

# Variational Restoration of Images with Space-Variant Blur

---

Leah Bar

Department of Electrical Engineering

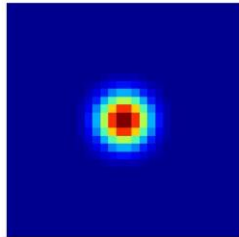
University of Minnesota



# Space-Invariant Model

- Image is degraded by blur and random (noise) processes.
- Blur is assumed as a linear shift invariant process with additive noise.

$$f * h + n = g$$



# Space Invariant Model – Variational Approach

Total Variation

$$F(f) = \int_{\Omega} (g - h^* f)^2 dx + \alpha \int_{\Omega} |\nabla f| dx$$

fitting term                      regularization

Rudin, Osher, Fetami 1992

# Space Invariant Model – Variational Approach

Total Variation

$$F(f) = \int_{\Omega} (g - h^* f)^2 dx + \alpha \int_{\Omega} |\nabla f| dx$$

fitting term

regularization

Rudin, Osher, Fetami 1992

$\Omega$ : image domain     $K$ : edge set     $f$ : recovered image     $g$ : observed image     $h$ : blur kernel

Mumford-Shah

$$F(f, K) = \int_{\Omega} (g - h^* f)^2 dx + \beta \int_{\Omega \setminus K} |\nabla f|^2 dx + \alpha \int_K d\sigma$$

gradients within  
segments

total edge length

Mumford and Shah, 1985

Bar, Sochen, Kiryati 2004

# Space Invariant Model – Variational Approach

Total Variation

$$F(f) = \int_{\Omega} (g - h^* f)^2 dx + \alpha \int_{\Omega} |\nabla f| dx$$

Rudin, Osher, Fetami 1992

fitting term

regularization

$\Omega$ : image domain     $K$ : edge set     $f$ : recovered image     $g$ : observed image     $h$ : blur kernel

Mumford-Shah

$$F(f, K) = \int_{\Omega} (g - h^* f)^2 dx + \beta \int_{\Omega \setminus K} |\nabla f|^2 dx + \alpha \int_K d\sigma$$

Mumford and Shah, 1985

Bar, Sochen, Kiryati 2004

gradients within  
segments

total edge length

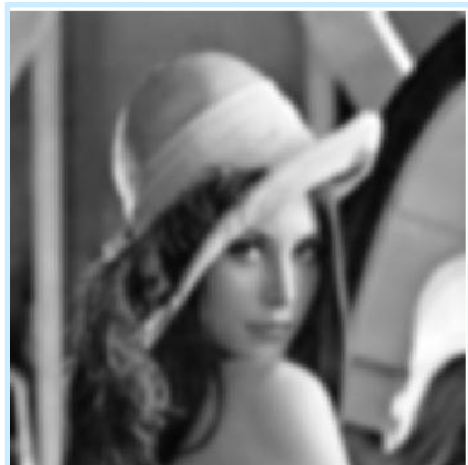
$\Gamma$ -convergence    Ambrosio and Tortorelli, 1990

$$F_{\varepsilon}(f, v) = \int_{\Omega} (g - h^* f)^2 dx + \beta \int_{\Omega} v^2 |\nabla f|^2 dx + \alpha \int_{\Omega} \left( \varepsilon |\nabla v|^2 + \frac{(v-1)^2}{4\varepsilon} \right) dx$$

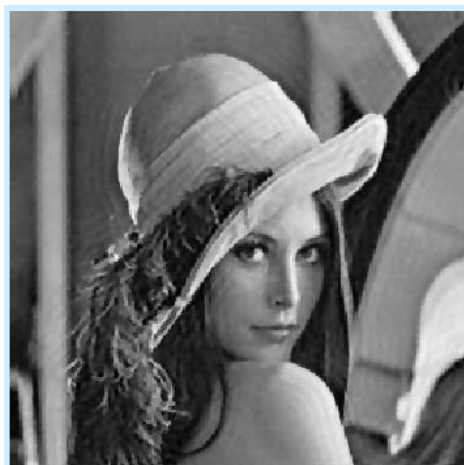
$v(x) \sim 0$  at edges     $v(x) \sim 1$  otherwise (in segments)

