

# Project 2

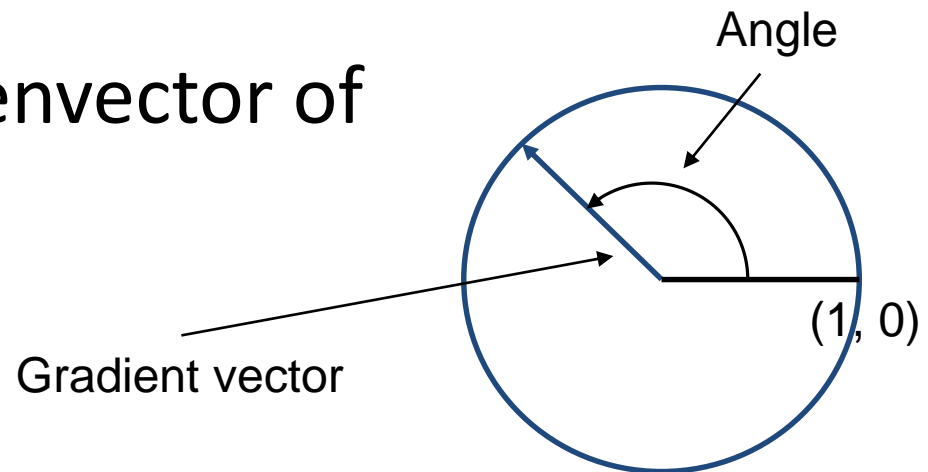
- Python, NumPy, SciPy
  - Tutorial link in writeup
- Main parts
  - Harris corner detection
  - MOPS feature descriptor
  - Simple feature matching

# Images + NumPy

- Image coordinates:  $x, y$
- Numpy array
  - Access like a matrix
  - Pixel at coordinate  $(x, y)$  is `image[y, x]`

# Keypoint orientation

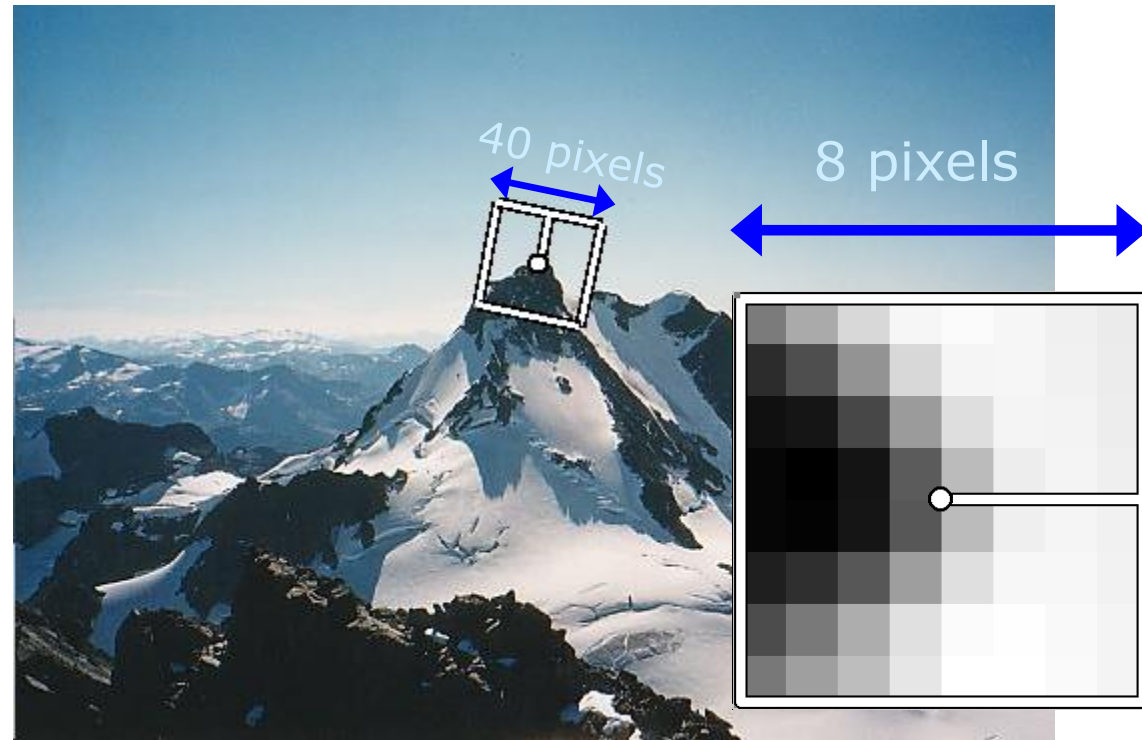
- Harris corner detector
  - Feature orientation angle *approximation*
    - *angle of gradient* computed on blurred image. 0 angle: counter clockwise from the point  $(1, 0)$
  - **Not** given by an eigenvector of the structure tensor



