

You have 60 minutes for the exam. It contains 4 sections, each is worth 15 points. Write your answers on a separate piece of paper.

MIPIA Test Exam

1 Gaussian Filtering

Question 1.

15 P.

A Gaussian filter with zero mean and the standard deviation σ is given as

$$g_{\sigma}(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{x^2}{2\sigma^2}\right)$$

and its Fourier transform as

$$G_{\sigma}(f) = \exp\left(-\frac{(\sigma 2\pi f)^2}{2}\right).$$

Show that the chaining of two Gaussian filters, using the standard deviations σ_1 and σ_2 respectively, is equivalent to one Gaussian filter using the standard deviation $\sigma_3 = \sqrt{\sigma_1^2 + \sigma_2^2}$

$$(g_{\sigma_1} * g_{\sigma_2})(x) = g_{\sigma_3}(x).$$

Solution 1.

$$(g_{\sigma_1} * g_{\sigma_2})(x) = F^{-1}(G_{\sigma_1}(f) \cdot G_{\sigma_2}(f)) \quad (1)$$

$$= F^{-1}\left(\exp\left(-\frac{\sigma_1^2(2\pi f)^2}{2}\right) \cdot \exp\left(-\frac{\sigma_2^2(2\pi f)^2}{2}\right)\right) \quad (2)$$

$$= F^{-1}\left(\exp\left(-\frac{\sigma_1^2(2\pi f)^2}{2} - \frac{\sigma_2^2(2\pi f)^2}{2}\right)\right) \quad (3)$$

$$= F^{-1}\left(\exp\left(-\frac{\sigma_1^2(2\pi f)^2 + \sigma_2^2(2\pi f)^2}{2}\right)\right) \quad (4)$$

$$= F^{-1}\left(\exp\left(-\frac{(\sigma_1^2 + \sigma_2^2)(2\pi f)^2}{2}\right)\right) \quad (5)$$

$$= F^{-1}\left(\exp\left(-\frac{(\sqrt{\sigma_1^2 + \sigma_2^2})^2(2\pi f)^2}{2}\right)\right) \quad (6)$$

$$= F^{-1}(G_{\sigma_3}(f)) \quad (7)$$

$$= g_{\sigma_3}(x) \quad (8)$$

2 Structure Tensor

Applying the tensor product to the gradients of an image f yields the structure tensor

$$\mathbf{J} = \nabla f (\nabla f)^T = \begin{pmatrix} f_x \\ f_y \end{pmatrix} (f_x, f_y) = \begin{pmatrix} f_x^2 & f_x f_y \\ f_x f_y & f_y^2 \end{pmatrix}$$

with f_x and f_y being the derivatives of f in x and y direction respectively.

Spatial averaging of the individual components of \mathbf{J} with the Gaussian kernel K_ϱ will result in the structure tensor

$$\mathbf{J}_{\varrho, \sigma} = K_\varrho * (\nabla f_\sigma \otimes \nabla f_\sigma)$$

with

$$\nabla f_\sigma = (\nabla K_\sigma) * f.$$

In this context, the standard deviations ϱ and σ act as regularization parameters.

Question 2.

7.5 P.

$\lambda_1, \lambda_2 \in \mathbb{R}$ with $\lambda_1 \geq \lambda_2$ are the eigenvalues and $\mathbf{v}_1, \mathbf{v}_2 \in \mathbb{R}^2$ are the eigenvectors of the structure tensor $\mathbf{J}_{\varrho, \sigma} \in \mathbb{R}^{2 \times 2}$ of an image f . Which conditions apply to the eigenvalues λ_1 and λ_2 when $\mathbf{J}_{\varrho, \sigma}$ denotes

1. a flat area?
2. a straight edge?
3. a corner?

Solution 2.

For the eigenvalues λ_1, λ_2 of a structure tensor $\mathbf{J}_{\varrho, \sigma}$ the following applies in the case of a

1. flat area: $\lambda_1 \approx \lambda_2 \approx 0$
2. straight edge: $\lambda_1 \gg \lambda_2 \approx 0$
3. corner: $\lambda_1 \geq \lambda_2 \gg 0$

Question 3.

