



Image warping - introduction

© 1997-2015 Josef Pelikán
CGG MFF UK Praha

pepca@cgg.mff.cuni.cz
<http://cgg.mff.cuni.cz/~pepca/>



Warping .. image deformation

- ◆ **texture mapping** in 3D rendering (after rasterization)
 - perspective distortion, mapping textures to arbitrary shapes
- ◆ correction of **geometric distortion** (digital image acquirement)
 - satellite and aerial photography
 - scanning of deformed documents
- ◆ **special effects** in TV, film and advertisement



Mathematical model

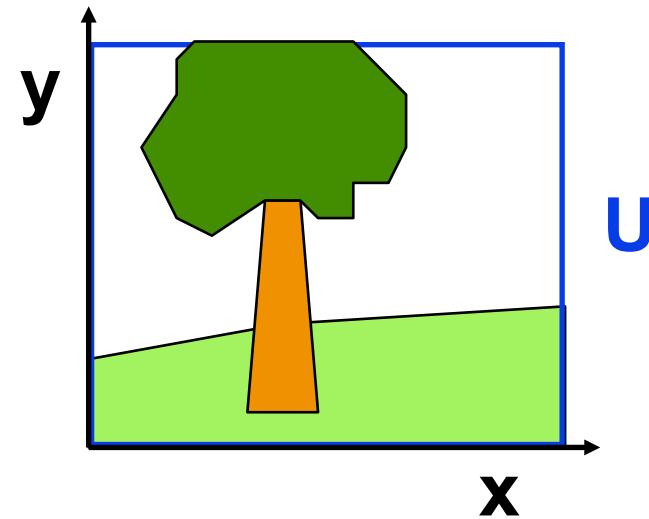
„image function”

$$f: U \subset \mathbb{R}^2 \rightarrow \mathbb{R}^n$$

$$f: [x, y] \rightarrow [a_1, a_2, \dots, a_n]$$

position on
the plane

image attributes
(color, transparency)





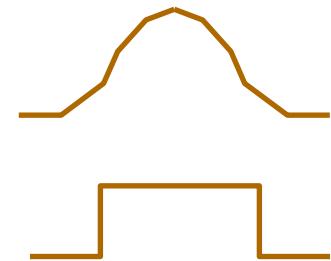
Spatial discretization

Digitized raster image:

$$I: \langle 0..m-1 \rangle \times \langle 0..n-1 \rangle \rightarrow \mathbb{R}^n$$

Digitization using filter d:

$$I_f(i, j) = \iint_{\mathbb{R}^2} f(x, y) \cdot \underline{d(x - i, y - j)} \, dx \, dy$$



d .. device characteristics
(optical system, CCD element)



Digital image reproduction

Reconstruction of a discrete image:

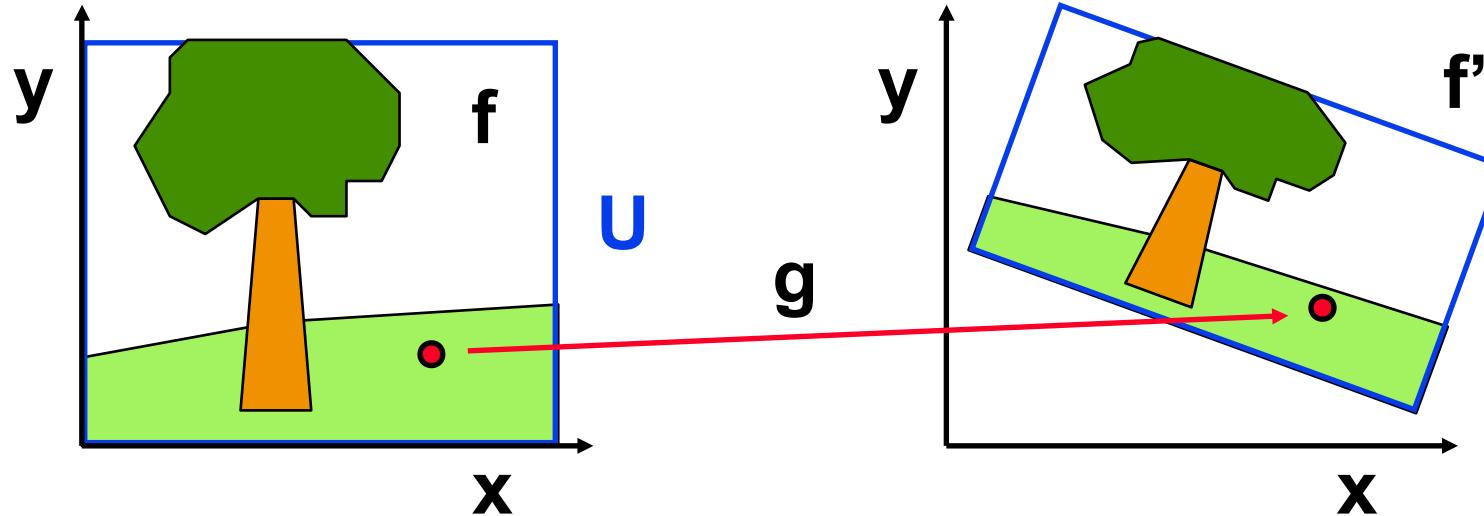
$$f^r(x, y) = \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} l_f(i, j) \cdot r(i - x, j - y)$$