# Deconvolution with Mumford-Shah Regularization

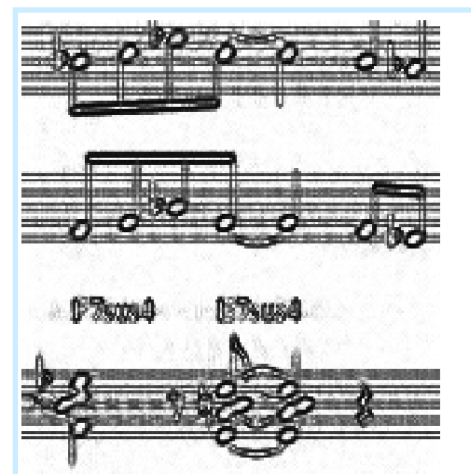
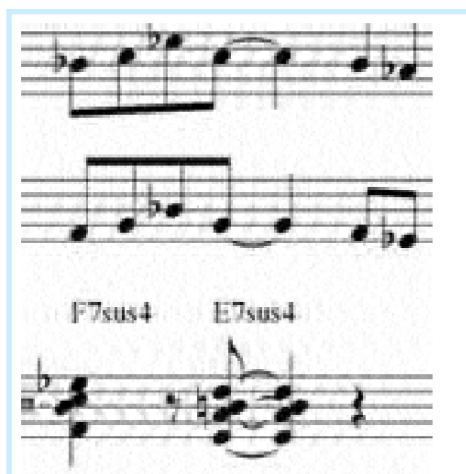
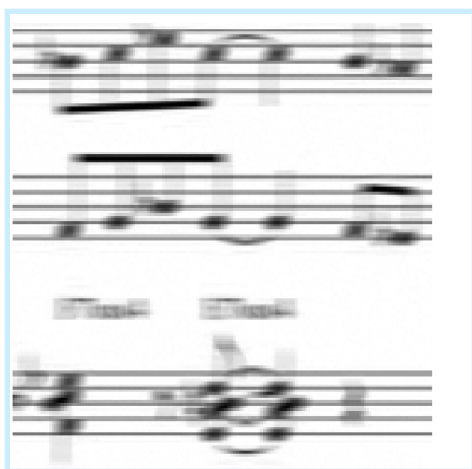
blurred



suggested restoration



computed edges ( $v$ )



# Space-Variant Image Restoration



- Assuming a piecewise point spread function
- Segmentation of sub-regions
- Space-variant deblurring

# Out of Focus Blur

$f_{fg}$  (closer)



$f_{bg}$



$$g = (f_{fg} * h) \cdot (1 - \chi_{bg} * h) + f_{bg} \cdot (\chi_{bg} * h)$$

$$g = f_{fg} \cdot \chi_{obj} + (f_{bg} * h) \cdot (1 - \chi_{obj})$$

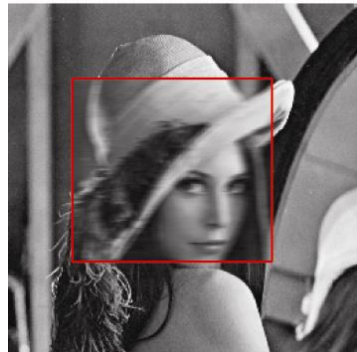
# Motion Blur



$$g = (f_{fg} \cdot \chi_{obj}) * h + f_{bg} (1 - h * \chi_{obj})$$

# Non-Blind Piecewise Deblurring

L. Bar, N. Sochen, N. Kiryati, SSVN 2007



Space-variant blur



Failure of region-wise deblurring

# Non-Blind Piecewise Deblurring

L. Bar, N. Sochen, N. Kiryati, SSVN 2007

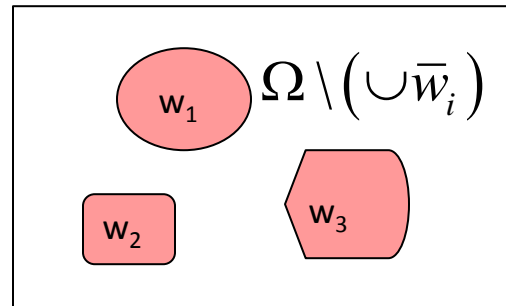


Space-variant blur



Failure of region-wise deblurring

Let:  $w_i$  = sub-region set  
 $h_i$  = kernels set  
 $h_b$  = background kernel

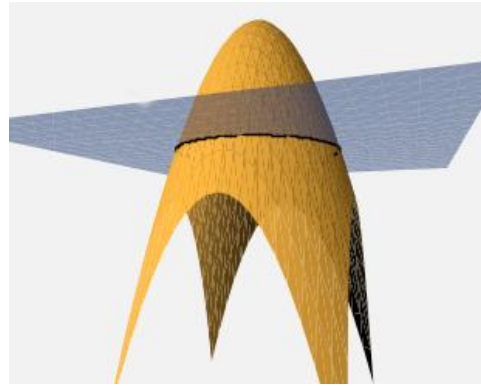


$$F(f) = \frac{1}{2} \sum_i \int_{w_i} (h_i * f - g)^2 dx + \frac{1}{2} \int_{\Omega \setminus (\cup \bar{w}_i)} (h_b * f - g)^2 dx + R^{MS}$$

