

Image Filtering (transformation)

그림 위치 옮김

Image_filtering.py



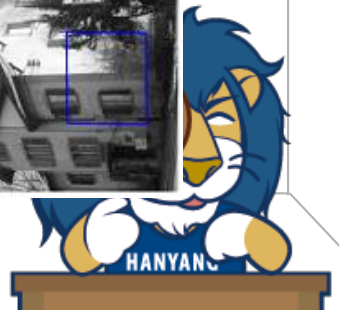
Application: Matching

- Affine template matching

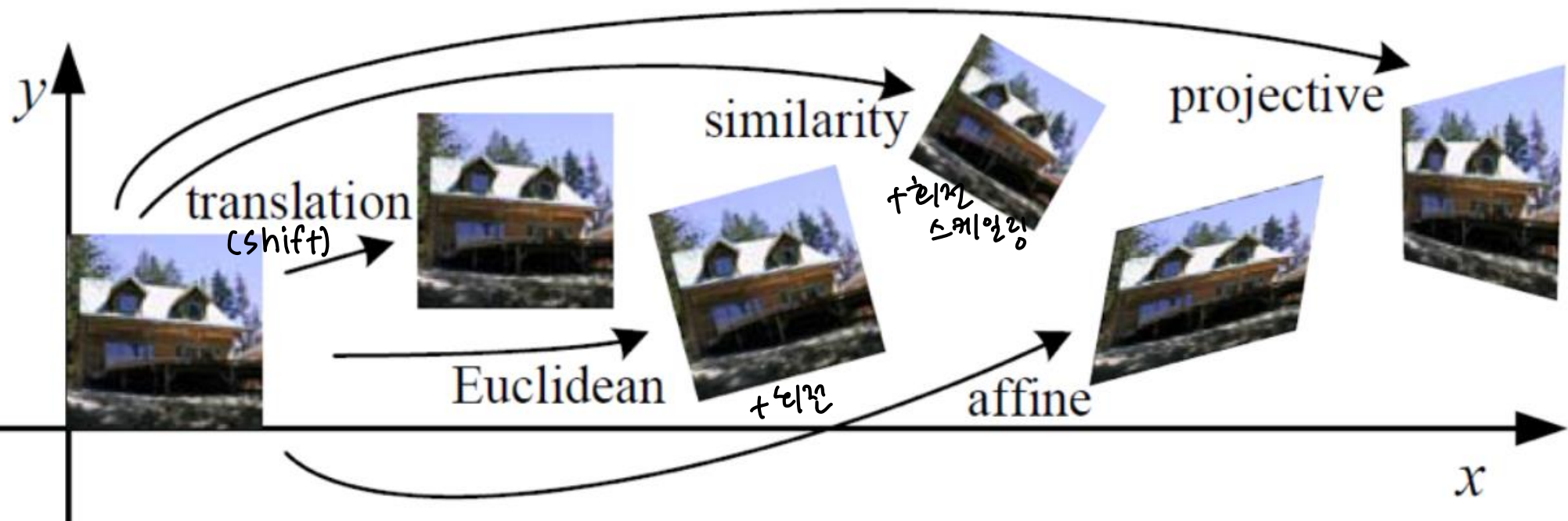
같은 그림. 촬영 시점 다름



Affine template matching results, Koreman et al. CVPR'13



2D Transformations



Application: Stitching

파노라마 이미지를 만드는 데 필요한 기술

- Basic tool for image registration and alignment