r .. output device characteristics
(impulse response)

.. we need f^r to be similar to f
(in frequency range defined by the Nyquist law)



Geometric transform



$$g: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$\begin{aligned}f(x, y) &= f'(g(x, y)) \\f(g^{-1}(u, v)) &= f'(u, v)\end{aligned}$$

But what about raster images?

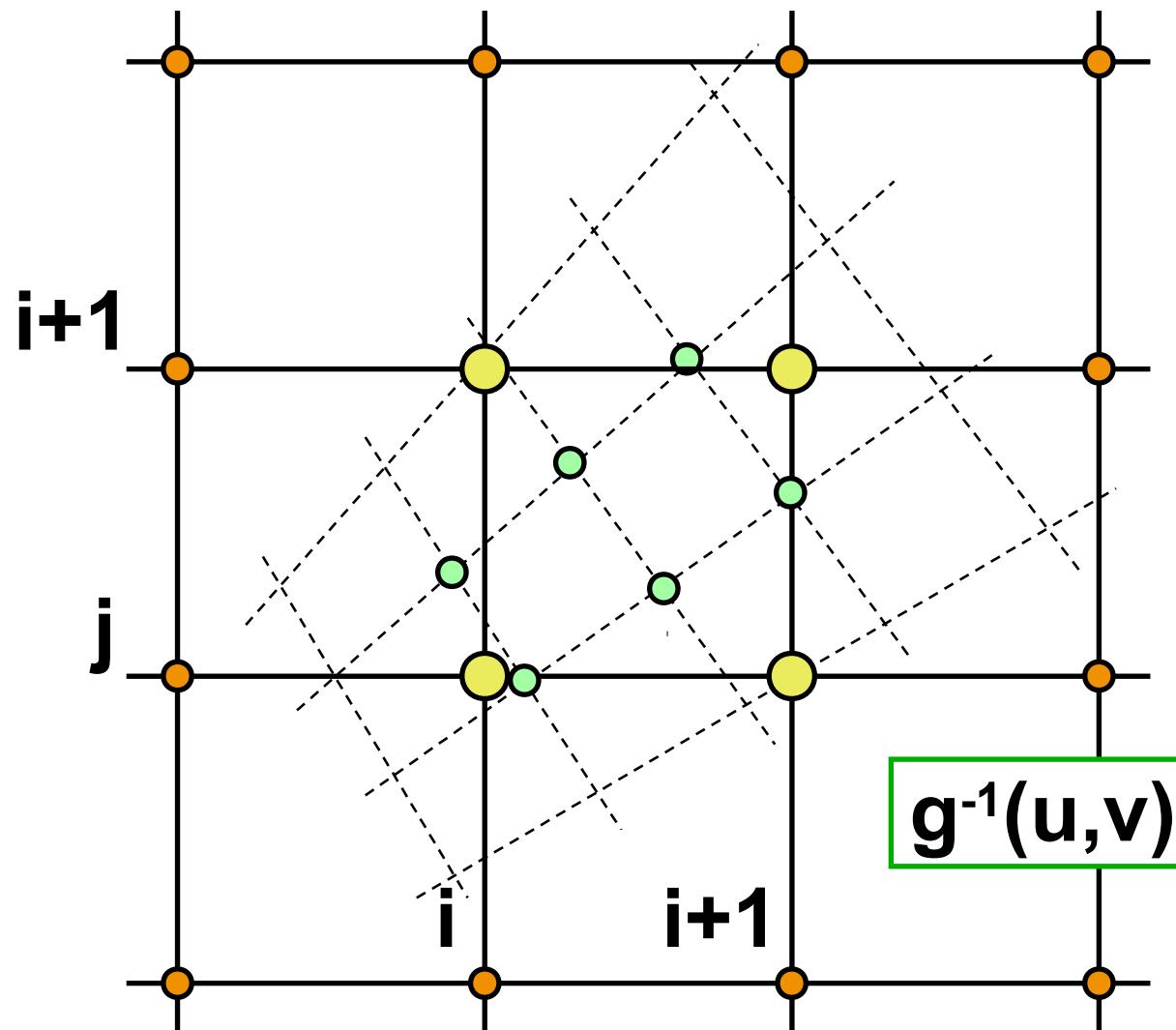
$$G(I_f) \simeq I_{f'}$$



Transform with interpolation

- digitization = **sampling**
 - digitizing filter = Dirac delta
- attributes (color) of transformed pixel computed by **approximation or interpolation**
 - inverse transform function \mathbf{g}^{-1} is needed
- „**rounding**“, **polynomial interpolation**
 - bilinear to bicubic interpolation/approximation is sufficient