7.5 P.

Figure 1-a shows an image of a fingerprint and figures b), c) and d) are showing the direction of the eigenvectors of the structure tensors of figure 1-a with varying parameters. The mathematical formulation of the structure tensor is given in the equations above.

1. What change of parameters causes the differences between figures 1-b, 1-c and 1-d?
2. What can you say about the changed parameters between figures 1-b, 1-c and 1-d, where are they increased and where are they decreased?

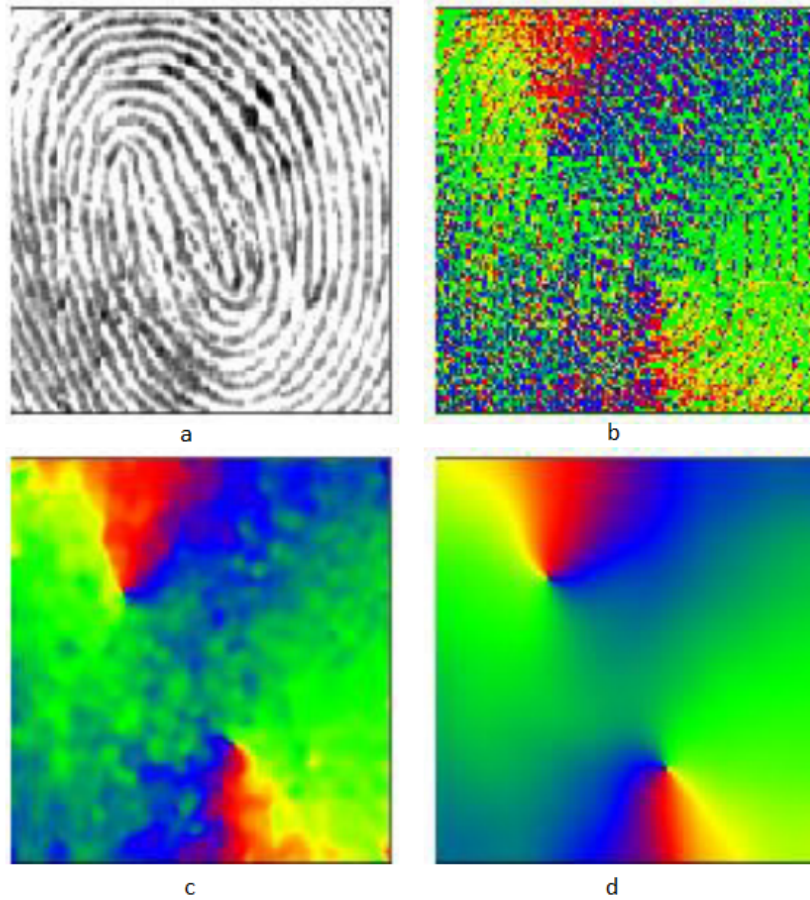


Figure 1: Subfigure a shows an image of a finger print and figures b, c and d are the computed structure tensor of image a with different parameter(s) (image: Joachim Weickert).

Solution 3.

The standard deviations ϱ and σ for the structure tensor $\mathbf{J}_{\varrho,\sigma}$ are the regularization parameters. Therefore, change in these parameters can result in differently localized structure tensors in figure 3-b, 3-c and 3-d.

For the sake of comparison, consider the σ to be the same in 3-b, 3-c and 3-d, we will have $\varrho_b < \varrho_c < \varrho_d$.

3 Epipolar Geometry

Question 4.

5 P.

Label the *epipole(s)*, *epipolar line(s)*, *epipolar plane(s)* and *baseline(s)* in figure 2.

Solution 4.

Epipole(s), epipolar line(s), epipolar plane(s) and baseline(s) are shown in figure 3.

