

Partial Differential Equations I

lecture by

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<https://github.com/andiv/PDE1>

Literature

Elliptic and parabolic partial differential equations:

- JÜRGEN JOST: *Partial Differential Equations*; Springer, 2007
ISBN: 978-0-387-49318-3; doi: 10.1007/978-0-387-49319-0
(good book, but not all details, small errors)
- LAWRENCE C. EVANS: *Partial Differential Equations*; American Mathematical Society, 2010; ISBN: 978-0-8218-4974-3
(part of the lecture follows this book, lots of details)
- DAVID GILBORG, NEIL S. TRUDINGER: *Elliptic Partial Differential Equations of second order*; Springer, 2001; ISBN: 3-540-41160-7
(classic textbook, complete treatment)

Hyperbolic partial differential equations (for the lecture “Partial Differential Equations II”):

- FRITZ JOHN: *Partial Differential Equations*; Springer, 1999
ISBN: 0-387-90609-6
- MICHAEL E. TAYLOR: *Partial Differential Equations I - III*; Springer, 1997
ISBN: 0-387-94653-5, 0-387-94651-9, 0-387-94652-7
(nice detailed text books)
- JOEL SMOLLER: *Shock waves and reaction-diffusion equations*; Springer, 1994
ISBN: 3-540-94259-9
(nicely presented, good motivations, covers most of the material)
- FRIEDRICH SAUVIGNY: *Partial Differential Equations I-II*; Springer, 2012
ISBN: 978-1-4471-2981-3, 978-1-4471-2984-4;
doi: 10.1007/978-1-4471-2981-3, 10.1007/3-540-27540-1
- and many more ...

Contents

1	A Brief Introduction	1
1.1	Definition (Partial Differential Equation)	1
1.2	Examples	1
1.3	Classification	4
1.4	Examples	4
2	Distributions and Fourier Transform	6
2.1	The Schwartz Space and Distributions	6
2.1.1	Definition (Multi-Index)	6
2.1.2	Definition (Schwartz Space)	7
2.1.3	Theorem (Criterion for Convergence)	7
2.1.4	Definition (Tempered Distribution)	8
2.1.5	Lemma (Criterion for Continuity)	8
2.1.6	Example (δ -Distribution)	9
2.1.7	Example (Integral Operator)	9
2.1.8	Remark (Schwartz Functions as Distributions)	11
2.1.9	Definition (regular/singular distribution)	11
2.1.10	Lemma (Multiplication of a Distribution by a Schwartz Function)	11
2.1.11	Example (Derivative of the δ -Distribution)	13
2.1.12	Definition (Distributional Derivative, Convolution)	13
2.1.13	Lemma (Commutativity and Associativity of the Convolution)	13
2.1.14	Proposition (Convolution is Continuous)	14
2.1.15	Definition (Convolution with Distribution)	15
2.2	The Fourier Transform	16
2.2.1	Definition (Fourier Transform)	16
2.2.2	Proposition (Fourier Transform)	16
2.2.3	Theorem (Plancherel, Convergence Generating Factor)	17
2.2.4	Definition (Fourier Transform of Distributions)	20
2.2.5	Theorem (Plancherel for Distributions)	20
2.2.6	Examples	20
2.2.7	Theorem (Fourier Transform is isometry)	22
2.3	Applications to Partial Differential Equations with Constant Coefficients	23
3	The Laplace Equation in $\Omega \subseteq \mathbb{R}^n$	24
3.0	Reminder	24
3.0.1	Theorem (Gauss's Theorem)	24
3.0.2	Theorem (Green's Identities)	24
3.1	Representation Formulas for Harmonic Functions	25
3.1.1	Definition (Harmonic Functions)	25
3.1.2	Definition (Fundamental Solution of Laplace Equation)	26

3.1.3	Theorem (Green's representation)	27
3.1.4	Corollary (Laplacian of Fundamental Solution is Delta Distribution)	28
3.2	The Green's Function	29
3.2.1	Definition (Green's function)	29
3.2.2	Proposition (Solution of Dirichlet Problem)	29
3.2.3	Theorem (Symmetry of the Green's Function)	30
3.2.4	Example (Green's function for $B_R(0)$)	32
3.2.5	Theorem (Poisson representation)	34
3.3	The Mean Value Theorem and the Maximum Principle for Harmonic Functions	35
3.3.1	Theorem (Mean Value Formulas)	35
3.3.2	Theorem (Strong Maximum Principle)	37
3.3.3	Theorem (Weak Maximum Principle)	38
3.3.4	Corollary (Uniqueness of Solutions of the Poisson Equation)	39
3.3.5	Corollary (Liouville's Theorem)	39
3.4	The Harnack Inequality for Harmonic Functions	40
3.4.1	Theorem (Harnack inequality)	40
3.4.2	Corollary (Harnack's Convergence Theorem)	41
4	The Perron Method of Sub- and Supersolution	43
Appendix		45
	Acknowledgements	45

1 A Brief Introduction

An ordinary differential equation (ODE) can be written as:

$$\frac{d}{dt}u(t) = \dot{u}(t) = v(t, u)$$

$$u : I \subseteq \mathbb{R} \rightarrow \mathbb{R}^N$$

This equation involves only derivatives with respect to *one* variable t .

$$\frac{\partial}{\partial x}f(x, y) + \frac{\partial}{\partial y}f(x, y) = 0$$

This is an example for a partial differential equation.