overlap 가장 높게
→ align
→ 붙여주기
→ 파노라마 이미지

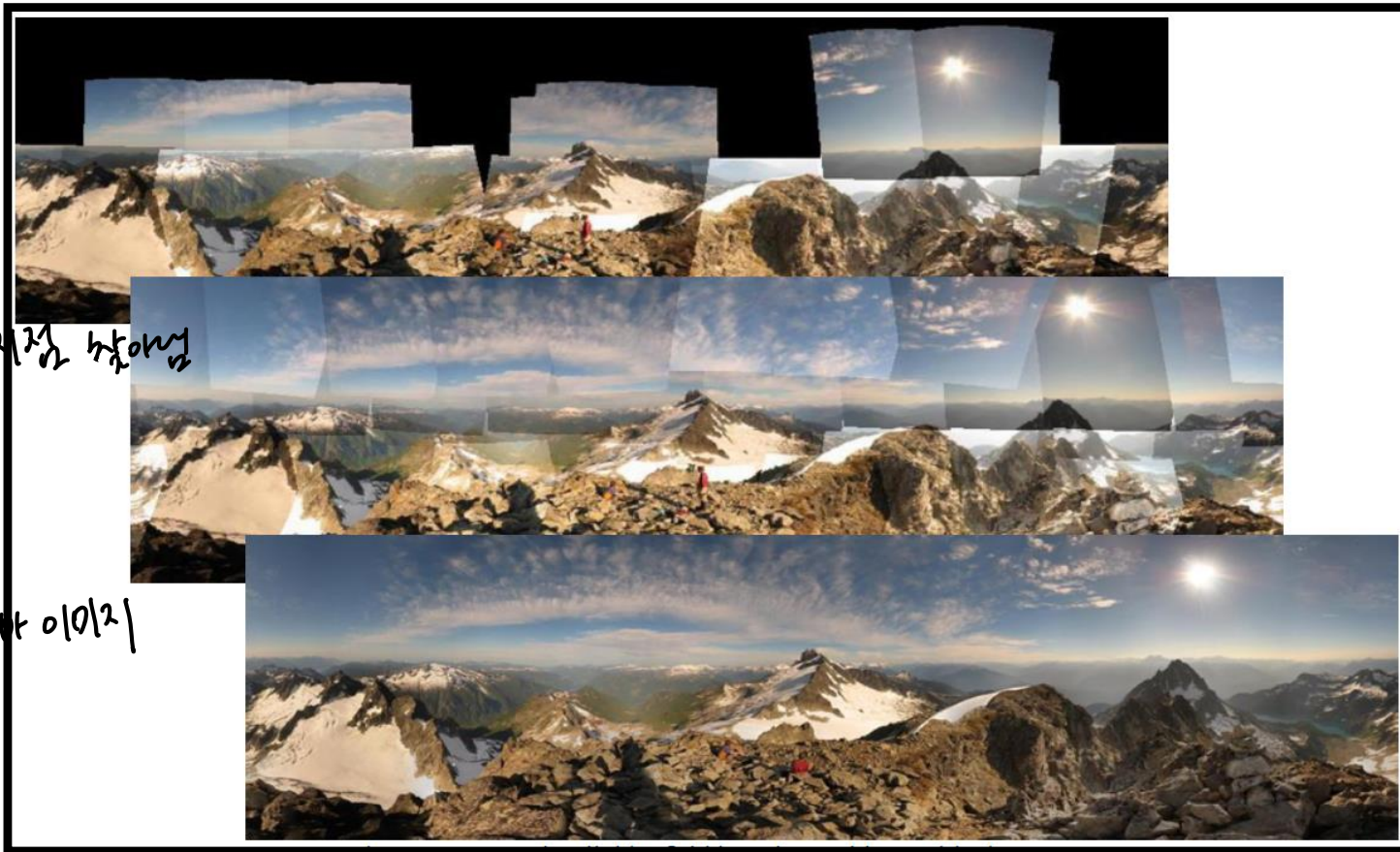


Image stitching result, Brown and Lowe. IJCV'07



2D Transformations

- Translation, Rotation, Scale, Similarity, Affine, Homography, Projective



Translation
shift



Rotation



Scale



Similarity

회전 + 크기일



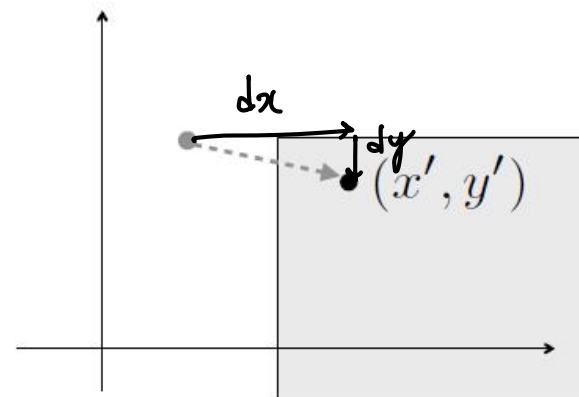
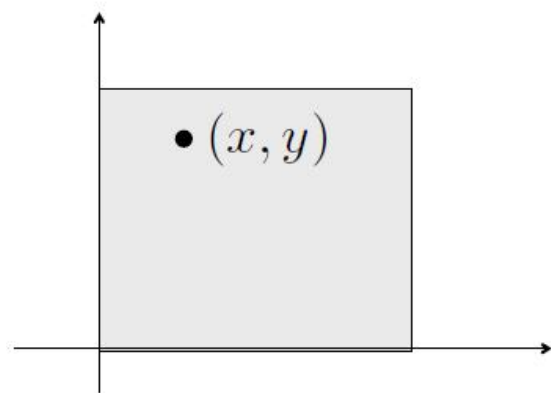
Affine