# Multiscale Oriented Patches descriptor






Take 40x40 square window around detected feature

- Prefilter (because we are subsampling)
- Scale to 1/5 size
- Rotate to horizontal
- Sample 8x8 square window centered at feature
- Intensity normalize the window by subtracting the mean, dividing by the standard deviation in the window
- You **don't** have to implement the multiscale part



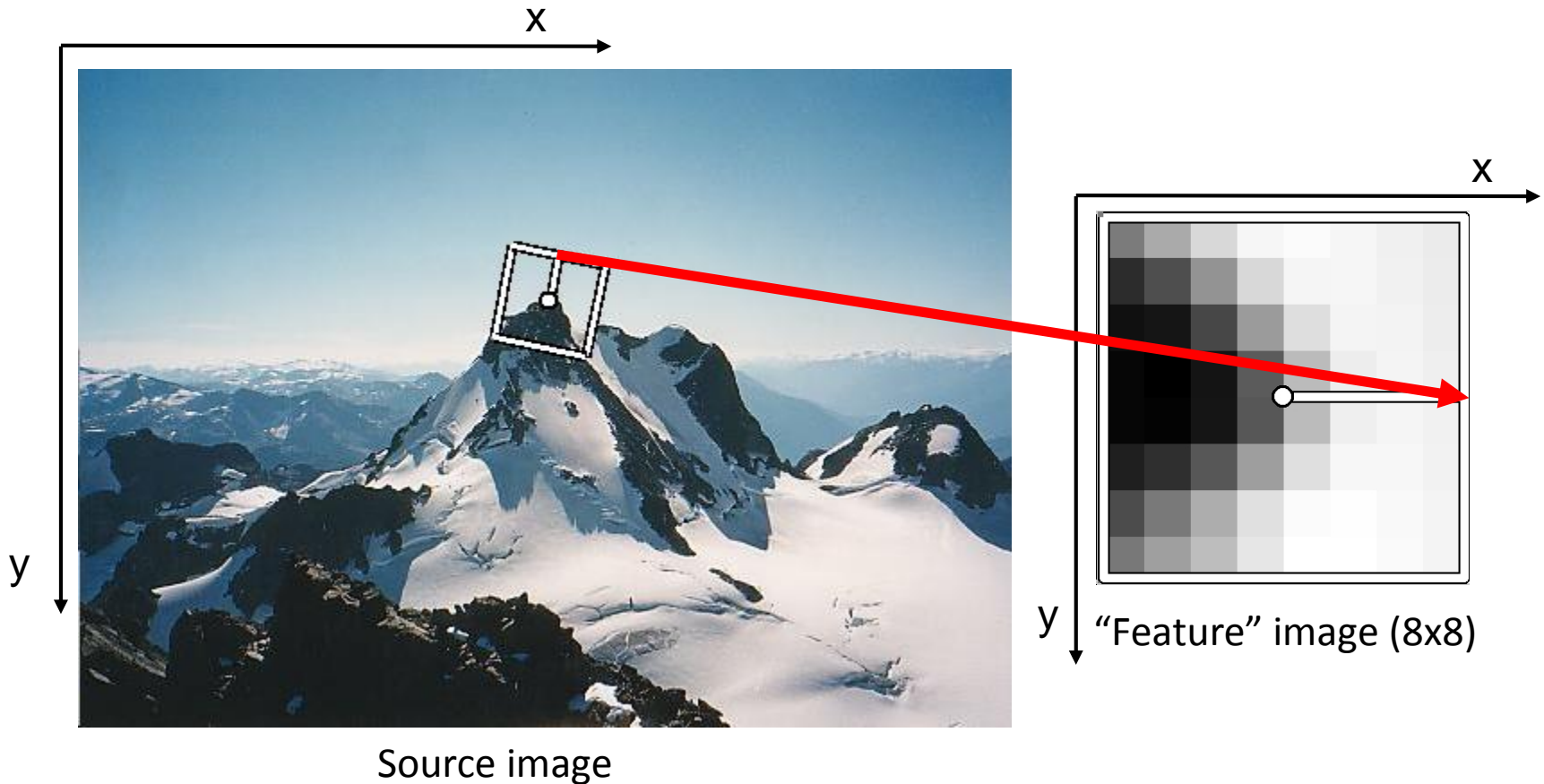
CSE 576: Computer Vision

# Affine transformation

Transformation	Matrix	# DoF	Preserves	Icon
translation	$\begin{bmatrix} I &   & t \end{bmatrix}_{2 \times 3}$	2	orientation	
rigid (Euclidean)	$\begin{bmatrix} R &   & t \end{bmatrix}_{2 \times 3}$	3	lengths	
similarity	$\begin{bmatrix} sR &   & t \end{bmatrix}_{2 \times 3}$	4	angles	
affine	$\begin{bmatrix} A \end{bmatrix}_{2 \times 3}$	6	parallelism	
projective	$\begin{bmatrix} \tilde{H} \end{bmatrix}_{3 \times 3}$	8	straight lines	

**Table 2.1** Hierarchy of 2D coordinate transformations. Each transformation also preserves the properties listed in the rows below it, i.e., similarity preserves not only angles but also parallelism and straight lines. The  $2 \times 3$  matrices are extended with a third  $[0^T \ 1]$  row to form a full  $3 \times 3$  matrix for homogeneous coordinate transformations.

# MOPS descriptor



- We provide you a routine, **cv2.warpAffine**, that can perform the resampling, transformation and cropping
- You have to pass a **forward** warping affine transformation matrix (2x3), multiplied from the **left**, the coordinates are represented as a **column vector**

