

## EXAMINATION MATHEMATICAL TECHNIQUES FOR IMAGE ANALYSIS

Course code: 2DMM10 & 8D020. Date: Wednesday January 27 2016. Time: 13:30–16:30. Place: AUD 12.

### Read this first!

- Use a separate sheet of paper for each problem. Write your name and student ID on each paper.
- The exam consists of 5 problems. The maximum credit for each item is indicated in the margin.
- The use of course notes is allowed, provided it is in immaculate state. The use of problem companion (“opgaven- en tentamenbundel”), calculator, laptop, smartphone, or any other equipment, is *not* allowed.
- Motivate your answers. You may provide your answers in Dutch or English.

*GOOD LUCK!*

### (25) 1. GROUP

**Definition 1** A *group* is a collection  $G$  together with an internal operation

$$\circ : G \times G \longrightarrow G : (x, y) \mapsto x \circ y,$$

such that

- the operation is associative, i.e.  $(x \circ y) \circ z = x \circ (y \circ z)$  for all  $x, y, z \in G$ ,
- there exists an identity element  $e \in G$  such that  $x \circ e = e \circ x = x$  for all  $x \in G$ , and
- for each  $x \in G$  there exists an inverse element  $x^{\text{inv}}$  such that  $x^{\text{inv}} \circ x = x \circ x^{\text{inv}} = e$ .

**Definition 2** Recall Definition 1. If, in addition,  $x \circ y = y \circ x$  for all  $x, y \in G$ , then the group is called *commutative*, or *abelian*.

### (10) a. Prove the following properties for a group $G$ :

- The identity element  $e \in G$  is unique.
- The identity element equals its own inverse:  $e^{\text{inv}} = e$ .
- The inverse  $x^{\text{inv}}$  of a given element  $x \in G$  is unique.
- The inverse of the inverse of a given element  $x \in G$  reproduces that element:  $(x^{\text{inv}})^{\text{inv}} = x$ .

The identity element  $e \in G$  is unique: Suppose  $e_1 \neq e_2$  are both identity elements in  $G$ . Consider the identity  $x \circ e = e \circ x = x$ , which holds for all  $x \in G$ , twice, once for  $(e, x) = (e_1, e_2)$  and once for  $(e, x) = (e_2, e_1)$ . This yields  $e_2 \circ e_1 = e_1 \circ e_2 = e_2$ , but at the same time  $e_1 \circ e_2 = e_2 \circ e_1 = e_1$ , which is a contradiction.

The identity element equals its own inverse: Take  $x = e$  in the second axiom. This yields  $e \circ e = e$ , which, by the last axiom, defines  $e = e^{\text{inv}}$  to be its own inverse.

The inverse  $x^{\text{inv}}$  of a given element  $x \in G$  is unique: Suppose  $y \neq z$  are both inverses of  $x \in G$ . On the one hand we have  $y \circ x \circ z = y \circ (x \circ z) = y \circ e = y$ . But at the same time we have  $y \circ x \circ z = (y \circ x) \circ z = e \circ z = z$ . This is a contradiction.

The inverse of the inverse of a given element  $x \in G$  reproduces that element: For ease of notation, denote the inverse of  $x$  by  $y = x^{\text{inv}}$ . Then  $x = e \circ x = (y^{\text{inv}} \circ y) \circ x = y^{\text{inv}} \circ (y \circ x) = y^{\text{inv}} \circ e = y^{\text{inv}}$ , i.e.  $x = (x^{\text{inv}})^{\text{inv}}$ .

(15) **b.** Consider the set of  $4 \times 4$  matrices

$$G = \left\{ \left( \begin{array}{cc|cc} \mathbf{I} & -\Theta & & \\ \Theta & \mathbf{I} & & \\ \hline & & \Theta & \\ & & & \mathbf{I} \end{array} \right) \middle| \Theta \stackrel{\text{def}}{=} \begin{pmatrix} \theta_{11} & \theta_{12} \\ 0 & \theta_{22} \end{pmatrix}, \mathbf{I} \stackrel{\text{def}}{=} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \theta_{ij} \in \mathbb{R} \right\}.$$

Which constraint on the parameters  $\theta_{11}, \theta_{12}, \theta_{22}$  should we impose in order for  $G$  to define an abelian group under matrix multiplication?