Fitting term

Mumford-Shah regularization

The contour is implicitly represented by the zero level-set of a function  $\phi$



$$C = \partial\omega = \{(x, y) \in \Omega : \phi(x, y) = 0\}$$

$$\text{inside}(C) = \omega = \{(x, y) \in \Omega : \phi(x, y) > 0\}$$

$$\text{outside}(C) = \Omega \setminus \omega = \{(x, y) \in \Omega : \phi(x, y) < 0\}$$

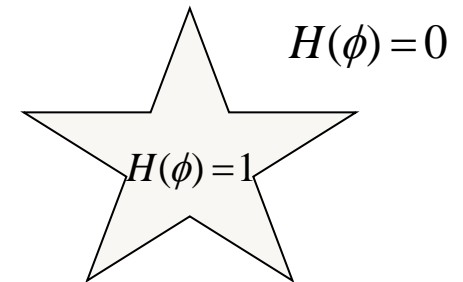
Heaviside function

$$H(s) = \begin{cases} 1 & s > 0 \\ 0 & \text{else} \end{cases}$$



$$\text{inside}(C) = \omega = \{(x, y) \in \Omega : H(\phi) = 1\}$$

$$\text{outside}(C) = \Omega \setminus \omega = \{(x, y) \in \Omega : H(\phi) = 0\}$$



$w_i$ : sub domain     $f$ : recovered image     $g$ : observed image     $h_i$ : blur kernels

$$\frac{1}{2} \sum_i \int_{w_i} (h_i * f - g)^2 dx + \frac{1}{2} \int_{\Omega \setminus (\cup \bar{w}_i)} (h_b * f - g)^2 dx + R^{MS}$$

Using level-set and  $\Gamma$ -convergence formulation

$$F_\varepsilon^{MS}(f, v) = \frac{1}{2} \sum_i \int_{\Omega} (h_i * f - g)^2 H(\phi_i) dx + \frac{1}{2} \int_{\Omega} (h_b * f - g)^2 \left[ 1 - \sum_i H(\phi_i) \right] dx$$

$$+ \beta \int_{\Omega} v^2 |\nabla f|^2 dx + \alpha \int_{\Omega} \left( \varepsilon |\nabla v|^2 + \frac{(v-1)^2}{4\varepsilon} \right) dx$$

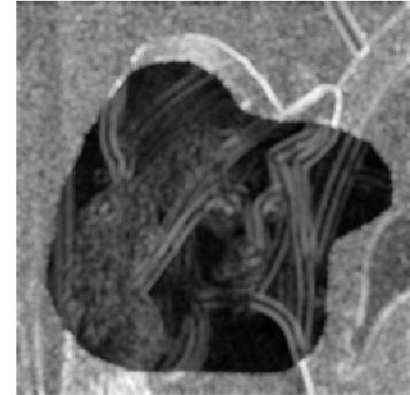
# Non-Blind Piecewise Deblurring - Results



# Semi-Blind Space Variant Restoration

Blur detection function

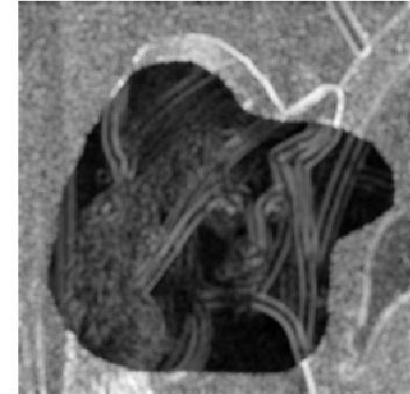
$$E(x) := \left| \log \left( \left| \nabla^2 g(x) \right| * B_r \right) \right|$$



# Semi-Blind Space Variant Restoration

Blur detection function

$$E(x) := \left| \log \left( \left| \nabla^2 g(x) \right| * B_r \right) \right|$$



Find a contour  $C$  and four constants  $c_1, c_2, c_3, c_4$  such that

$$F(C, \vec{c}) = \frac{\lambda_1}{2} \int_{\text{inside}(C) \cap K} (E - c_1)^2 dx + \frac{\lambda_2}{2} \int_{\text{outside}(C) \cap K} (E - c_2)^2 dx$$

$$+ \frac{\lambda_3}{2} \int_{\text{inside}(C) \setminus K} (E - c_3)^2 dx + \frac{\lambda_4}{2} \int_{\text{outside}(C) \setminus K} (E - c_4)^2 dx$$