Homography



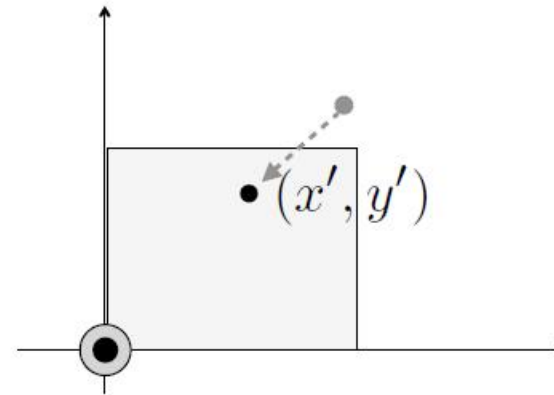
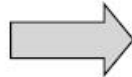
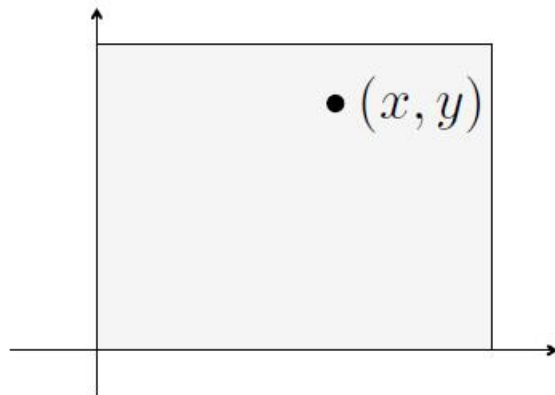
Translation



$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x + dx \\ y + dy \end{pmatrix}$$



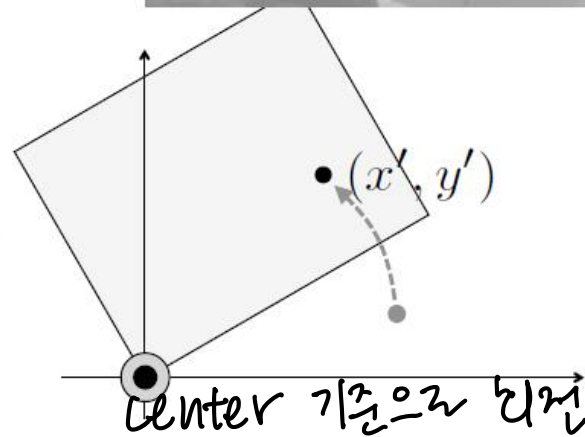
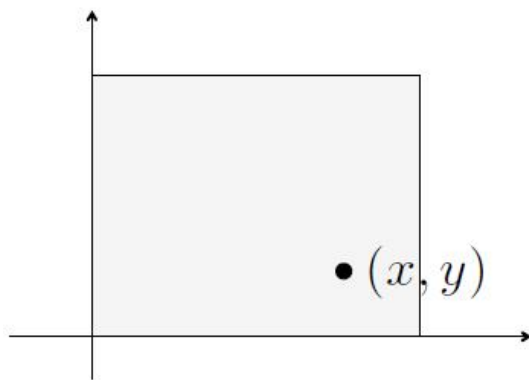
Scale



$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} sx \\ sy \end{pmatrix}$$



Rotation



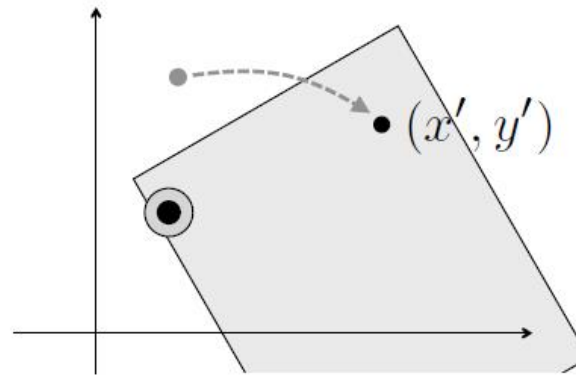
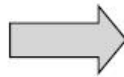
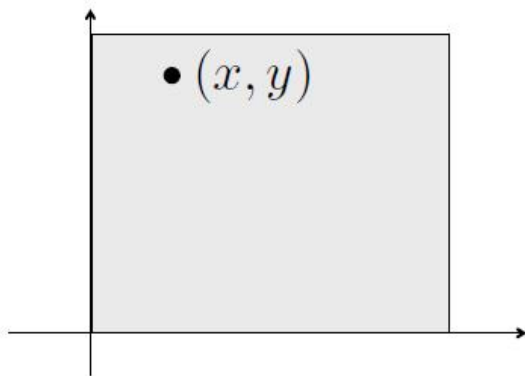
$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{pmatrix}$$