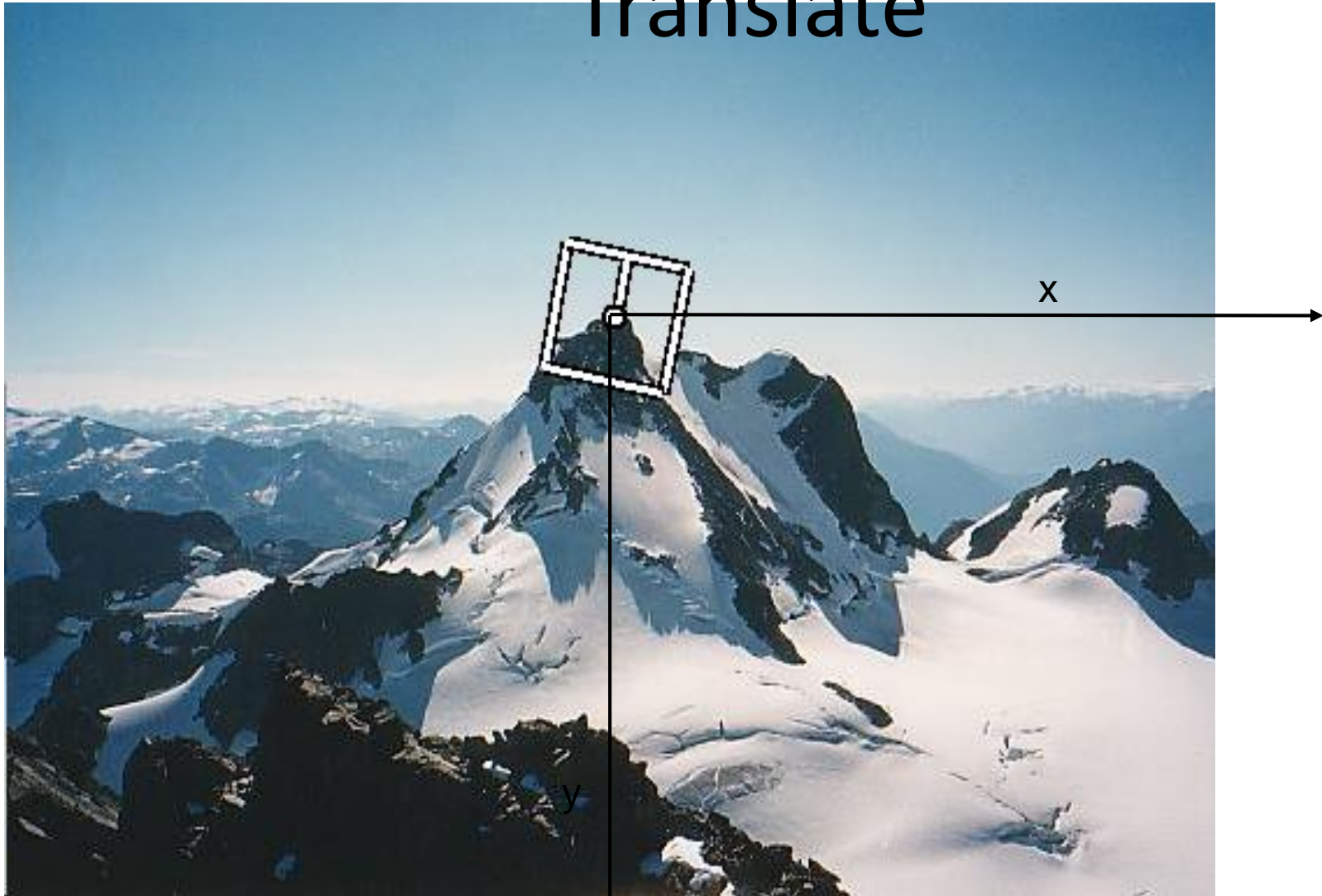
# MOPS descriptor

- You can combine transformations together to get the final transformation
- Pass this transformation matrix to `cv2.warpAffine`