$$+ \oint_C F_{GAC}(|\nabla E|) ds$$

$$F_{GAC} = \frac{\mu}{1 + (V_T^2 |\nabla E|^2) / \gamma} + \nu$$

# Semi-Blind Space Variant Restoration

observed



recovered



blur detection function

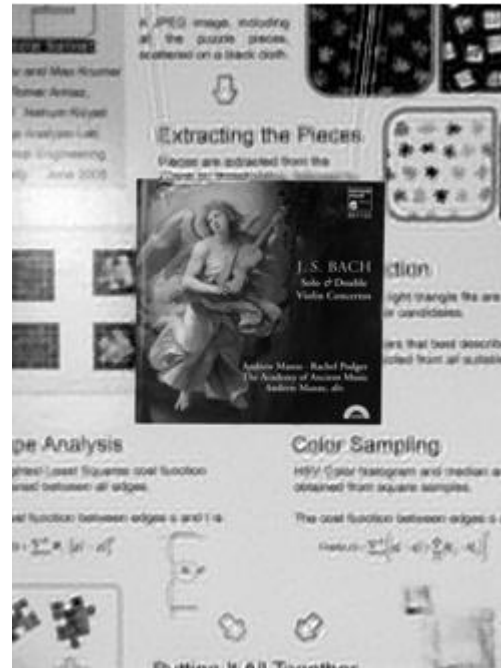


# Semi-Blind Space Variant Restoration

observed



recovered

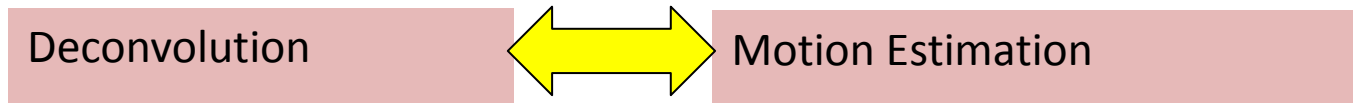
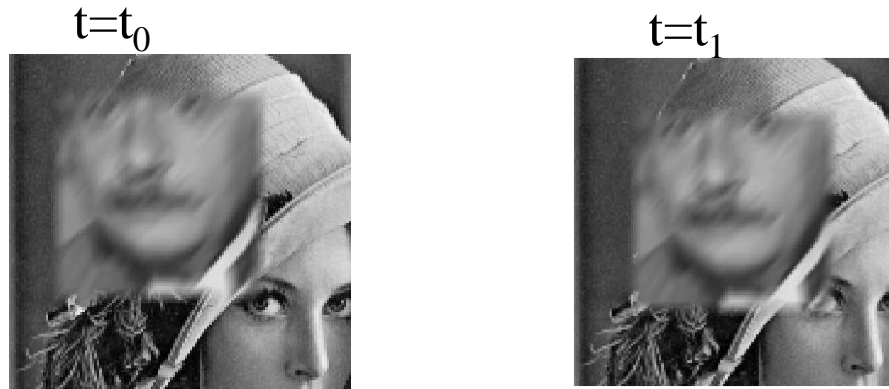


blur detection function



# Motion Estimation and Restoration of Blurred Video

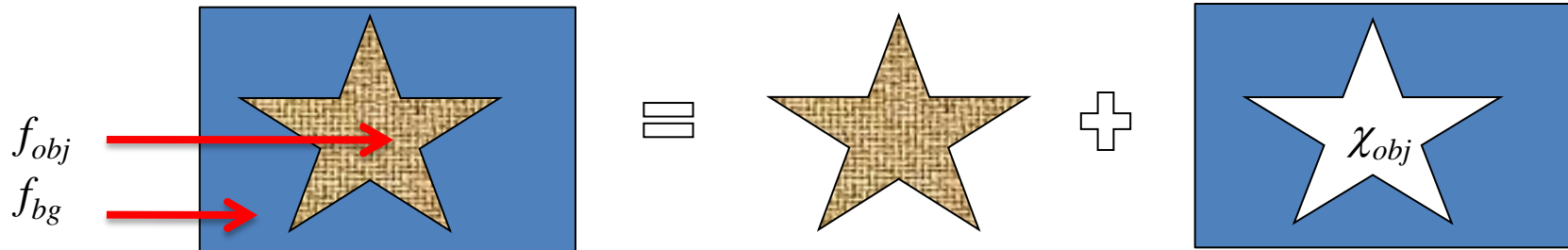
L. Bar, B. Berkels, G. Sapiro, M. Rumpf, ICCV 2007



Given 2 or more video frames: estimate the

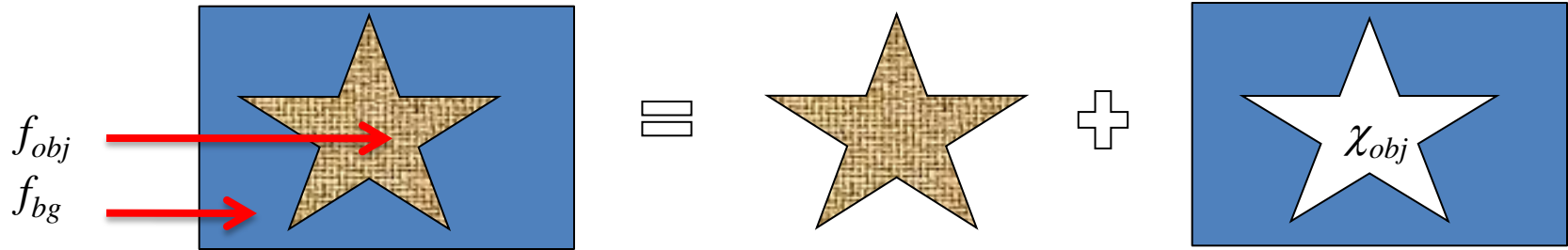
- velocity of the moving object
- sharp version of the object
- object characteristic function (segmentation)