Interpolation in source coord. system



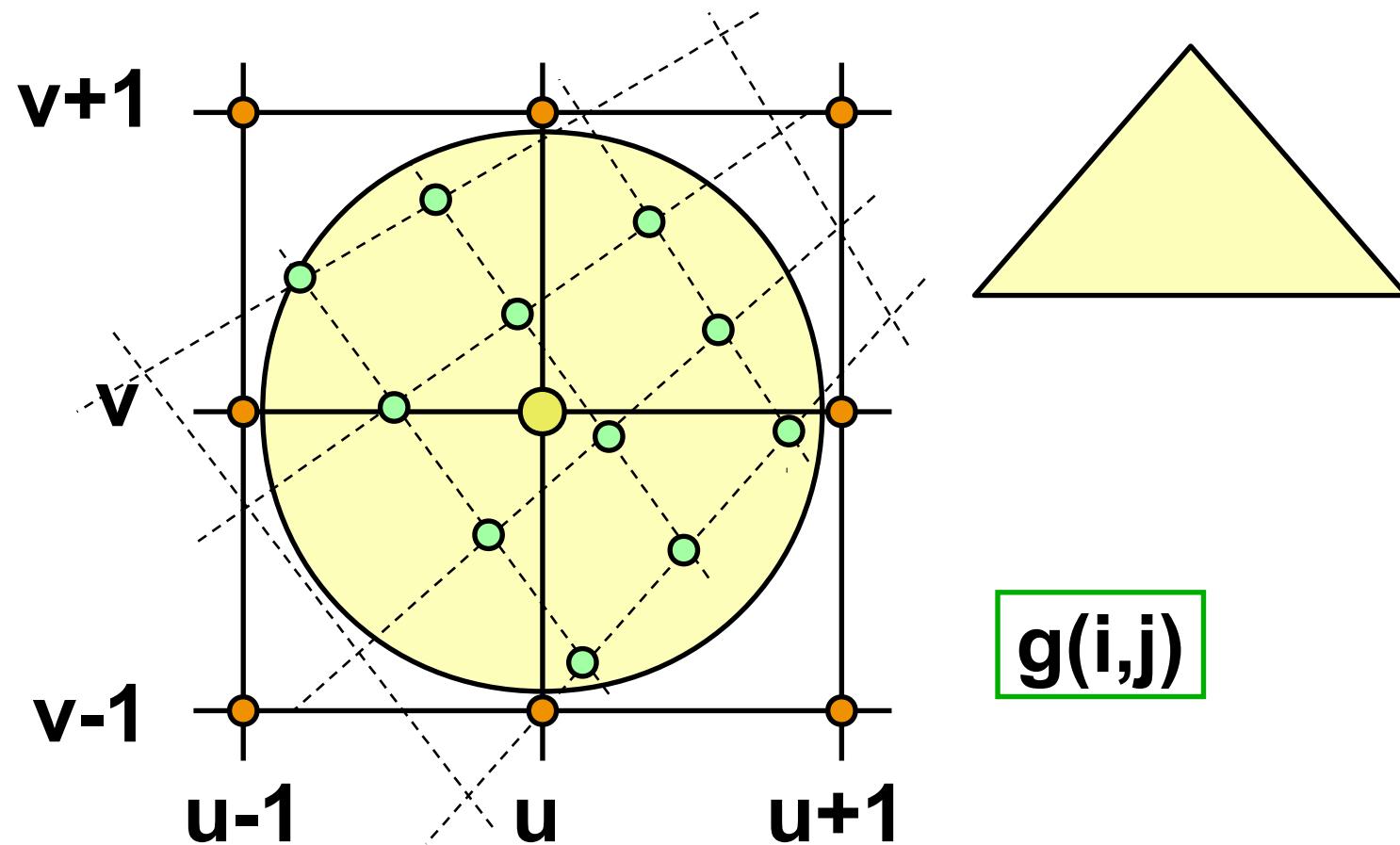


Transform with filtering

- **pixel-area model**
 - digitizing filter has areal support (e.g. box or conical filter)
- **source pixels** are projected to target coordinate system
 - only **g** is needed
- suitable also for **contractive transforms**
 - isometric transform \Rightarrow image is blurred
 - big contraction \Rightarrow high computing time (speedup needed)