Taking  $\Psi$  to be a matrix of the same form as  $\Theta$ , but with entries  $\psi_{ij}$  instead of  $\theta_{ij}$  ( $i, j = 1, 2$ ), we have

$$\left( \begin{array}{cc|cc} \mathbf{I} & -\Theta & & \\ \Theta & \mathbf{I} & & \\ \hline & & \mathbf{I} & -\Psi \\ & & \Psi & \mathbf{I} \end{array} \right) = \left( \begin{array}{cc|cc} \mathbf{I} - \Theta\Psi & -(\Theta + \Psi) & & \\ \Theta + \Psi & \mathbf{I} + \Theta\Psi & & \end{array} \right),$$

thus  $\Theta\Psi$  must be the  $2 \times 2$  null matrix. Note that the required constraint must be formulated in terms of a condition on the generic form of  $\Theta$ . Working this out for the given parametric form of  $\Theta$  and  $\Psi$  yields

$$\begin{pmatrix} \theta_{11} & \theta_{12} \\ 0 & \theta_{22} \end{pmatrix} \begin{pmatrix} \psi_{11} & \psi_{12} \\ 0 & \psi_{22} \end{pmatrix} = \begin{pmatrix} \theta_{11}\psi_{11} & \theta_{11}\psi_{12} + \theta_{12}\psi_{22} \\ 0 & \theta_{22}\psi_{22} \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}.$$

This must hold for *all* coefficients, whence (take  $\Theta = \Psi$  for instance) it follows that  $\theta_{11} = \theta_{22} = 0$ , leaving  $\theta_{12}$  undetermined, whence

$$\Theta \stackrel{\text{def}}{=} \begin{pmatrix} 0 & \theta \\ 0 & 0 \end{pmatrix},$$

with  $\theta \in \mathbb{R}$ .



(20) **2. METRIC SPACE**

**Definition 3** Let  $V$  be a vector space over  $\mathbb{K}$ . A *distance* on  $V$  is a nondegenerate positive definite symmetric mapping  $d : V \times V \rightarrow \mathbb{R}$  such that for all  $u, v, w \in V, \lambda \in \mathbb{K}$ ,

- $d(v, w) \geq 0$  and  $d(v, w) = 0$  if and only if  $v = w$ ,
- $d(v, w) = d(w, v)$ ,
- $d(v, w) \leq d(v, u) + d(u, w)$ .

We consider (2+1)-dimensional Euclidean spacetime  $M = \mathbb{R}^3 = \{(\vec{x}, t) \mid x = (x_1, x_2) \in \mathbb{R}^2 \text{ and } t \in \mathbb{R}\}$ , interpreted as a vector space with the usual definitions of vector addition and scalar multiplication, and furnished with a bivariate operator

$$d : M \times M \rightarrow \mathbb{R} : ((\vec{x}, t), (\vec{y}, u)) \mapsto d((\vec{x}, t), (\vec{y}, u)) \stackrel{\text{def}}{=} \begin{cases} \|\vec{y} - \vec{x}\| & \text{if } t = u, \\ |u - t| & \text{if } t \neq u, \end{cases}$$

in which  $\|\vec{y} - \vec{x}\| = \sqrt{(y_1 - x_1)^2 + (y_2 - x_2)^2}$ , i.e. the Euclidean distance between  $\vec{x}$  and  $\vec{y}$ . You may take it for granted that this Euclidean distance does indeed define a distance according to Definition 3.

In order to verify whether  $d$  defines a distance, please answer all problems under a, b and c below.

- (10) **a.** Prove or disprove the first axiom (positivity and nondegeneracy).  
(5) **b.** Prove or disprove the second axiom (symmetry).  
(5) **c.** Prove or disprove the third axiom (triangle inequality).

The mapping  $d$  fails to be a distance. Symmetry and positive nondegeneracy are fulfilled:

- a.  $d((\vec{x}, t), (\vec{y}, u)) \geq 0$ ; in particular equality clearly implies  $t = u$ , so that  $0 = d((\vec{x}, t), (\vec{y}, u)) = \|\vec{y} - \vec{x}\|$ , which holds iff  $\vec{x} = \vec{y}$ ;  
b.  $d((\vec{x}, t), (\vec{y}, u)) = d((\vec{y}, u), (\vec{x}, t))$  by virtue of the symmetries  $\|\vec{y} - \vec{x}\| = \|\vec{x} - \vec{y}\|$  and  $|u - t| = |t - u|$ .

However, the triangle inequality fails, for consider the distance between two distinct spacetime points for equal time  $t$ .

- c. On the one hand we have  $d((\vec{y}, t), (\vec{x}, t)) = \|\vec{y} - \vec{x}\| \stackrel{\text{def}}{=} \Delta > 0$ . If  $(\vec{z}, v)$  is any other spacetime point with  $v \neq t$ , then  $d((\vec{x}, t), (\vec{z}, v)) + d((\vec{z}, v), (\vec{y}, t)) = 2|v - t|$ . This may be smaller than  $\Delta$ , viz. if  $|v - t| < \Delta/2$ .



