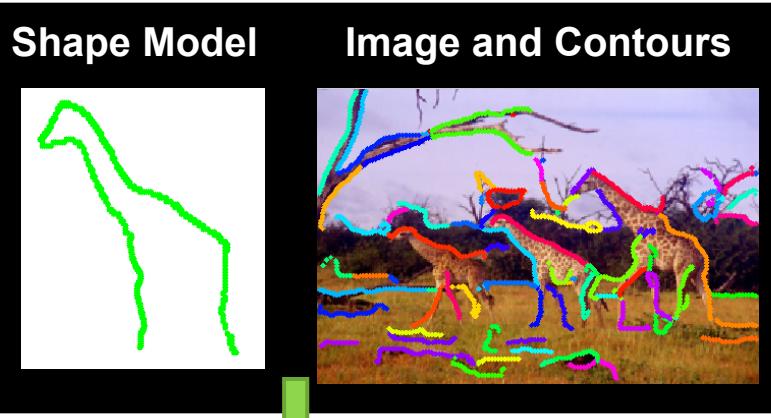


**Constrained
Histogram
Matching**

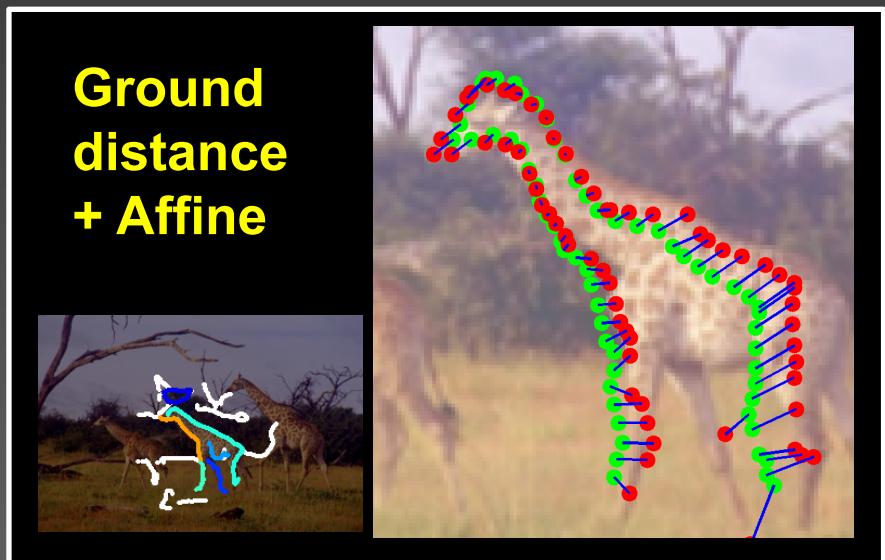


**For each
detection
candidate**