Similarity Transform (if no scale change → Euclidean)

shift
rotation

scaling, rotation, translation 다 포함됨



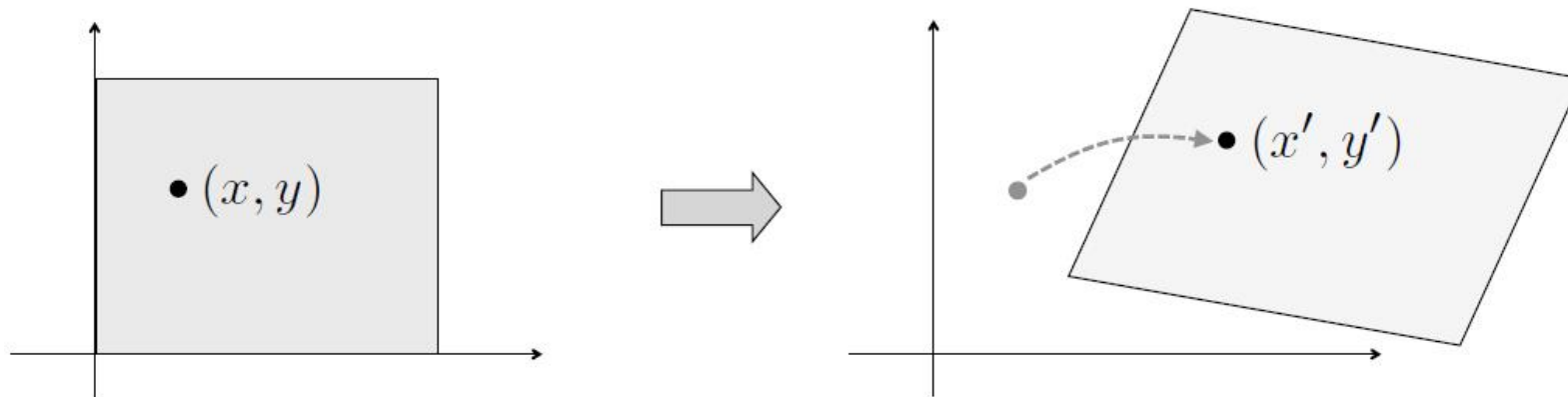
$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} sx \cos \theta - sy \sin \theta + dx \\ sx \sin \theta + sy \cos \theta + dy \end{pmatrix}$$

파라미터 4개 필요



Affine Transform

- Preserve points, lines, planes, & parallel lines remain parallel



$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} a_0x + a_1y + a_2 \\ a_3x + a_4y + a_5 \end{pmatrix}$$

파라미터 6개 필요



Summary: 2D Transformations

- Summary (자유도 DOF=?)
필요한 파라미터 개수

Translation

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x + dx \\ y + dy \end{pmatrix} \quad \text{DOF} = 2$$

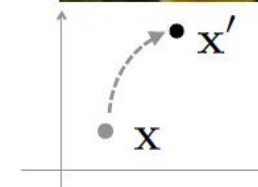
$$\mathbf{x}' = \mathbf{x} + \mathbf{d}$$



Scale

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} sx \\ sy \end{pmatrix} \quad \text{DOF} = 1$$

$$\mathbf{x}' = s\mathbf{x}$$



Rotation

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{pmatrix} \quad \text{DOF} = 1$$

$$\mathbf{x}' = R\mathbf{x} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \mathbf{x}$$

Similarity / Euclidean

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} sx \cos \theta - sy \sin \theta + dx \\ sx \sin \theta + sy \cos \theta + dy \end{pmatrix}$$

$$\begin{aligned} \mathbf{x}' &= sR\mathbf{x} + \mathbf{d} \\ \mathbf{x}' &= R\mathbf{x} + \mathbf{d} \end{aligned}$$

자유도 DOF가 적을수록
변환 결과가 제한적

Affine

$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} a_0x + a_1y + a_2 \\ a_3x + a_4y + a_5 \end{pmatrix} \quad \begin{aligned} \text{DOF} &= 4 \\ \text{DOF} &= 6 \end{aligned}$$

$$\mathbf{x}' = A\mathbf{x} + \mathbf{d} = \begin{pmatrix} a_0 & a_1 \\ a_3 & a_4 \end{pmatrix} \mathbf{x} + \begin{pmatrix} a_2 \\ a_5 \end{pmatrix}$$



Euclidean (or Cartesian) Coordinate

- Computational cost
 - What if we transform five times
 - $\begin{pmatrix} x' \\ y' \end{pmatrix} = A(A(A(A(A \begin{pmatrix} x \\ y \end{pmatrix} + B) + B) + B) + B) + B$
 - 5 multiplications, 5 additions
 - Huge computations with millions of points

좀 더 쉽게 할 수 있는 좌표계



Homogeneous Coordinate

- Representation

- Change 2D into 3D by adding 1's at the end

$$\bullet \begin{pmatrix} x' \\ y' \\ 1 \end{pmatrix} = \overset{\text{affine transform}}{\begin{pmatrix} a_0 & a_1 & a_2 \\ a_3 & a_4 & a_5 \\ 0 & 0 & 1 \end{pmatrix}} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = H \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

- Computational cost

- What if we transform five times

$$\bullet \begin{pmatrix} x' \\ y' \\ 1 \end{pmatrix} = \text{HHHHH} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} \quad \text{연산 간단해짐}$$

- 1 multiplication only! (HHHHH is known or precomputable)

1000000 pixel \rightarrow 1,000,000 만 연산



2D Transformations using Homogeneous Coordinates

- DOF=?