# Blur Formation



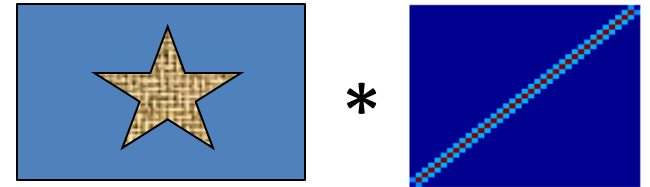
$$f(t, x) = f_{obj}(x - vt)\chi_{obj}(x - vt) + f_{bg} [1 - \chi_{obj}(x - vt)]$$

# Blur Formation



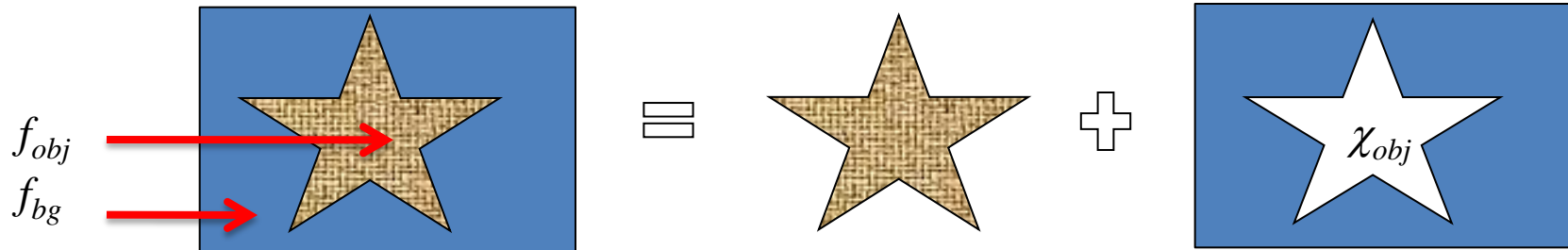
$$f(t, x) = f_{obj}(x - vt)\chi_{obj}(x - vt) + f_{bg} [1 - \chi_{obj}(x - vt)]$$

$$g_i(x) = \frac{1}{T} \int_{t_i - T/2}^{t_i + T/2} f(x - sv) ds = f * h_v(x)$$



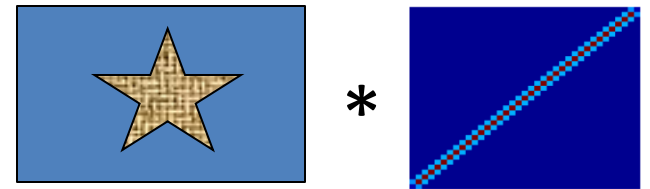
$$h_v(x) = \frac{1}{T\|v\|} \delta(v \cdot x^\perp) H(T\|v\| - 2\|x\|)$$

# Blur Formation



$$f(t, x) = f_{obj}(x - vt)\chi_{obj}(x - vt) + f_{bg} [1 - \chi_{obj}(x - vt)]$$

$$g_i(x) = \frac{1}{T} \int_{t_i - T/2}^{t_i + T/2} f(x - sv) ds = f * h_v(x)$$



$$h_v(x) = \frac{1}{T\|v\|} \delta(v \cdot x^\perp) H(T\|v\| - 2\|x\|)$$

$$g_i(t, x) = \left( (f_{obj}\chi_{obj}) * h_v \right) (x - vt_i) + f_{bg} [1 - (\chi_{obj} * h_v)(x - vt_i)]$$



# Objective Functional

$$\mathcal{E}(\chi_{obj}, f_{obj}, \nu) =$$

$$\sum_{i=1}^2 \int_{\Omega} \left\{ \left( (f_{obj} \chi_{obj}) * h_{\nu} \right) (x - t_i \nu) + f_{bg}(x) \left[ 1 - (\chi_{obj} * h_{\nu})(x - t_i \nu) \right] - g_i \right\}^2 dx + \int_{\Omega} \beta |\nabla f_{obj}| + \nu \int_{\Omega} |\partial \chi_{obj}| dx$$

$$\chi_{obj} = \text{Heaviside}(\phi)$$



$$\mathcal{E}(\phi, f_{obj}, \nu) =$$

$$\sum_{i=1}^2 \int_{\Omega} \left\{ \left( f_{obj} H(\phi) * h_{\nu} \right) (x - t_i \nu) + f_{bg}(x) \left[ 1 - (H(\phi) * h_{\nu})(x - t_i \nu) \right] - g_i \right\}^2 dx + \int_{\Omega} \beta |\nabla f_{obj}| + \nu \int_{\Omega} |\nabla H(\phi)| dx$$

How to optimize?