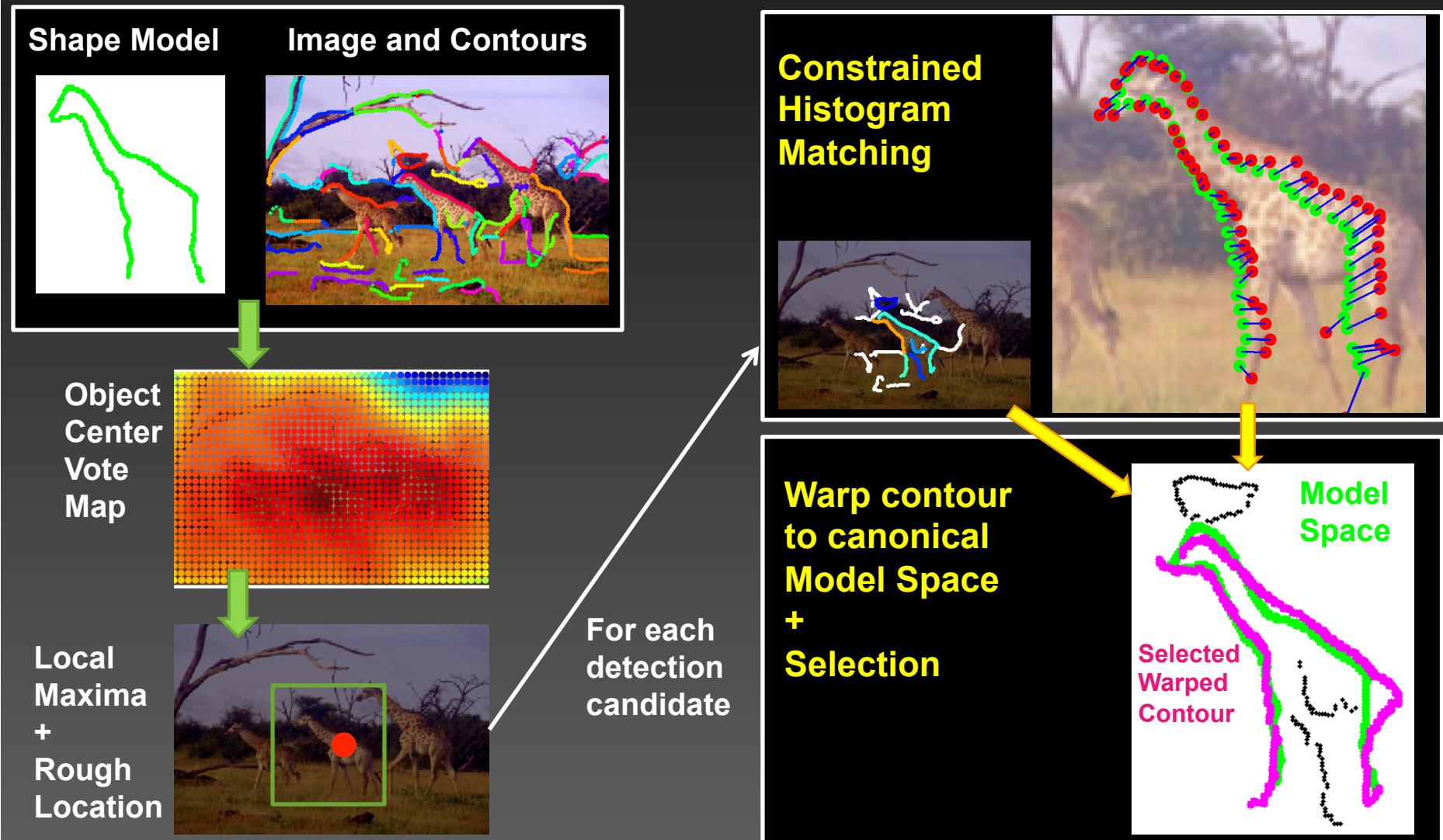
Detection Example



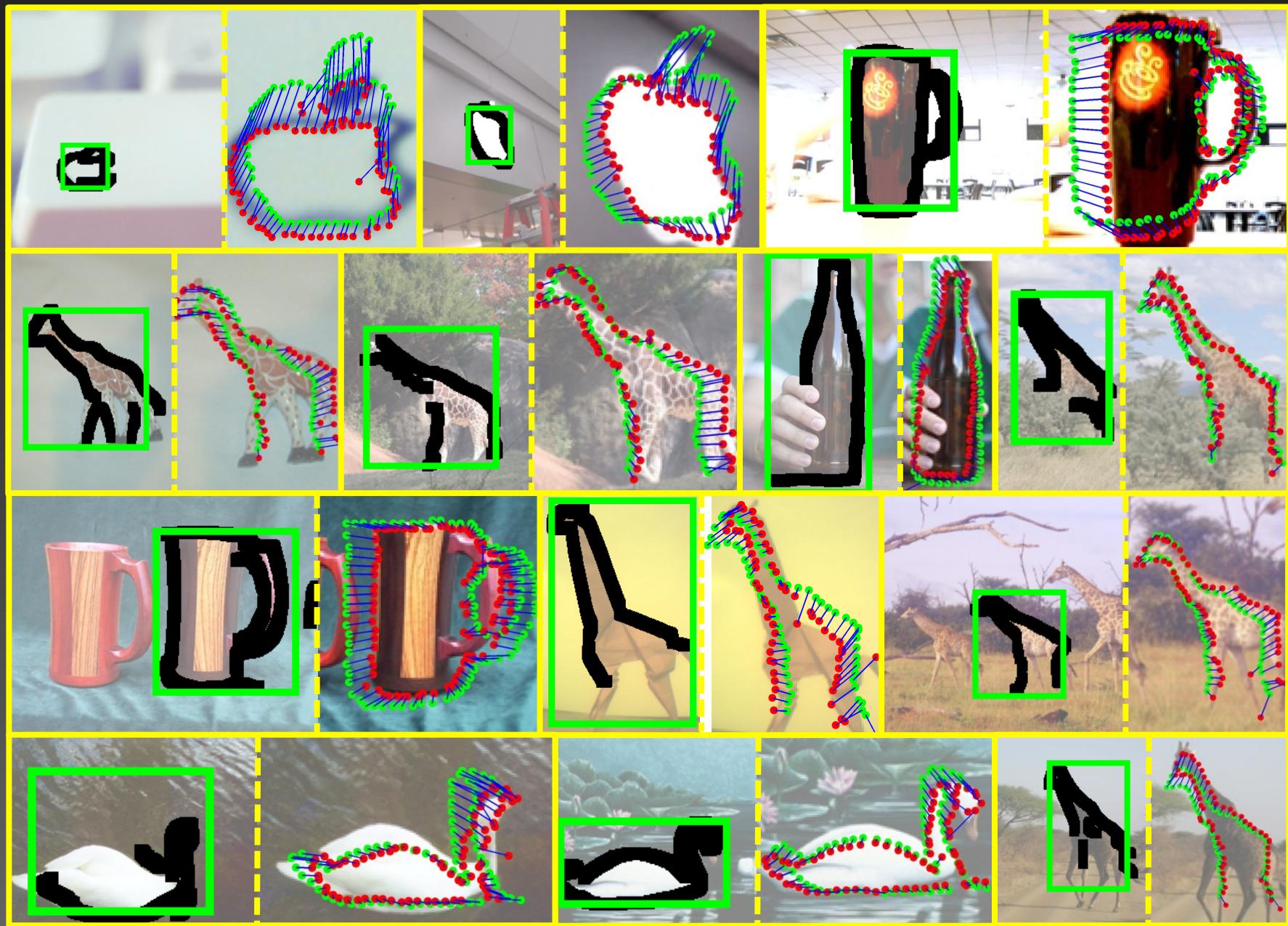
For each
detection
candidate



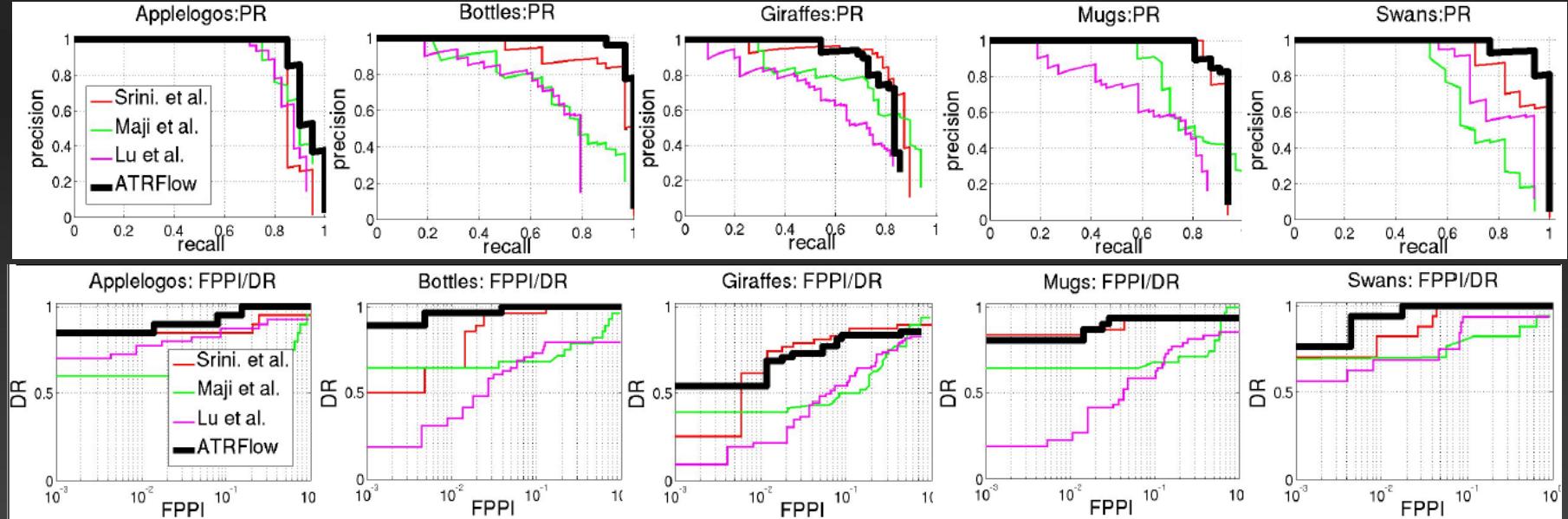
Discrimination in Canonical Space