Transform with filtering



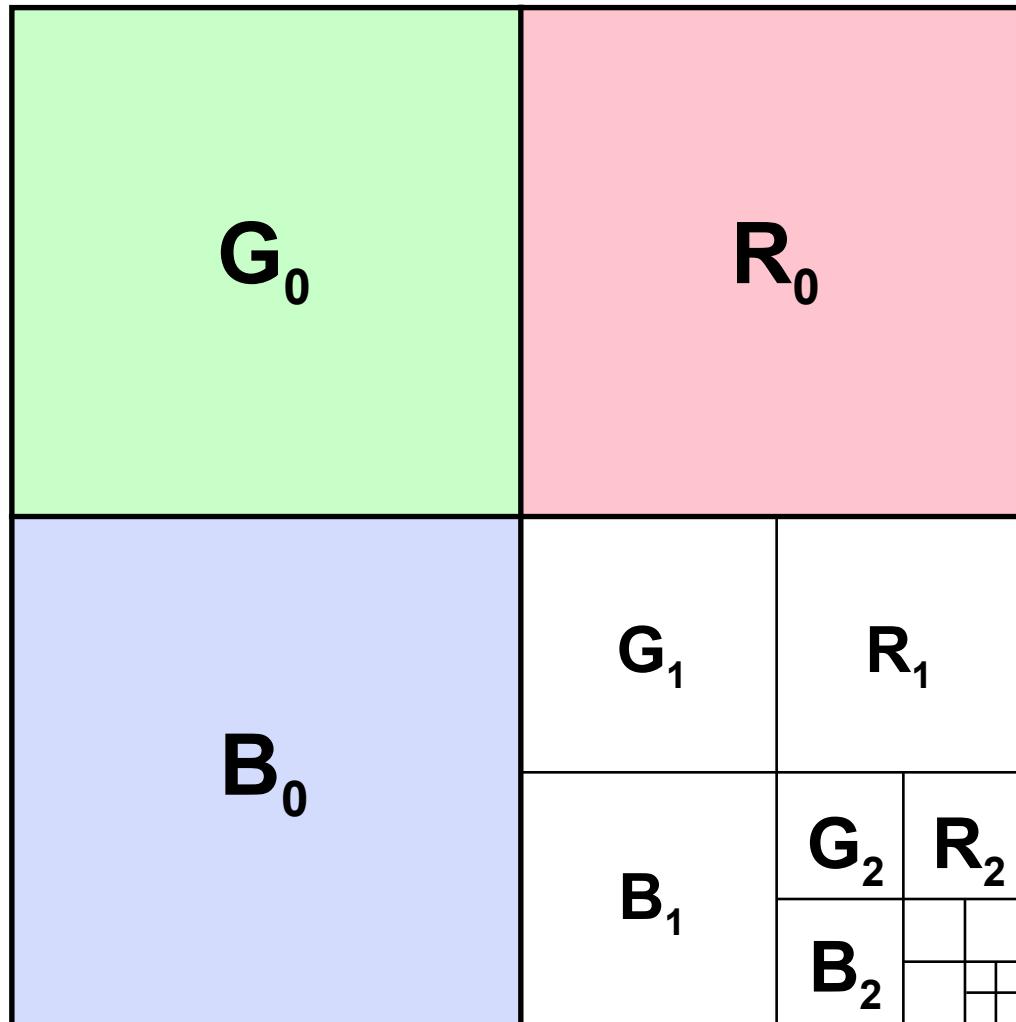


„MIP map“ (*multum in parvo*)

- **pyramidal representation** of the source image (hierarchy)
 - preprocessing = pre-filtering
- useful mostly for **transforms with big contraction**
 - substantial speedup
 - used mostly for **texture mapping** on the GPU (distant objects)
- **compact storage**
 - only **4/3** size of original RGB image



„MIP map”



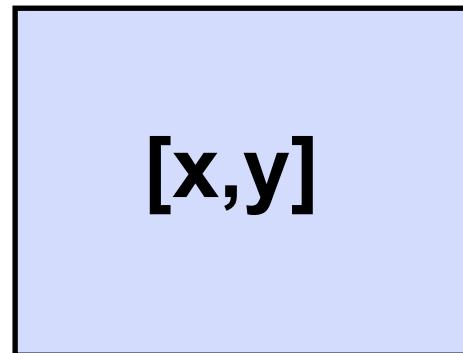


Multi-pass algorithms

- transform is factored into several **consecutive steps**
 - each step works either **on rows** or **on columns** of an image