## (20) 3. DISTRIBUTION THEORY

**Definition 4** The class  $\mathcal{S}(\mathbb{R}^n)$  of *smooth test functions of rapid decay*,  $\phi : \mathbb{R}^n \rightarrow \mathbb{K}$ , a.k.a. *Schwartz functions*, is defined as follows:

$$\phi \in \mathcal{S}(\mathbb{R}^n) \quad \text{iff} \quad \phi \in C^\infty(\mathbb{R}^n) \quad \text{and} \quad \sup_{x \in \mathbb{R}^n} |x^\alpha \nabla_\beta \phi(x)| < \infty,$$

for all multi-indices  $\alpha$  and  $\beta$ .

**Theorem 1** *If  $T \in \mathcal{S}'(\mathbb{R}^n)$ , then there exist a constant  $c > 0$  and multi-indices  $\alpha, \beta$  such that*

$$|T[\phi]| \leq c \sup_{x \in \mathbb{R}^n} |x^\alpha \nabla_\beta \phi(x)|.$$

- (10) **a.** Prove Theorem 1 for the case of a regular tempered distribution  $T \stackrel{\text{def}}{=} T_f$ , i.e. a tempered distribution of the form

$$T_f[\phi] = \int_{\mathbb{R}^n} f(x) \phi(x) dx,$$

with  $f \in L^1(\mathbb{R}^n)$ .

Using the Hölder inequality (in †) we find

$$|T_f[\phi]| = \left| \int_{\mathbb{R}^n} f(x) \phi(x) dx \right| \leq \int_{\mathbb{R}^n} |f(x) \phi(x)| dx = \|f\phi\|_1 \leq \|f\|_1 \|\phi\|_\infty = c \sup_{x \in \mathbb{R}^n} |\phi(x)|,$$

in which  $c = \|f\|_1$ . Note that we may take trivial multi-indices in the estimation in this case:  $\alpha = \beta = 0 \in \mathbb{N}^n$ .

- (10) **b.** Show that it also holds if  $T \stackrel{\text{def}}{=}} \nabla_\alpha T_f$ , i.e. for any  $\alpha$ -partial derivative of a tempered distribution.

The space  $\mathcal{S}(\mathbb{R}^n)$  is closed under differentiation. This means that  $(-1)^\beta \nabla_\beta \phi \in \mathcal{S}(\mathbb{R}^n)$  for all multi-indices  $\beta$ . Since, by definition,  $\nabla_\beta T_f[\phi] = (-1)^\beta T_f[\nabla_\beta \phi]$  we have, using the conclusion of problem a,  $|\nabla_\beta T_f[\phi]| = |T_f[\nabla_\beta \phi]| \leq c \sup_{x \in \mathbb{R}^n} |\nabla_\beta \phi(x)|$  for some  $c \in \mathbb{R}$  and given  $\beta$ . Note that we may take  $\alpha$  trivial in Theorem 1 in this case.



**(20) 4. DISTRIBUTION THEORY & FOURIER ANALYSIS (EXAM APRIL 11, 2014, PROBLEM 4)**

The Fourier transform  $\widehat{T} \in \mathcal{S}'(\mathbb{R}^n)$  of a distribution  $T \in \mathcal{S}'(\mathbb{R}^n)$  is defined as follows:

$$\widehat{T}(\phi) = T(\widehat{\phi}) \quad \text{for all test functions } \phi \in \mathcal{S}(\mathbb{R}^n), \quad (\star)$$

in which

$$\widehat{\phi}(\omega) = \int_{\mathbb{R}^n} e^{-i\omega \cdot x} \phi(x) dx.$$

The purpose of this problem is to motivate this definition.

To this end, consider any function  $f : \mathbb{R}^n \rightarrow \mathbb{C}$  of polynomial growth for which the Fourier integral

$$\widehat{f}(\omega) = \int_{\mathbb{R}^n} e^{-i\omega \cdot x} f(x) dx$$

is well-defined and yields a function  $\widehat{f} : \mathbb{R}^n \rightarrow \mathbb{C}$  of polynomial growth. Denote by  $T_f \in \mathcal{S}'(\mathbb{R}^n)$  the corresponding *regular* tempered distribution, i.e.

$$T_f(\phi) = \int_{\mathbb{R}^n} f(x) \phi(x) dx$$

for all test functions  $\phi \in \mathcal{S}(\mathbb{R}^n)$ . It is then natural to define  $\widehat{T}_f \stackrel{\text{def}}{=} T_{\widehat{f}}$ , i.e.

$$\widehat{T}_f(\phi) \stackrel{\text{def}}{=} \int_{\mathbb{R}^n} \widehat{f}(\xi) \phi(\xi) d\xi. \quad (*)$$

- (10) **a.** Show that this definition (\*) implies  $\widehat{T}_f(\phi) = T_f(\widehat{\phi})$ , consistent with (\*), for any  $\phi \in \mathcal{S}(\mathbb{R}^n)$ .