Results on ETHZ



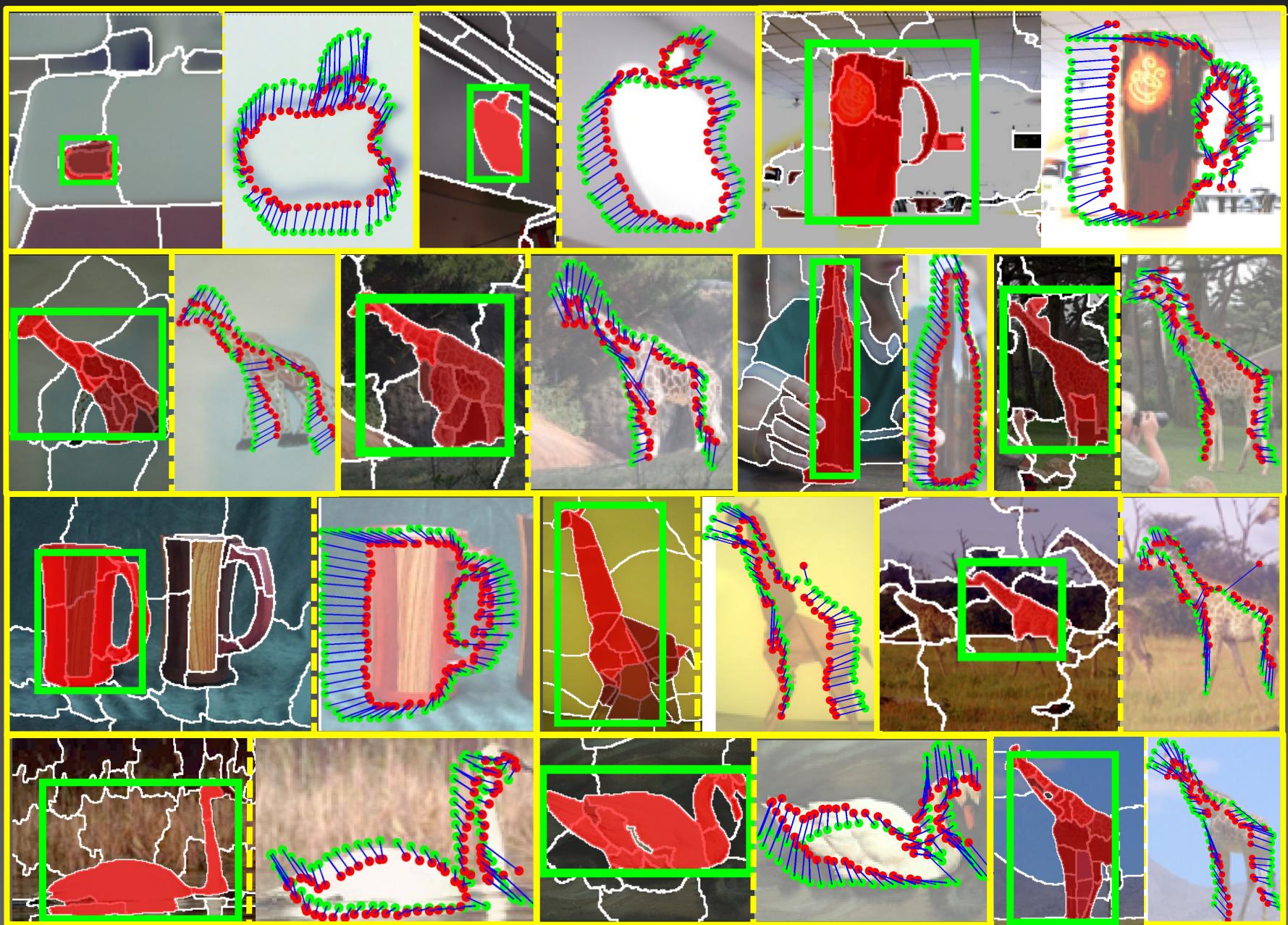
Quantitative Evaluation



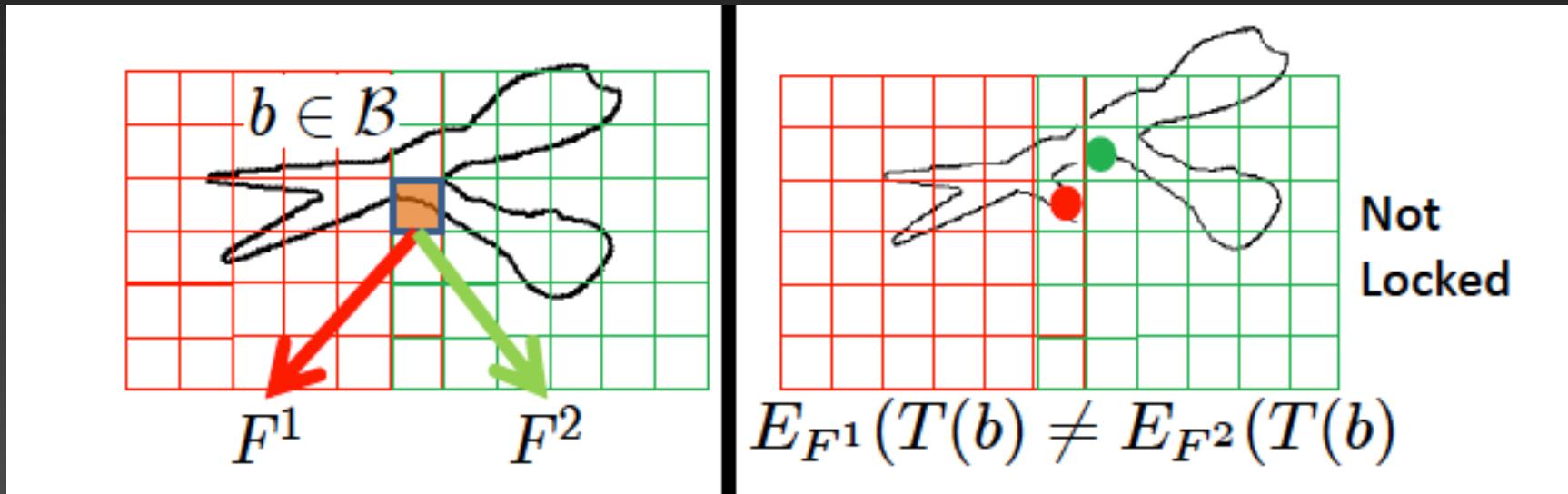
Average
Precision

	Applelogos	Bottles	Giraffes	Mugs	Swans	Mean
Attribute Flow	0.930	0.977	0.783	0.895	0.972	0.911
Ma et al. [6]	0.881	0.920	0.756	0.868	0.959	0.877
Srini. et al. [10]	0.845	0.916	0.787	0.888	0.922	0.872
Maji et al. [7]	0.869	0.724	0.742	0.806	0.716	0.771
Lu et al. [5]	0.844	0.641	0.617	0.643	0.798	0.709
Toshev et al. [§] [12]	0.983	0.936	0.713	0.718	0.973	0.865