$T = ?$



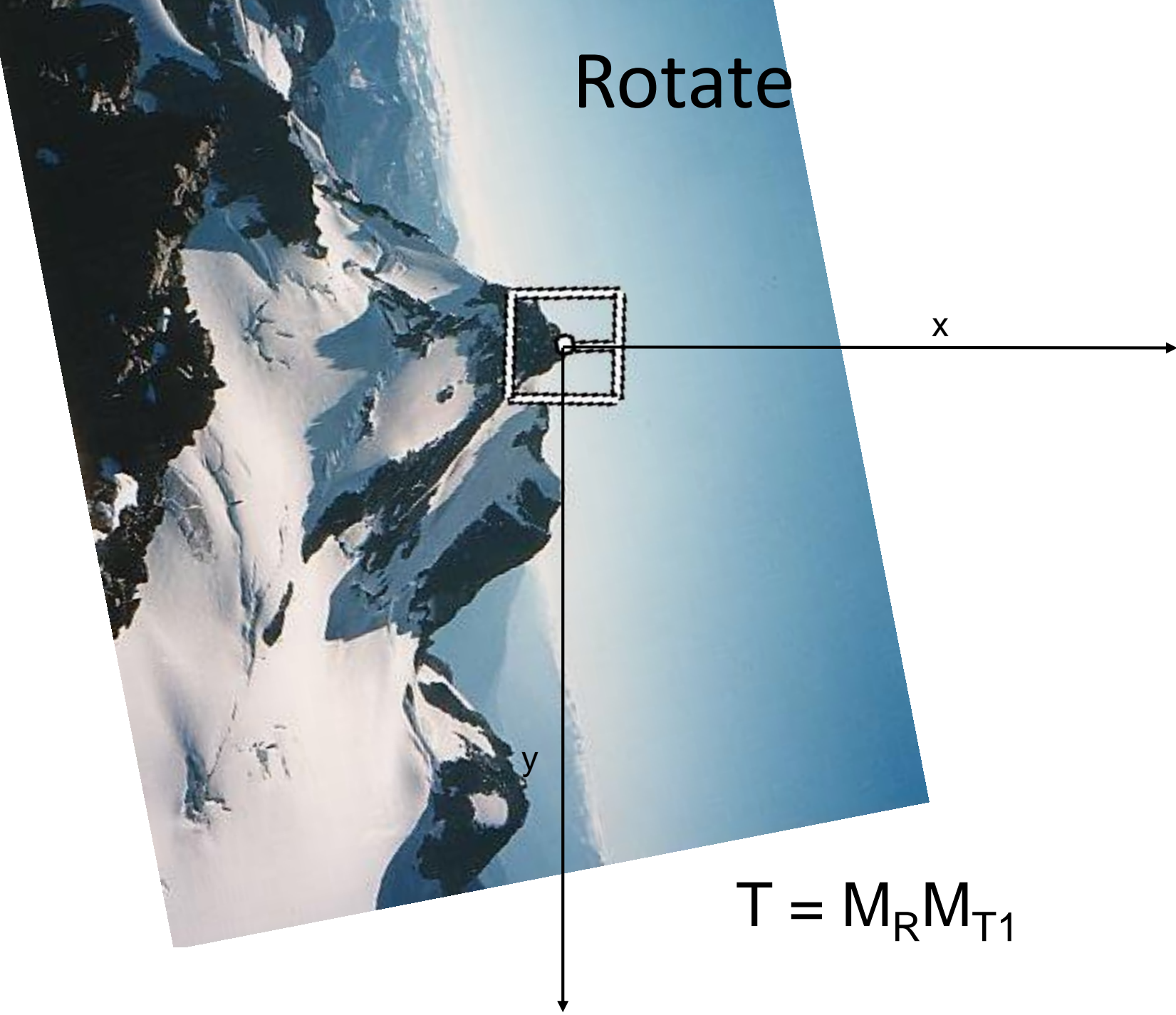
# Translate



$$T = M_{T_1}$$