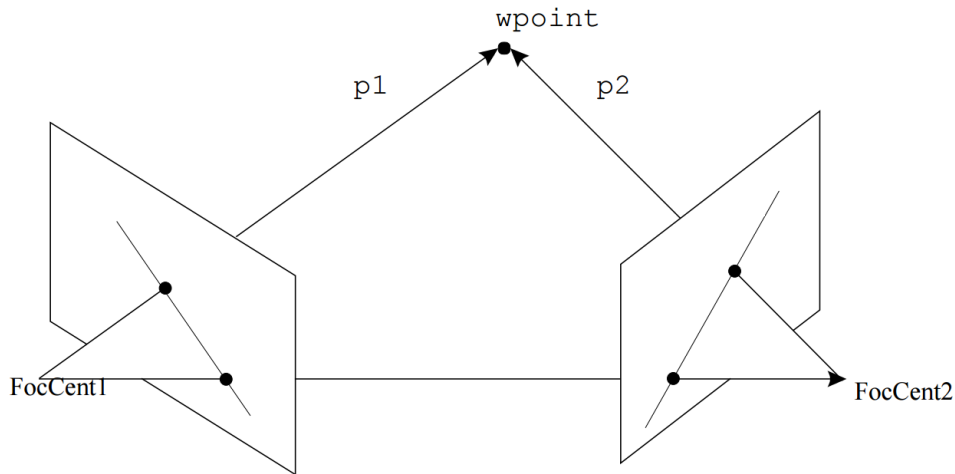


Figure 2: Epipolar geometry.

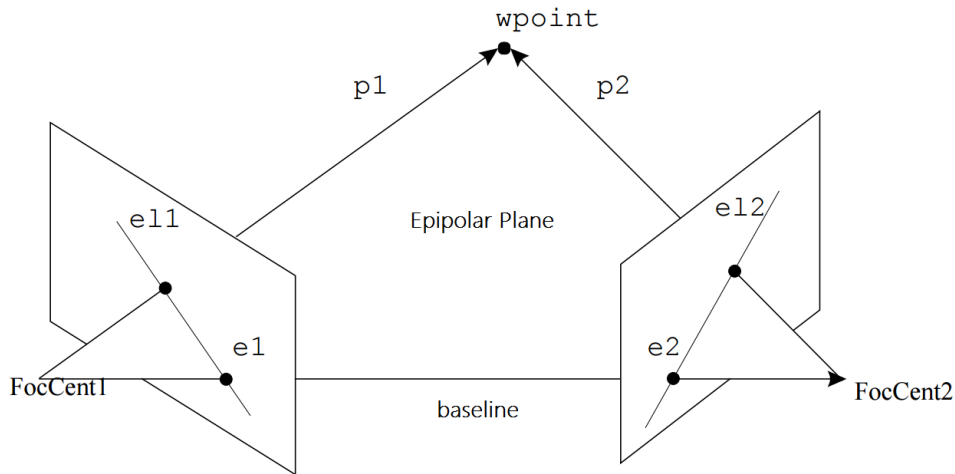


Figure 3: Epipolar geometry.

Question 5.

10 P.

1. Describe the *epipolar constraint*.
2. Consider a 3D world point $\mathbf{w} \in \mathbb{R}^3$. \mathbf{w} is mapped to $\mathbf{p} \in \mathbb{R}^3$ in the left image and to $\mathbf{q} \in \mathbb{R}^3$ in the right image. \mathbf{p}^c and \mathbf{q}^c denote the points corresponding to the world point \mathbf{w} in 3-D camera coordinates. $\mathbf{t} \in \mathbb{R}^3$ denotes the translation vector and $\mathbf{R} \in \mathbb{R}^{3 \times 3}$ denotes the rotation matrix.

Derive the epipolar constraint, i.e. the essential matrix \mathbf{E} mathematically. You can use the the equations below as a starting point.

- $\mathbf{q}^c = \mathbf{R}(\mathbf{p}^c - \mathbf{t})$.
- \mathbf{p}^c , $\mathbf{p}^c - \mathbf{t}$ and \mathbf{t} lie on the same plane, i.e. $(\mathbf{p}^c - \mathbf{t})^T (\mathbf{t} \times \mathbf{p}^c) = 0$.

Solution 5.