Translation: 유중세리만 눈알지니머스

$$\mathbf{x}' = \mathbf{x} + \mathbf{t} \qquad \mathbf{x}' = \begin{pmatrix} I & \mathbf{t} \end{pmatrix} \bar{\mathbf{x}}$$

Euclidean, rigid body (rotation + translation):

$$\mathbf{x}' = R \mathbf{x} + \mathbf{t} \qquad \mathbf{x}' = \begin{pmatrix} R & \mathbf{t} \end{pmatrix} \bar{\mathbf{x}}$$

$$R = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

Scaled rotation (similarity):

$$\mathbf{x}' = sR \mathbf{x} + \mathbf{t} \qquad \mathbf{x}' = \begin{pmatrix} sR & \mathbf{t} \end{pmatrix} \bar{\mathbf{x}}$$

Affine:

$$\mathbf{x}' = A \bar{\mathbf{x}}$$

$$A = \begin{pmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \end{pmatrix}$$

Projective (homography):

$$\bar{\mathbf{x}}' = H \bar{\mathbf{x}}$$

$$H = \begin{pmatrix} h_{00} & h_{01} & h_{02} \\ h_{10} & h_{11} & h_{12} \\ h_{20} & h_{21} & h_{22} \end{pmatrix}$$

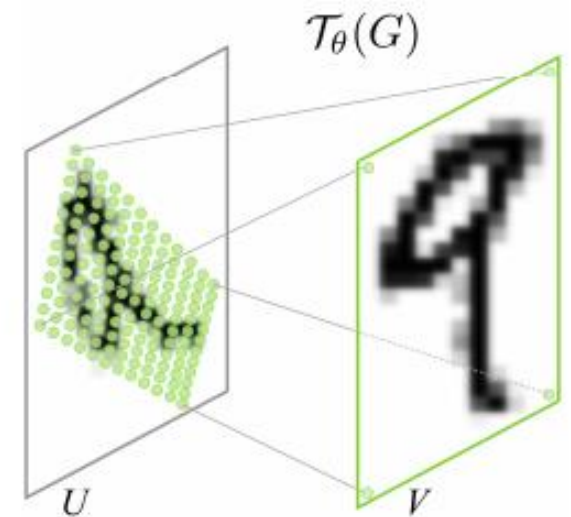
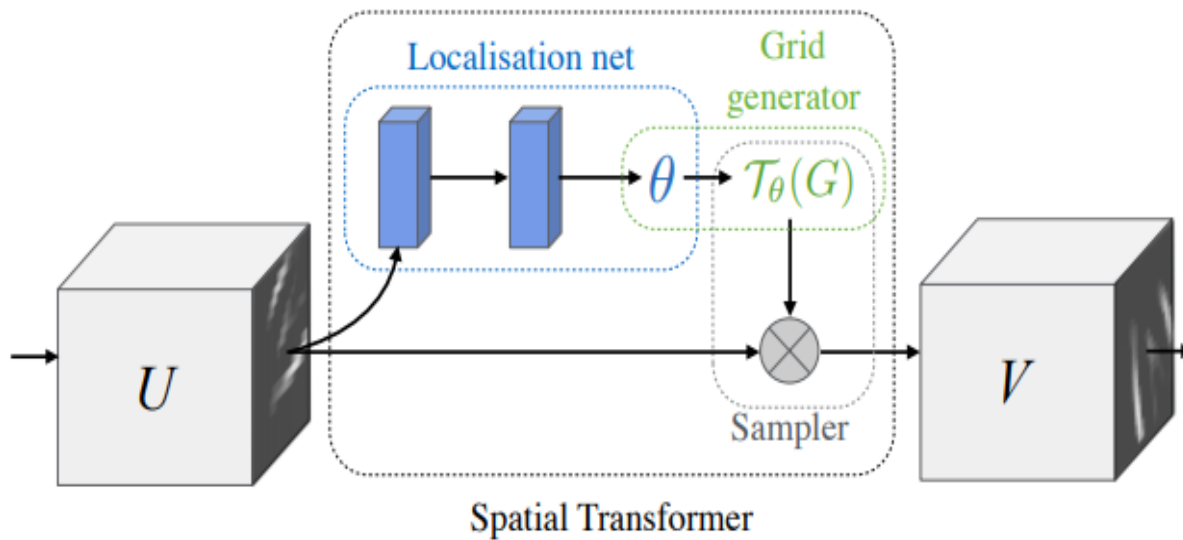


How can we compute transformation parameters?