# Initialization

Assume that there is no blur

$$\mathcal{E}_{init}(\phi, \nu) =$$

$$\sum_{i=1}^2 \int_{\Omega} \left\{ (f_{obj} H(\phi))(x - t_i \nu) + f_{bg}(x) [1 - H(\phi)(x - t_i \nu)] - g_i \right\}^2 dx + \nu \int_{\Omega} |\nabla H(\phi)| dx$$

$$f_{obj} = g_1, \quad \phi^0 = \text{cone}, \quad k = 0$$

do{

$$\phi^{k+1} = \phi^k - \tau^{\phi} \nabla_{\phi} \mathcal{E}_{init}[\phi^k, \nu^k] * G_{\sigma}$$

$$\nu^{k+1} = \nu^k - \tau^{\nu} \nabla_{\nu} \mathcal{E}_{init}[\phi^{k+1}, \nu^k] * G_{\sigma}$$

$k = k + 1$

}while  $(\|\phi^{k+1} - \phi^k\|, \|\nu^{k+1} - \nu^k\| > \delta)$

  $\phi^{init}, \nu^{init}$

# Main Loop

$$\mathcal{E}(\phi, f_{obj}, v) =$$

$$\sum_{i=1}^2 \int_{\Omega} \left\{ (f_{obj} H(\phi) * h_v)(x - t_i v) + f_{bg}(x) [1 - (H(\phi) * h_v)(x - t_i v)] - g_i \right\}^2 dx + \int_{\Omega} \beta |\nabla f_{obj}| + v \int_{\Omega} |\nabla H(\phi)| dx$$

$$f_{obj}^0 = g_1, \quad \phi^{init}, \quad v^{init} \quad k = 0$$

do{

$$\phi^{k+1} = \phi^k - \tau^{\phi} \nabla_{\phi} \mathcal{E}[\phi^k, v^k, f_{obj}^k] * G_{\sigma}$$

$$v^{k+1} = v^k - \tau^v \nabla_v \mathcal{E}[\phi^{k+1}, v^k, f_{obj}^k] * G_{\sigma}$$

$$f_{obj}^{k+1} = f_{obj}^k - \tau^f \nabla_f \mathcal{E}[\phi^{k+1}, v^{k+1}, f_{obj}^k] * G_{\sigma}$$

$$k = k + 1$$

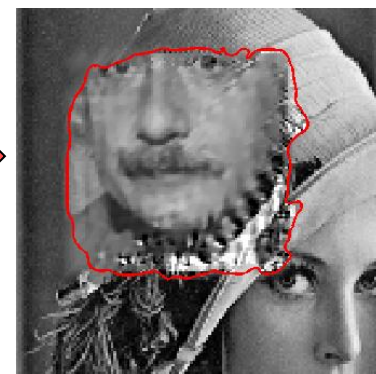
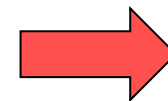
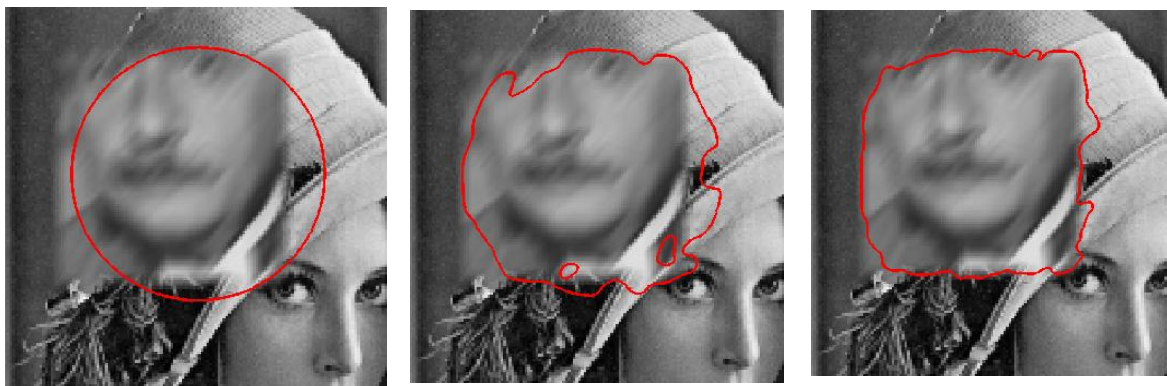
$$\} \text{while } \left( \|\phi^{k+1} - \phi^k\|, \|v^{k+1} - v^k\|, \|f_{obj}^{k+1} - f_{obj}^k\| > \delta \right)$$