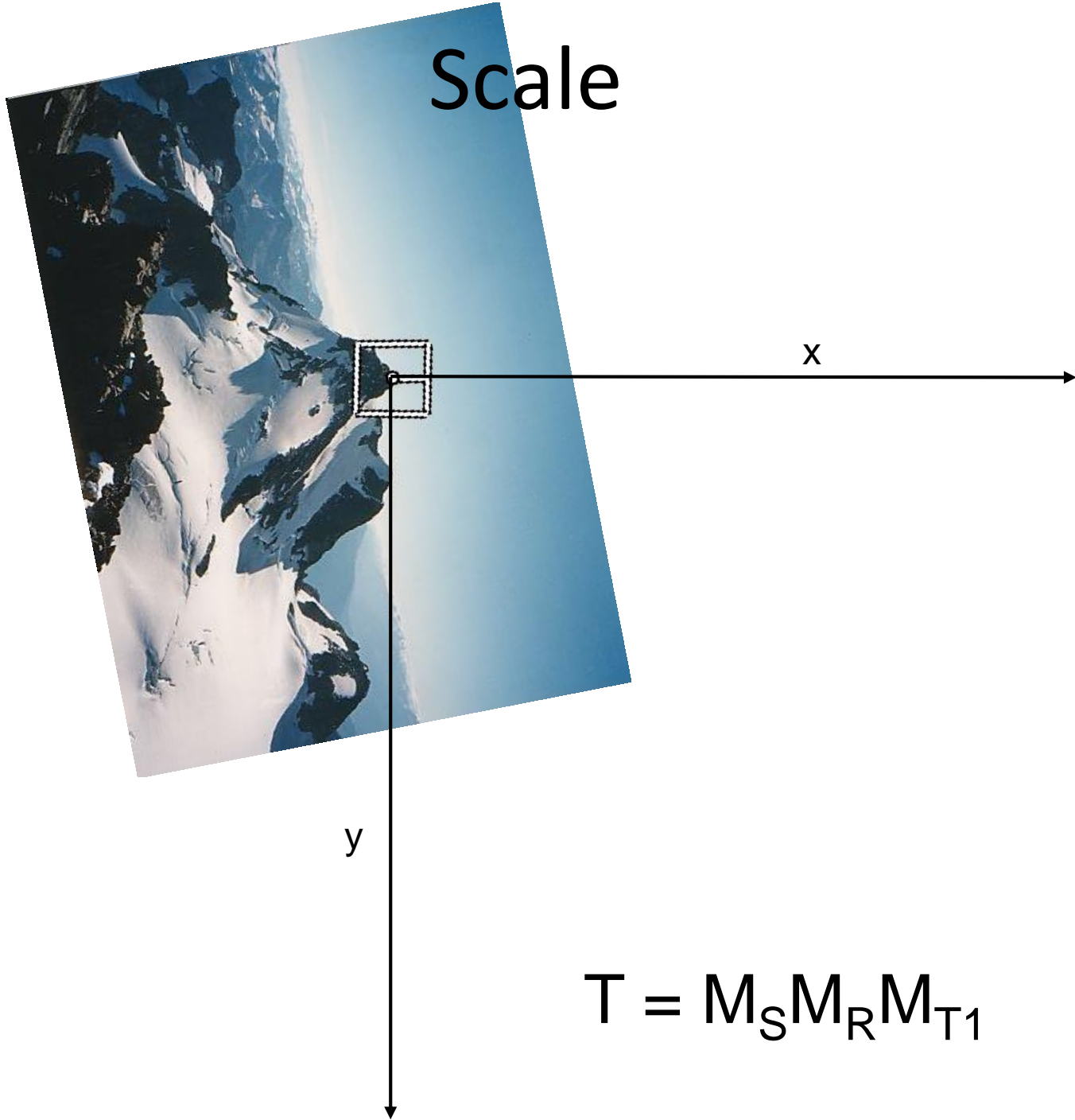


Rotate

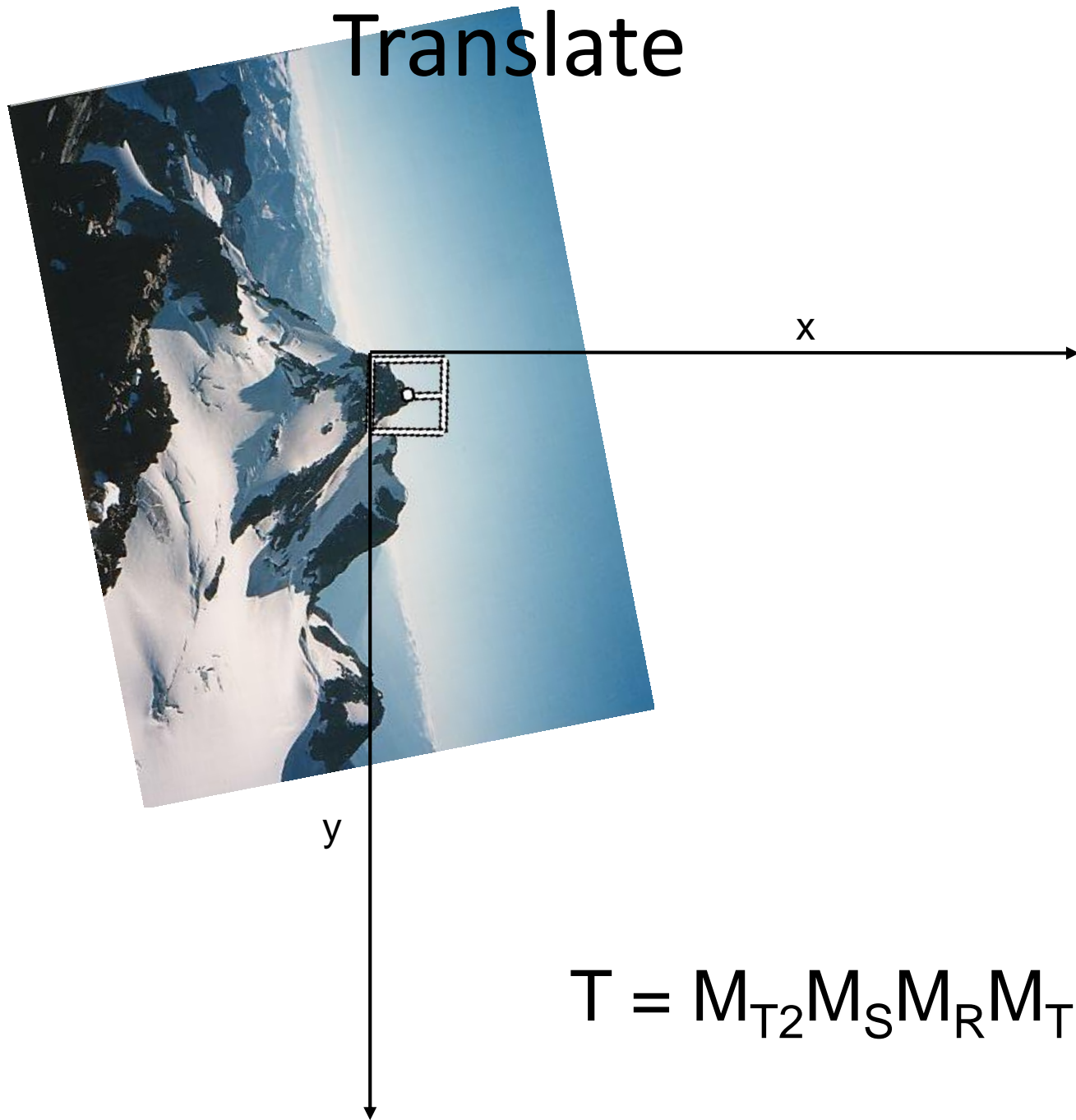