- Spatial Transformer Networks

그냥 직접 사기에 적합한
transform parameter return

NN가 파라미터 찾아줄 수 있음
어떻게 변환 했는지



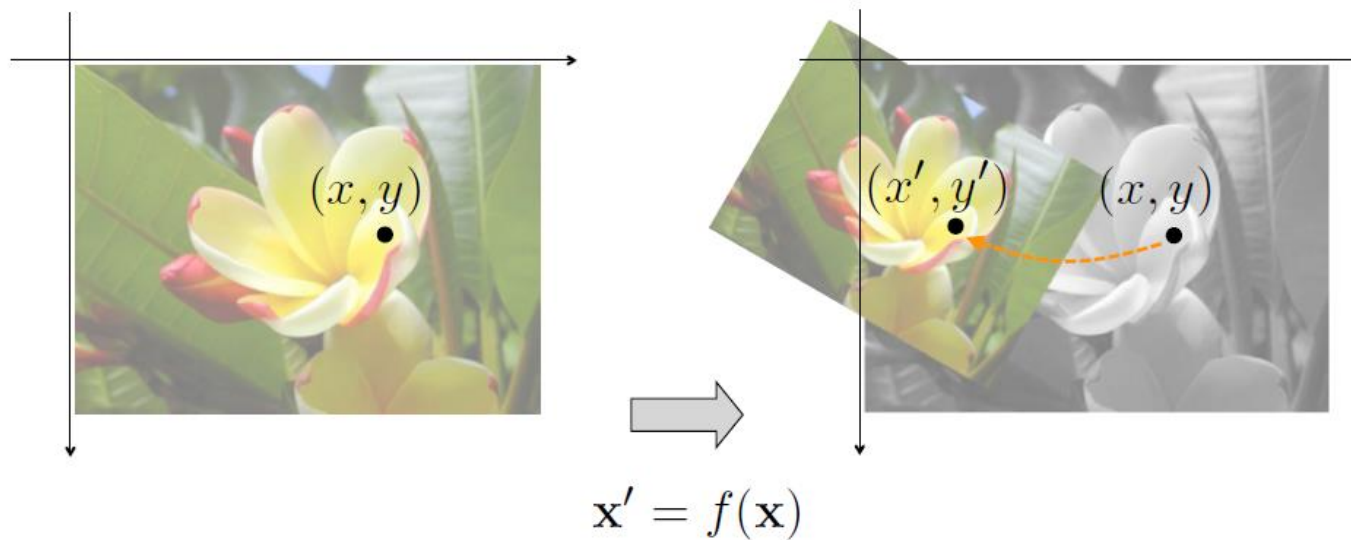
Jaderberg et al. NIPS'15



Image Warping

평면을 찌그러뜨리는 기술

- Easy to transfer a point with a given transformation
- How can we obtain a transformed (warped) image?

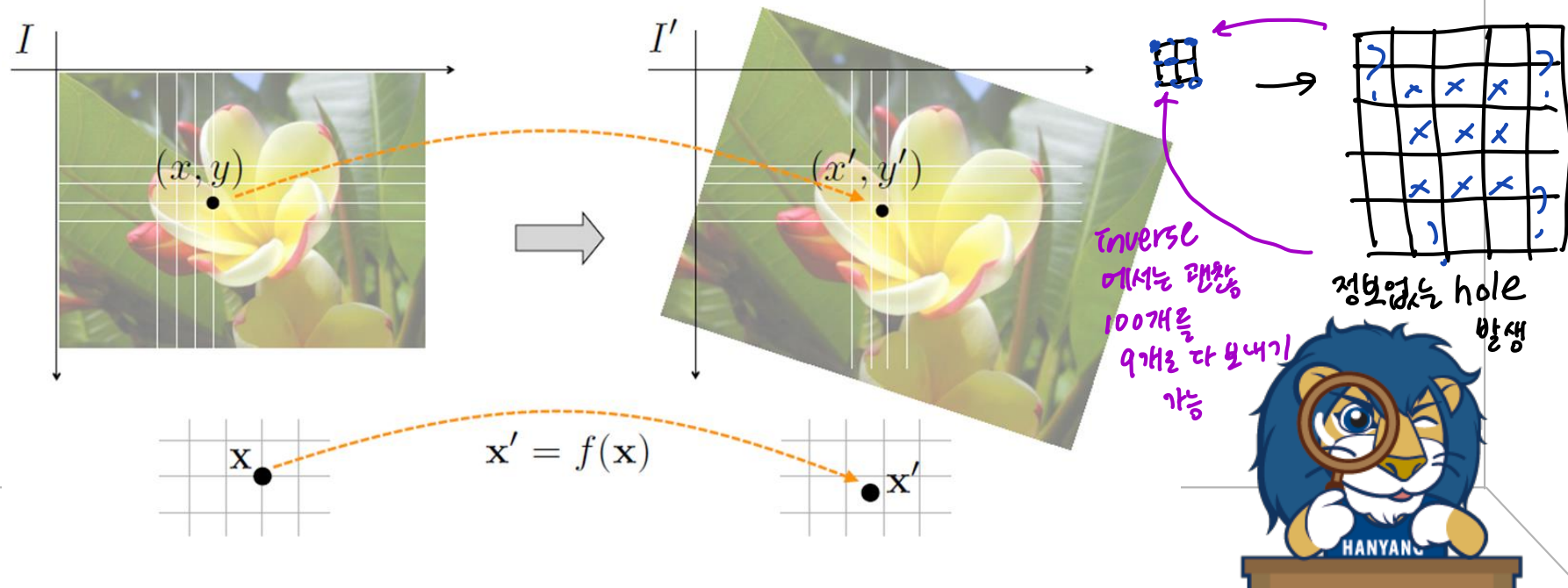
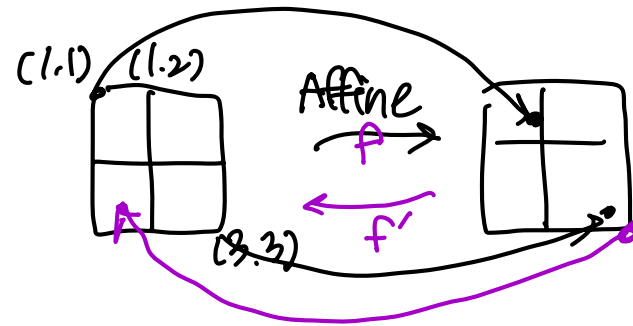