  $\phi, v, f_{obj}$

# Motion Estimation and Restoration of Blurred Video - Results

Synthetic but challenging example

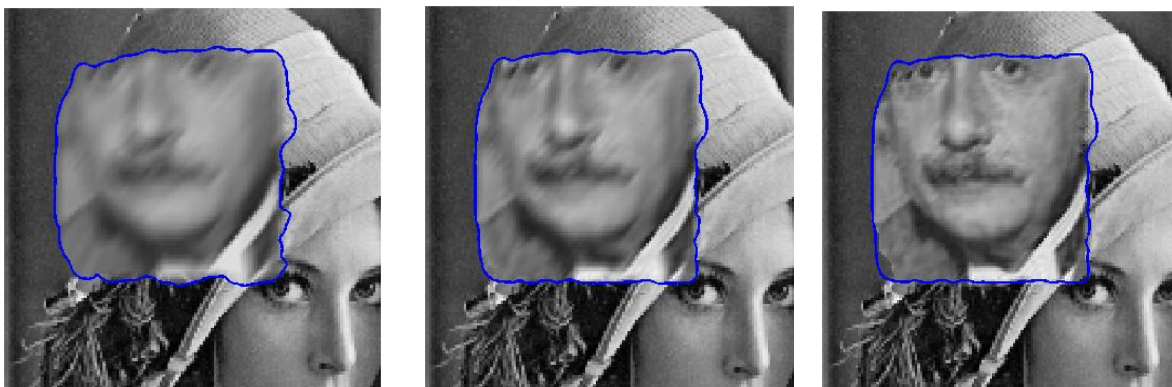
Initialization



ground truth:  $v=[6,7]$

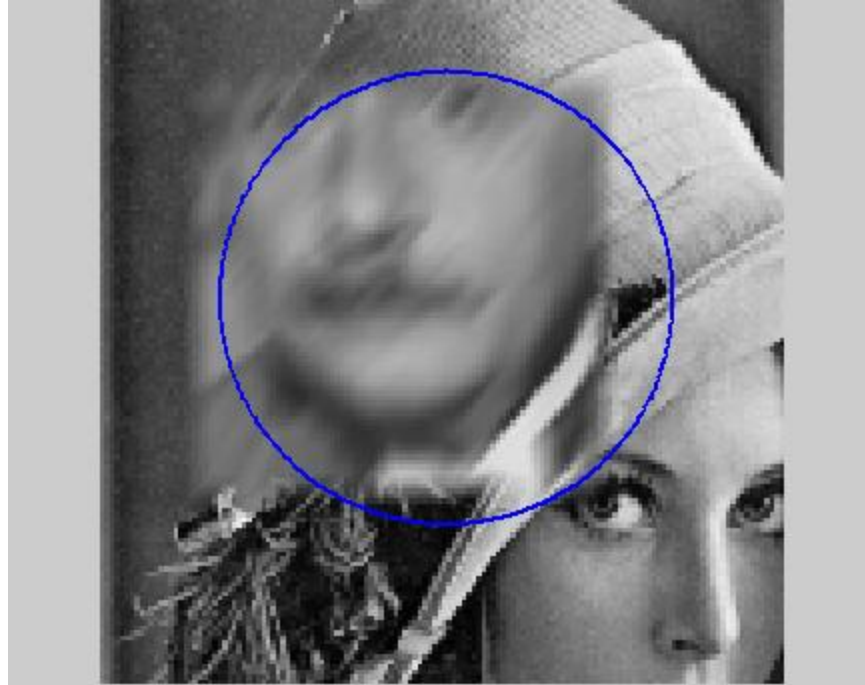
wrong model  
 $v=[5.78,6.8]$

Main loop



right model  
 $v=[5.98,7.009]$

# Motion Estimation and Restoration of Blurred Video - Results



$$v_{\text{truth}} = [6,7]$$

$$v_{\text{init}} = [5.78,6.8]$$

$$v_{\text{final}} = [5.98,7.009]$$

# Motion Estimation and Restoration of Blurred Video - Results



Ground truth:  $v=[10,0]$

wrong model:  
 $v=[9.46, -0.0213]$



right model:  
 $v=[9.48, -0.007]$



# Motion Estimation and Restoration of Blurred Video - Results



$$v_{\text{truth}} = [10, 0]$$

$$v_{\text{init}} = [9.46, -0.0213]$$