1. Projection of a 3D world point \mathbf{w} on the left/right image plane lies on the epipolar line on the left/right image plane.
- 2.

$$\mathbf{q}^c = \mathbf{R}(\mathbf{p}^c - \mathbf{t}) \Rightarrow \mathbf{R}^T \mathbf{q}^c = \mathbf{p}^c - \mathbf{t} \quad (9)$$

\mathbf{p}^c , $\mathbf{p}^c - \mathbf{t}$ and \mathbf{t} lie on the same plane, i.e.

$$(\mathbf{p}^c - \mathbf{t})^T (\mathbf{t} \times \mathbf{p}^c) = 0. \quad (10)$$

We then get

$$(\mathbf{R}^T \mathbf{q}^c)^T (\mathbf{t} \times \mathbf{p}^c) = \mathbf{q}^{c^T} \mathbf{R} (\mathbf{t} \times \mathbf{p}^c) = \mathbf{q}^{c^T} \underbrace{\mathbf{R}[\mathbf{t}]_{\times}}_{\mathbf{E}} \mathbf{p}^c = \mathbf{q}^{c^T} \mathbf{E} \mathbf{p}^c \quad (11)$$

with \mathbf{E} being the essential matrix.

3. Same solution as the second part.

4 Variational Calculus

Question 6.

15 P.

The Euler-Lagrange equation

$$\frac{\delta}{\delta \mathbf{f}} F(x, \mathbf{f}(x), \mathbf{f}'(x)) - \frac{d}{dx} \frac{\delta}{\delta \mathbf{f}'} F(x, \mathbf{f}(x), \mathbf{f}'(x)) = 0$$

is satisfied for the functional

$$I(\mathbf{f}) = \int_{x_1}^{x_2} F(x, \mathbf{f}(x), \mathbf{f}'(x)) dx$$

with $x_1, x_2 \in \mathbb{R}^2$ and \mathbf{f}' being the first degree derivative of \mathbf{f} , when $\mathbf{f}(x)$ is a minimum for $I(\mathbf{f})$.

The integral

$$L(c) = \int_a^b \sqrt{1 + (c'(x))^2} dx$$

gives the length for the curve described by the function $c(x) \in \mathbb{R}$ between the points $a, b \in \mathbb{R}^2$.

1. Minimize L with respect to c to find the minimal function c_0 .
2. How can c_0 be interpreted?

Solution 6.

We try to find the minimum c_0 for the functional L so we apply the Euler-Lagrange equation to L . The integrand from L

$$L(c) = \int_a^b \underbrace{\sqrt{1 + (c'(x))^2}}_{\text{integrand}} dx \quad (12)$$

is now

$$F(x, c(x), c'(x)) = \sqrt{1 + (c'(x))^2}. \quad (13)$$

The derivatives of F with respect to $c(x)$ and $c'(x)$ are

$$\frac{\delta}{\delta c} F(x, c(x), c'(x)) = 0 \quad (14)$$

$$\frac{\delta}{\delta c'} F(x, c(x), c'(x)) = \frac{c'(x)}{\sqrt{1 + (c'(x))^2}}. \quad (15)$$

These are inserted into the Euler-Lagrange equation

$$\frac{\delta}{\delta c} F(x, c(x), c'(x)) - \frac{d}{dx} \frac{\delta}{\delta c'} F(x, c(x), c'(x)) = 0 \quad (16)$$

$$0 - \frac{d}{dx} \frac{c'(x)}{\sqrt{1 + (c'(x))^2}} = 0. \quad (17)$$

Since the derivative above is 0, its parent function must be a constant

$$k = \frac{c'(x)}{\sqrt{1 + (c'(x))^2}}. \quad (18)$$

This can be solved for $c'(x)$ by

$$c'(x) = k \sqrt{1 + (c'(x))^2} \quad (19)$$

$$(c'(x))^2 = k^2 (1 + (c'(x))^2) = k^2 + k^2 (c'(x))^2 \quad (20)$$

$$(c'(x))^2 - k^2 (c'(x))^2 = k^2 \quad (21)$$

$$(c'(x))^2 (1 - k^2) = k^2 \quad (22)$$

$$(c'(x))^2 = \frac{k^2}{(1 - k^2)} \quad (23)$$

$$c'(x) = \frac{k}{\sqrt{1 - k^2}}. \quad (24)$$

$c'(x)$ is a constant which leads to its parent function

$$c_0(x) = \frac{k}{\sqrt{1 - k^2}} x + t = mx + t \quad (25)$$

with $m, t \in \mathbb{R}$ being constants. The optimum $c_0(x)$ describes a line, which is the shortest connection between the points a and b .