Results on ETHZ using Segments

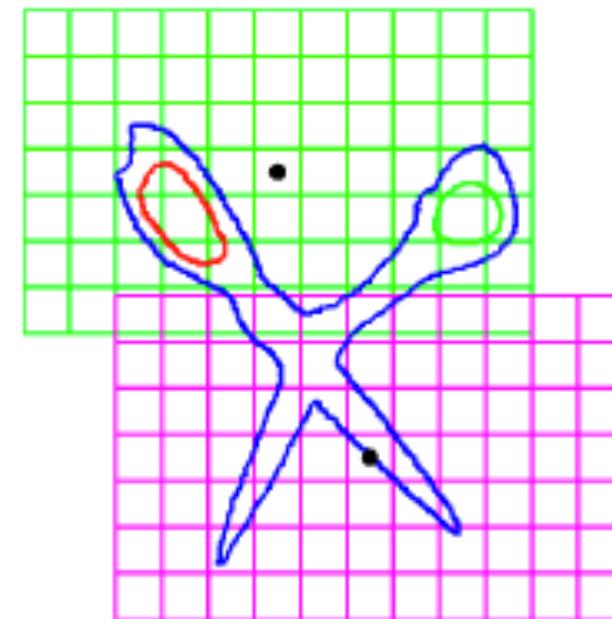


Piecewise affine Attribute Flow:





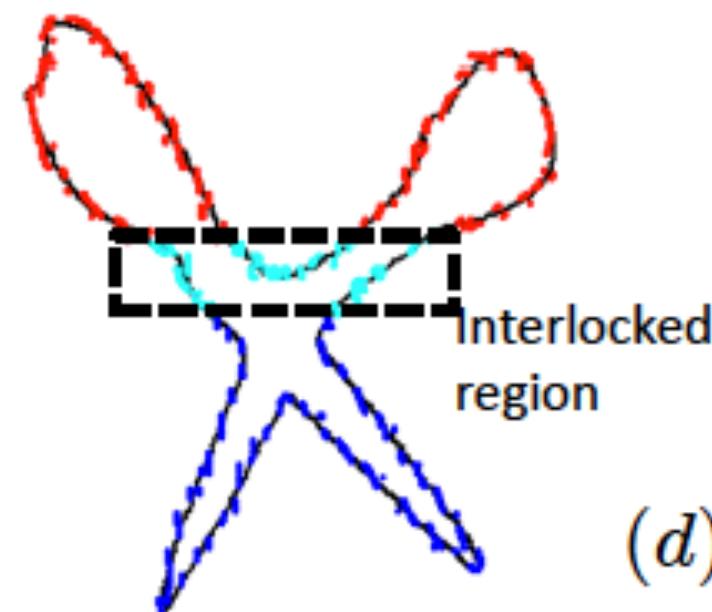
(a)



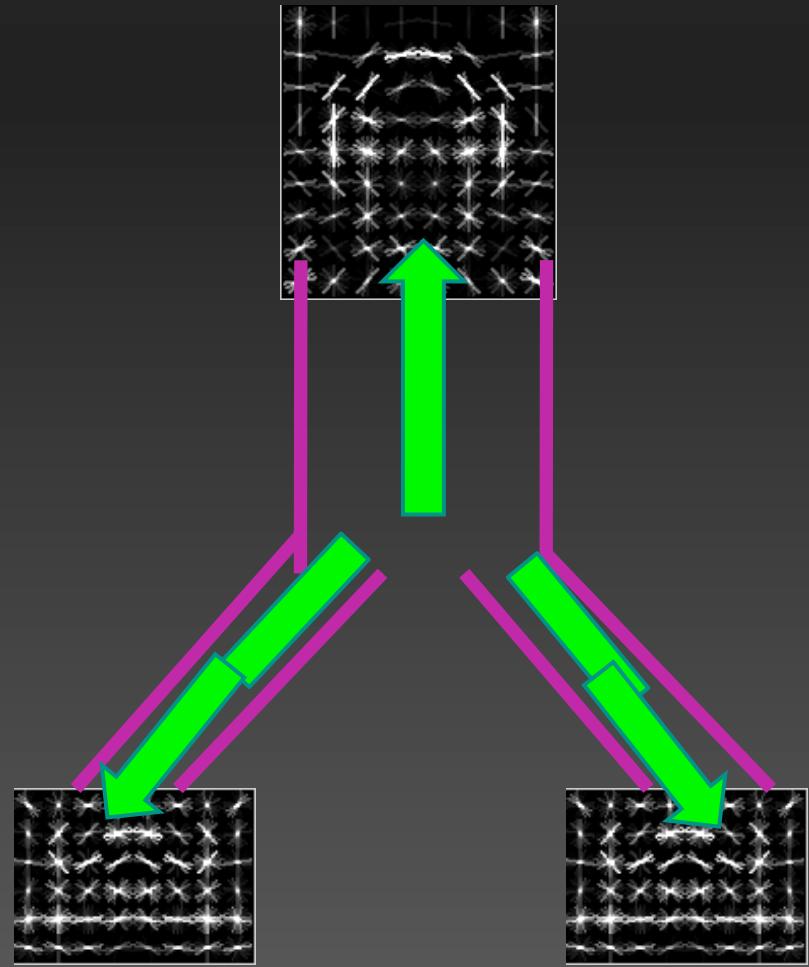
(b)