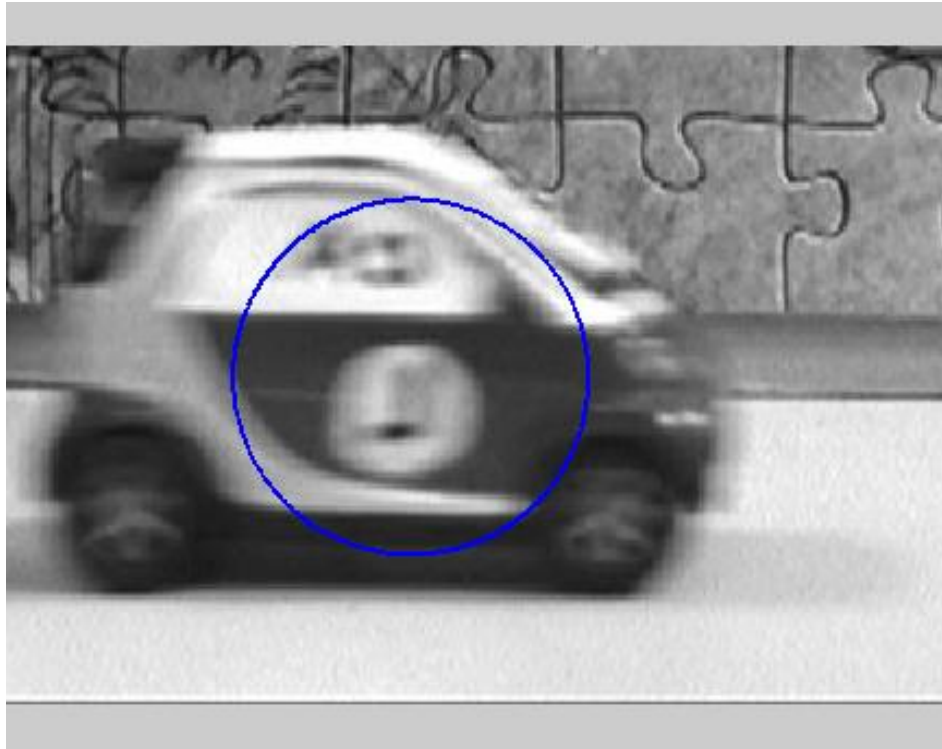
$$v_{\text{final}} = [9.48, -0.007]$$

# Motion Estimation and Restoration of Blurred Video - Results



Original movie

# Motion Estimation and Restoration of Blurred Video - Results

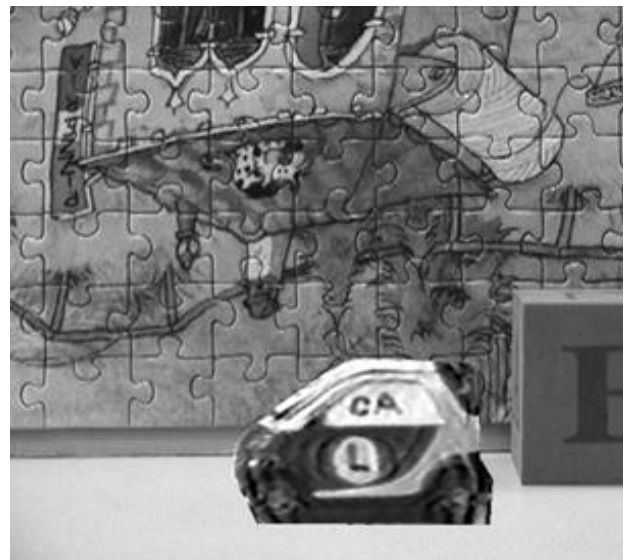
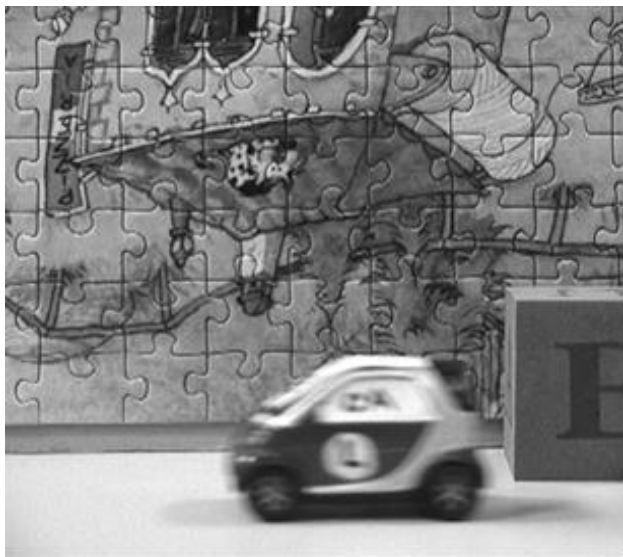


$$v_{\text{truth}} = ?$$

$$v_{\text{init}} = [18.67, 0.0588]$$

$$v_{\text{final}} = [18.66, 0.0473]$$

# Motion Estimation and Restoration of Blurred Video - Results



# Conclusions

- Space variant restoration is an ongoing and challenging research field
- A piecewise point spread function was used for out of focus and motion blur
- Promising results for synthetic/real images if using the physical model of the blur formation.

Thank you for your attention!