$$\begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} a_0 & a_1 & a_2 \\ a_3 & a_4 & a_5 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$



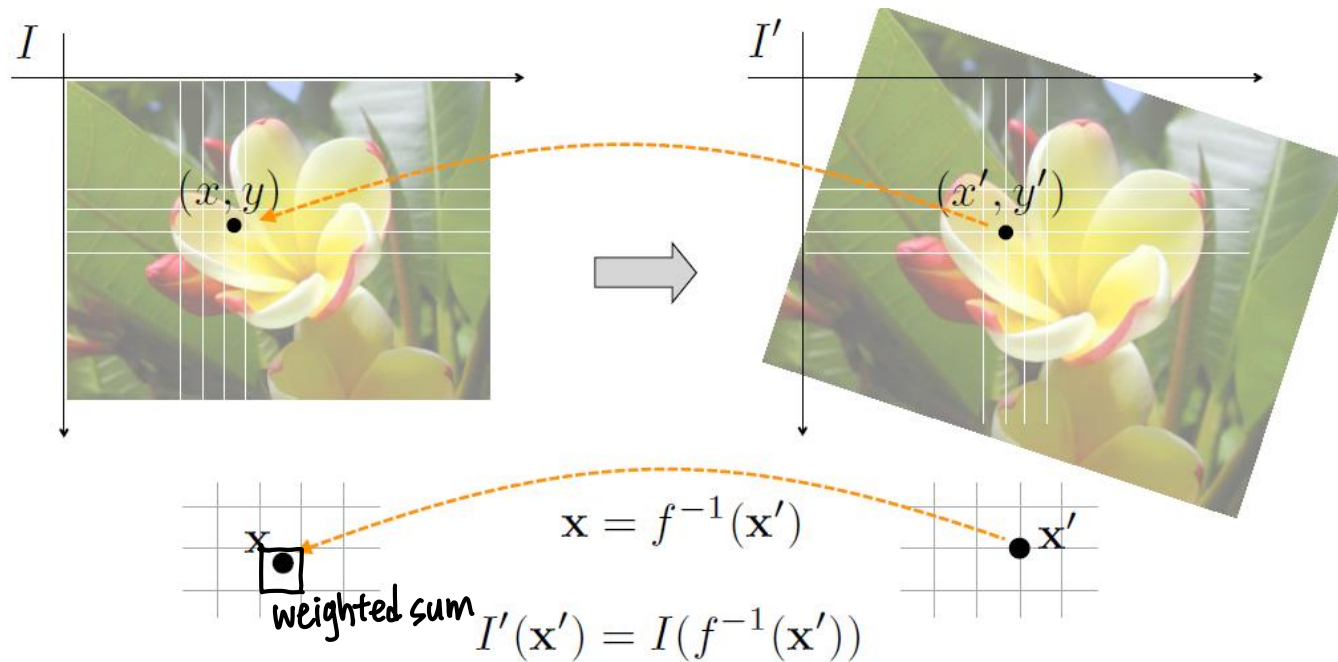
Forward Warping

- Pros
 - Easy to calculate target points
- Cons
 - Cracks & holes, floating point in the source image
 - Need interpolation