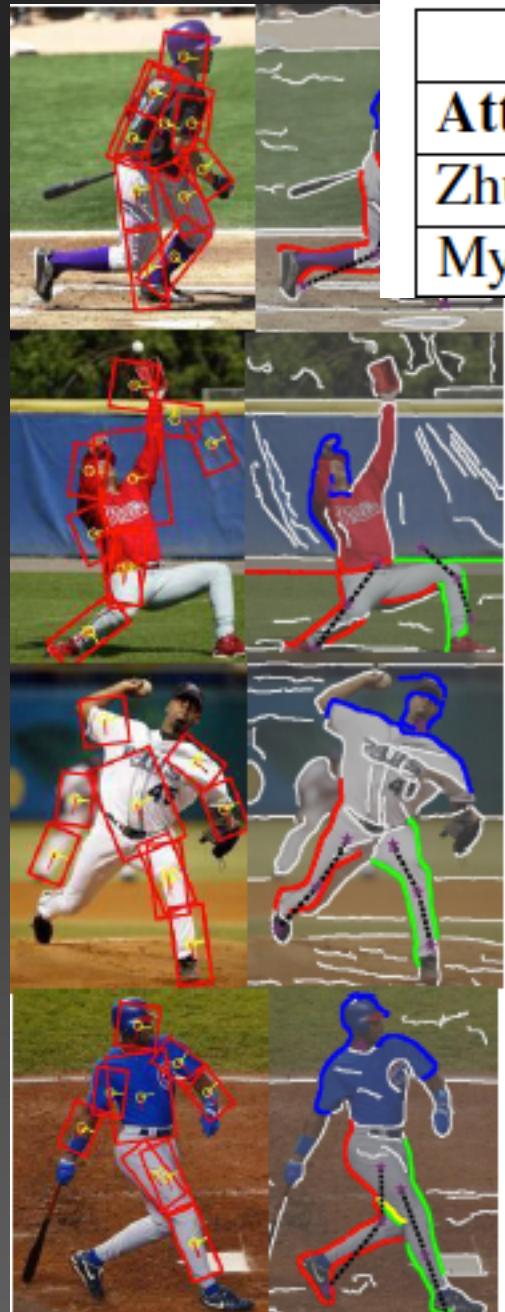


(c)