- **faster computation**
 - simpler filtering
 - two 1D filters are faster than a 2D one
- **bottleneck problem**
 - partial mapping is highly contractive or even non-injective

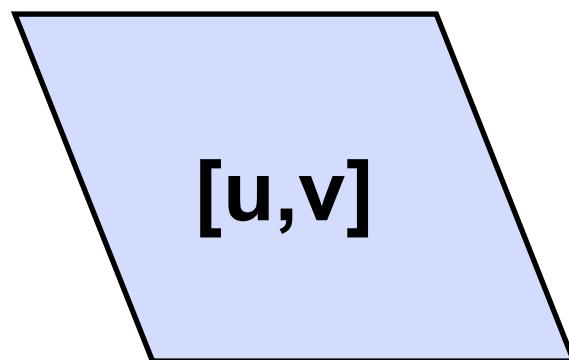


Two-pass rotation

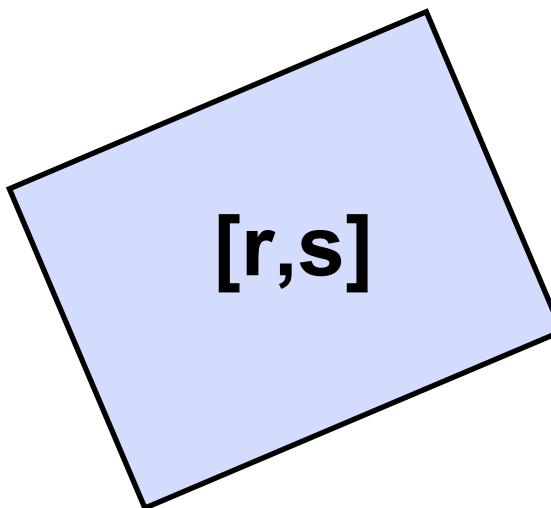


$$\begin{aligned}[u,v] &= f(x,y) = [f_1(x,y), y] \\ [r,s] &= g(u,v) = [u, g_1(u,v)] \\ &= [f_1(x,y), g_1(f_1(x,y), v)]\end{aligned}$$

↓
 f



→
 g





Derivation

Target transform (rotation by α):

$$\begin{aligned} r &= x \cdot \cos\alpha - y \cdot \sin\alpha \\ s &= x \cdot \sin\alpha + y \cdot \cos\alpha \end{aligned}$$