$$T = M_R M_{T1}$$

# Scale



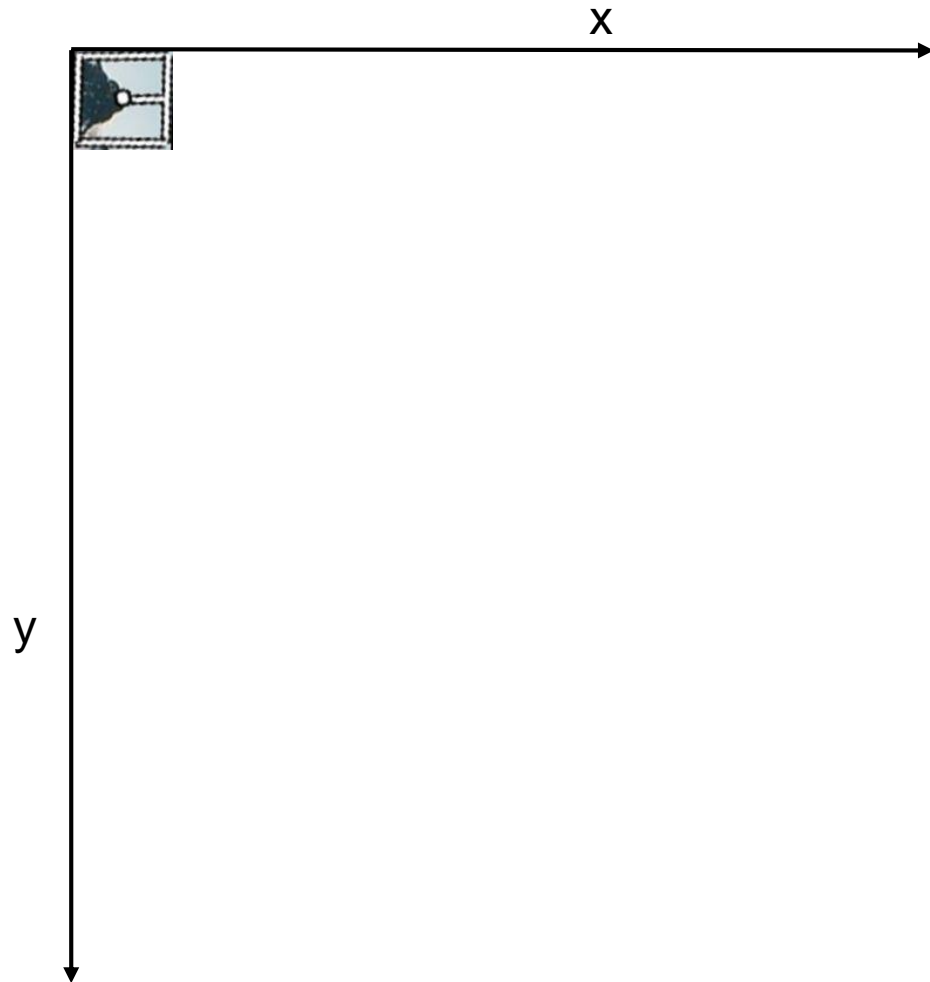
# Translate



$$T = M_{T2}M_S M_R M_{T1}$$

# Crop

- *cv2.warpAffine* also takes care of the cropping



# Demo