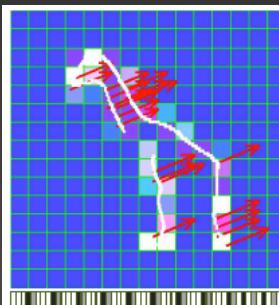
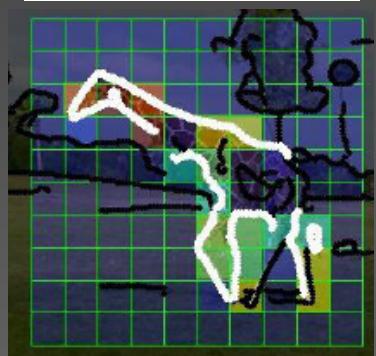
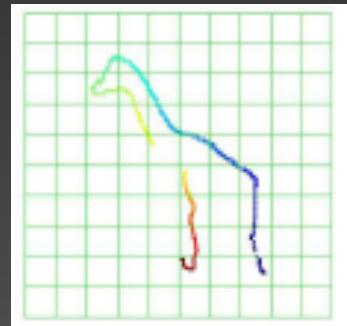
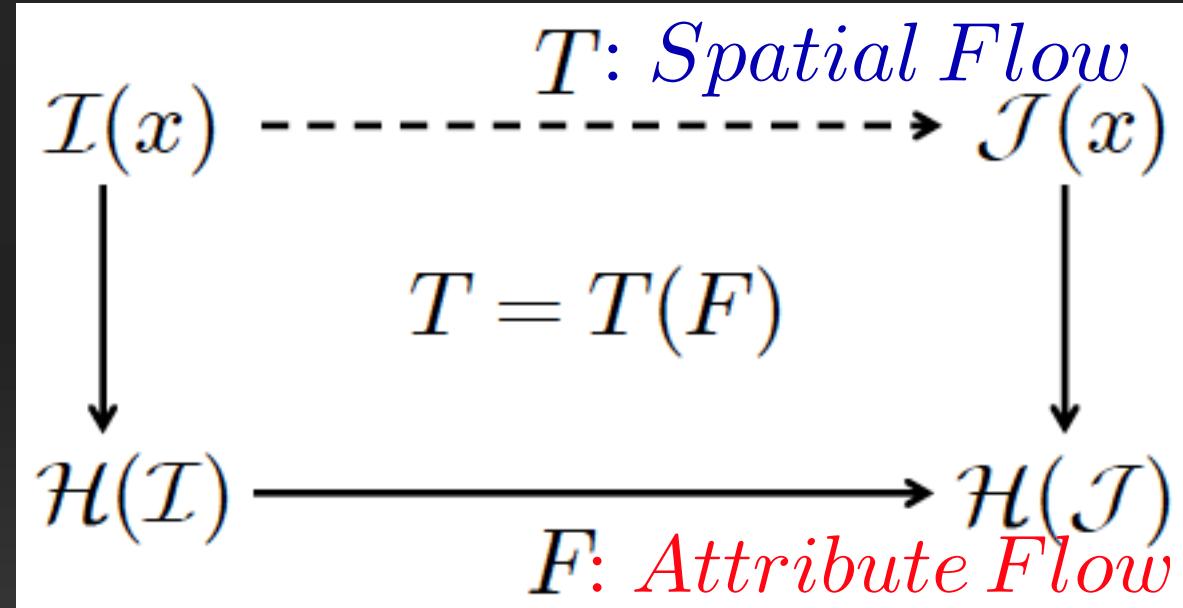
(d)



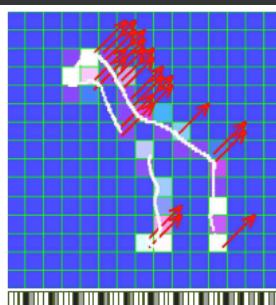


	Head	Hips	Knees	Ankles	Mean
Attribute Flow	12	27	24	27	26.5
Zhu et al. [15]	N/A	N/A	N/A	N/A	30
Mykhaylo et al. [1]	24	36.5	43	71	47

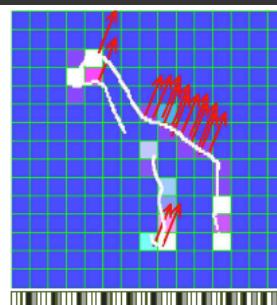




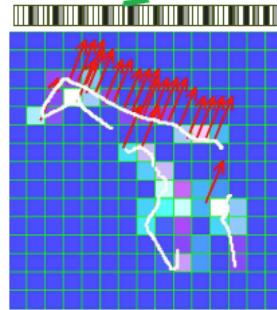
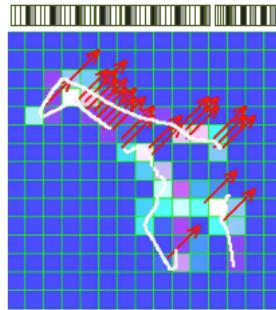
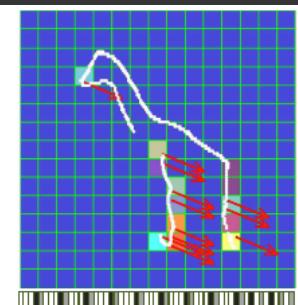
Flow across
Spatial Bins



Flow across
Orientation Bins



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