First transform f :

$$\begin{array}{ll} \underline{u = f_1(x, y)} & \underline{v = y} \end{array}$$

Second transform g :

$$\begin{array}{ll} \underline{r = u = f_1(x, y)} & \underline{s = g_1(u, v) = g_1[f_1(x, y), y]} \end{array}$$



Derivation

Horizontal shear for the first pass:

$$f_1(x, y) = x \cdot \cos\alpha - y \cdot \sin\alpha$$

Vertical shear for the second pass:

$$g_1(u, v) = x \cdot \sin\alpha + y \cdot \cos\alpha = x \cdot \sin\alpha + v \cdot \cos\alpha$$

$$u = f_1(x, v) = x \cdot \cos\alpha - v \cdot \sin\alpha$$

$$x = (u + v \cdot \sin\alpha) / \cos\alpha$$

$$g_1(u, v) = u \cdot \tan\alpha + v \cdot \sec\alpha$$



Three-pass rotation

Rotation matrix factoring:

$$\begin{bmatrix} \cos\alpha & -\sin\alpha \\ \sin\alpha & \cos\alpha \end{bmatrix} = \\ = \begin{bmatrix} 1 & -\tan\alpha/2 \\ 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ \sin\alpha & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & -\tan\alpha/2 \\ 0 & 1 \end{bmatrix}$$

sufficient for angles: $-\pi/2 \leq \alpha \leq \pi/2$



Arbitrary angle

- **multi-pass methods** cannot be used for wide range of angles. Suitable angle ranges:
 - $-45^\circ \leq \alpha \leq 45^\circ$ for two-pass rotation
 - $-90^\circ \leq \alpha \leq 90^\circ$ for three-pass rotation
- ◆ **auxiliary rotation** (multiple of 90°)
 - simple, fast
 - no image degradation

General separable transform

Separable transform:

$$\mathbf{h}(\mathbf{x}, \mathbf{y}) = [\mathbf{h}_1(\mathbf{x}, \mathbf{y}), \mathbf{h}_2(\mathbf{x}, \mathbf{y})] = \underline{\mathbf{g}(\mathbf{f}(\mathbf{x}, \mathbf{y}))}$$

$$\underline{\mathbf{f}(\mathbf{x}, \mathbf{y})} = [\mathbf{f}_1(\mathbf{x}, \mathbf{y}), \mathbf{y}]$$

$$\underline{\mathbf{g}(\mathbf{u}, \mathbf{v})} = [\mathbf{u}, \mathbf{g}_1(\mathbf{u}, \mathbf{v})]$$

$$\mathbf{f}_1(\mathbf{x}, \mathbf{y}) = \mathbf{h}_1(\mathbf{x}, \mathbf{y})$$

$$\mathbf{g}_1(\mathbf{u}, \mathbf{v}) = \mathbf{h}_2(\varphi(\mathbf{u}, \mathbf{v}), \mathbf{v})$$

if $\exists \varphi(\mathbf{u}, \mathbf{v})$ such that
 $\underline{\mathbf{x} = \varphi(\mathbf{u}, \mathbf{v})}$



Image degradation

Inapplicable areas: low values of the derivative

$$\frac{\partial h_1(x, y)}{\partial x}$$

or high values of the derivative

$$\frac{\partial g_1(u, v)}{\partial v}$$

Sometimes even the inverse function could not
be defined $\phi(u, v)$!

⇒ concurrent processing of images I and I^T



Optimized algorithm

- ◆ concurrent processing of original \mathbf{I} and transposed \mathbf{I}^T image
 - both branches use the same method (two-pass algorithm using separable transforms \mathbf{h} and \mathbf{h}^T)
- ◆ result pixels are composed from $\mathbf{h}(\mathbf{I})$ and $(\mathbf{h}^T(\mathbf{I}^T))^T$
 - comparing degradation pixel by pixel
- ◆ even for „non-injective“ deformations!
 - image „folding“..



The End

More info:

- **J. Foley, A. van Dam, S. Feiner, J. Hughes:**
Computer Graphics, Principles and Practice, 815-832
- **J. Gomes et al.:** *Warping and Morphing of Graphical Objects*, Course Notes - SIGGRAPH'95