(Hint: In (\*) apply the Fourier reconstruction formula in the following form:  $\phi(\xi) = \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{i\xi \cdot x} \widehat{\phi}(x) dx$ .)

Following the hint (in the quality marked by \*) we write

$$\widehat{T_f(\phi)} \stackrel{\text{def}}{=} T_{\widehat{f}}(\phi) \stackrel{\text{def}}{=} \int_{\mathbb{R}^n} \widehat{f}(\xi) \phi(\xi) d\xi \stackrel{*}{=} \int_{\mathbb{R}^n} \widehat{f}(\xi) \left( \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{i\xi \cdot x} \widehat{\phi}(x) dx \right) d\xi.$$

Interchanging the order of integration this is seen to be equivalent to

$$\widehat{T_f(\phi)} = \int_{\mathbb{R}^n} \left( \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} \widehat{f}(\xi) e^{i\xi \cdot x} d\xi \right) \widehat{\phi}(x) dx \stackrel{\circ}{=} \int_{\mathbb{R}^n} f(x) \widehat{\phi}(x) dx = T_f(\widehat{\phi}).$$

In  $\circ$  the Fourier reconstruction formula has been used once more, this time for the function  $f$ .

This result justifies the general definition (\*), which also holds if  $T \in \mathcal{S}'(\mathbb{R}^n)$  is not regular.

As an example, consider the (non-regular) Dirac point distribution  $\delta \in \mathcal{S}'(\mathbb{R}^n)$ , defined by  $\delta(\phi) = \phi(0)$  for all  $\phi \in \mathcal{S}(\mathbb{R}^n)$ .

- (10) **b.** Use the general definition (\*) to prove that  $\widehat{\delta} = T_1$ . Here  $T_1 \in \mathcal{S}'(\mathbb{R}^n)$  is the regular tempered distribution corresponding to the constant function  $1 : \mathbb{R}^n \rightarrow \mathbb{C} : x \mapsto 1(x) = 1$ .

Using the distributional definition of the Fourier transform (in \*) we find that

$$\widehat{\delta}(\phi) \stackrel{*}{=} \delta(\widehat{\phi}) \stackrel{\text{def}}{=} \widehat{\phi}(0) \stackrel{\dagger}{=} \int_{\mathbb{R}^n} \phi(x) dx \stackrel{\text{def}}{=} \int_{\mathbb{R}^n} 1(x) \phi(x) dx \stackrel{\text{def}}{=} T_1(\phi).$$

Note that  $\dagger$  uses a special case of the definition of the Fourier transform, viz.

$$\widehat{\phi}(\omega) = \int_{\mathbb{R}^n} e^{-i\omega \cdot x} \phi(x) dx,$$

for  $\omega = 0 \in \mathbb{R}^n$ .



## (15) 5. FOURIER ANALYSIS & PDE THEORY

Consider the initial value problem

$$\begin{cases} \frac{\partial u}{\partial t} = c \frac{\partial u}{\partial x} \\ u(x, 0) = f(x), \end{cases}$$

for a function  $u : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ , for a given  $f \in C^1(\mathbb{R})$  and constant parameter  $c \in \mathbb{R}$ .

Find the solution for  $u$  in terms of  $f$  and  $c$  by Fourier transformation

$$u(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{i\omega x} \widehat{u}(\omega, t) d\omega.$$

You may assume that  $f$  has a well-defined Fourier transform  $\widehat{f}$ .

Substituting the Fourier formula yields

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} e^{i\omega x} \left( \frac{d\widehat{u}}{dt}(\omega, t) - i\omega c \widehat{u}(\omega, t) \right) d\omega = 0,$$

with

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} e^{i\omega x} (\widehat{u}(\omega, 0) - \widehat{f}(\omega)) d\omega = 0.$$

This holds iff

$$\begin{cases} \frac{d\widehat{u}}{dt}(\omega, t) - i\omega c \widehat{u}(\omega, t) & = & 0, \\ \widehat{u}(\omega, 0) & = & \widehat{f}(\omega). \end{cases}$$

The solution is  $\widehat{u}(\omega, t) = \widehat{f}(\omega) e^{i\omega ct}$ . Fourier inversion yields

$$u(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{i\omega(x+ct)} \widehat{f}(\omega) d\omega = f(x + ct).$$

This reveals that the initial condition is a snapshot of a travelling wave with profile  $f$  propagating with velocity  $c$  (in negative  $x$ -direction if  $c > 0$ , in positive  $x$ -direction if  $c < 0$ , and “frozen” if  $c = 0$ ).

THE END