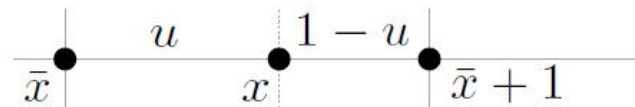
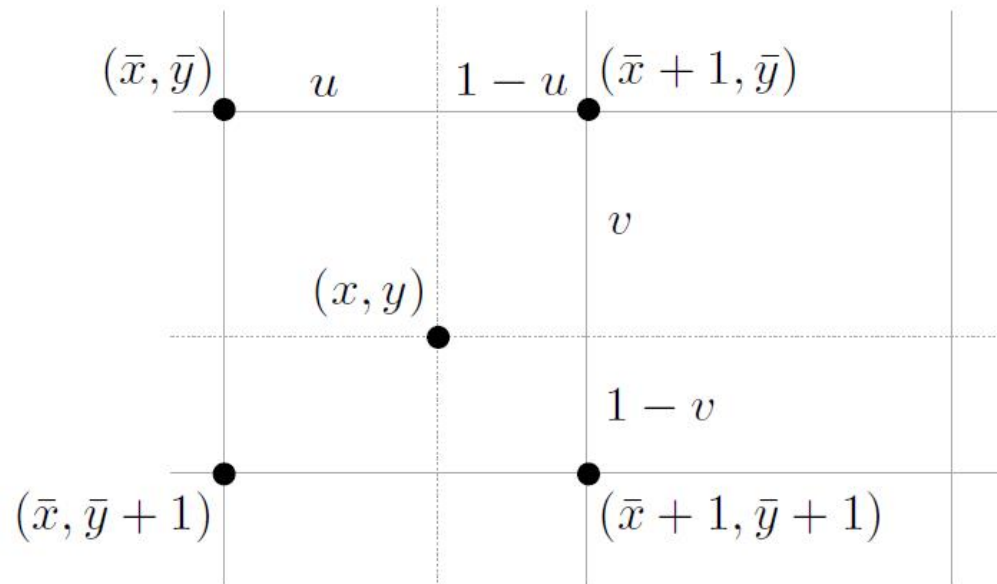
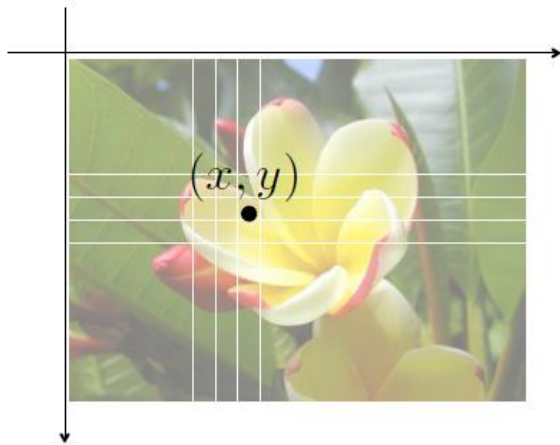


Inverse Warping

- Pros
 - Avoid cracks and holes
- Cons
 - Floating point in the source image
 - Need interpolation



Linear Interpolation



$$f(x) = (1 - u) f(\bar{x}) + u f(\bar{x} + 1)$$

$$\begin{aligned} I(x, y) = & (1 - u)(1 - v) I(\bar{x}, \bar{y}) + \\ & u(1 - v) I(\bar{x} + 1, \bar{y}) + \\ & (1 - u)v I(\bar{x}, \bar{y} + 1) + \\ & uv I(\bar{x} + 1, \bar{y} + 1) \end{aligned}$$



Recent Warping Technique

- Learning-based approach



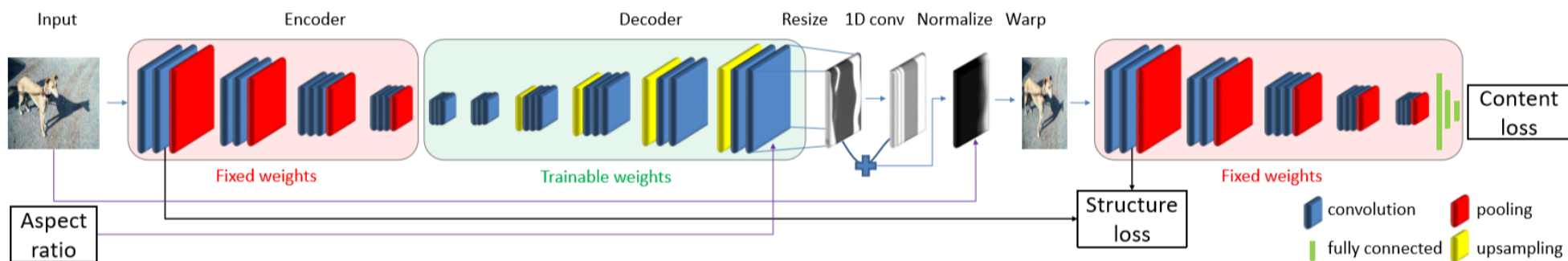
Original image



Warping results

WSSDCNN, Cho et al. ICCV'17

- Network architecture



Thank you!