1.1 Definition (Partial Differential Equation)

A *partial differential equation* (PDE) is a (scalar) equation, which involves partial derivatives of an unknown function $u : \Omega \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$. We always assume that $\Omega \subseteq \mathbb{R}^n$ is open.

More generally, a *system of partial differential equations* is a system of equations involving partial derivatives of a function $u : \Omega \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^N$.

Similarly one can define partial differential equations on manifolds.

For ordinary differential equations we considered the initial-value problem:

$$\dot{u}(t) = v(t, u) \qquad u(t_0) = u_0$$

For partial differential equations one considers

- the initial-value problem and
- the boundary-value problem.

1.2 Examples

1. Cauchy-Riemann equations: Let

$$f : \Omega \overset{\text{open}}{\subseteq} \mathbb{C} \rightarrow \mathbb{C}$$

be holomorphic.

$$f = a + \mathbf{i}b \qquad a := \operatorname{Re}(f) \qquad b := \operatorname{Im}(f)$$

$$\frac{\partial a}{\partial x} = \frac{\partial b}{\partial y} \qquad \frac{\partial b}{\partial x} = -\frac{\partial a}{\partial y}$$

This is a system of two partial differential equations.

$$u := \begin{pmatrix} a \\ b \end{pmatrix} \quad u : \Omega \subseteq \mathbb{C} = \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$\frac{\partial^2 a}{\partial x^2} + \frac{\partial^2 a}{\partial y^2} = \frac{\partial b}{\partial x \partial y} - \frac{\partial b}{\partial y \partial x} = 0$$

$$\Rightarrow \underbrace{\left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} \right)}_{=: \Delta} a = 0 \quad \Delta b = 0$$

This is the *Laplace equation* with the *Laplace operator* (or *Laplacian*) Δ . Solutions of the Laplace equation are called *harmonic functions*.

2. Let (M, g) be a Riemannian manifold. Here exists the Laplace-Beltrami operator Δ .

– In the special case $M = \mathbb{R}^n$ we have:

$$\Delta = \frac{\partial^2}{\partial x_1^2} + \dots + \frac{\partial^2}{\partial x_n^2}$$

$$\Delta \varphi = 0$$

– With the Riemannian metric g_{ij} we can define:

$$\Delta \varphi = g^{ij} \nabla_i \nabla_j \varphi = \operatorname{div}(\operatorname{grad}(\varphi)) = \frac{1}{\sqrt{\det(g)}} \partial_j \left(\sqrt{\det(g)} g^{jk} \partial_k \varphi \right)$$

This gives an elliptic equation.

3. Newton's gravitational law: Let $\varrho(x)$ be the mass density and $\varphi(x)$ the Newtonian potential.

$$\Delta \varphi = \underbrace{-4\pi \varrho}_{\text{inhomogeneity}}$$

Such an inhomogeneous Laplace equation is usually referred to as Poisson equation and it is elliptic.

4. Heat flow equation (Wärmeleitungsgleichung): Let $\varphi(t, x)$ be the temperature at time $t \in \mathbb{R}$ and position $x \in \mathbb{R}^n$.

$$\partial_t \varphi(t, x) = \Delta \varphi(t, x)$$

This is a parabolic equation.

5. The Schrödinger equation is a parabolic equation:

$$\mathbf{i} \hbar \partial_t \psi(t, x) = \left(-\frac{\hbar^2}{2m} \Delta + V \right) \psi(t, x)$$

Additionally to the the heat flow equation there is the potential V , but more important there is a factor of \mathbf{i} in front of the partial derivative. The time-independent Schrödinger equation is:

$$E \psi(x) = \left(-\frac{\hbar^2}{2m} \Delta + V \right) \psi(x)$$

This is similar to the Poisson equation and also elliptic.

6. The wave equation

$$(\partial_t^2 - \Delta_x) \psi(t, x) = 0$$

is hyperbolic. We will consider it in the lecture “Partial Differential Equations II”.

7. Maxwell’s equations: $E(t, x)$ is the electric field and $B(t, x)$ the magnetic field.

$$\begin{array}{ll} \operatorname{div}(E) = 4\pi \varrho & \text{Gauss law} \\ \operatorname{rot}(E) = -\partial_t B & \text{Maxwell} \\ \operatorname{div}(B) = 0 & \\ \operatorname{rot}(B) = 4\pi j - \partial_t E & \text{Faraday} \end{array}$$

This is a system of 8 partial differential equation.

8. Einstein’s field equation:

$$R_{ij} - \frac{1}{2} R g_{ij} = 4\pi \kappa T_{ij}$$

This is a geometric partial differential equation. R_{ij} is the Ricci curvature, R the scalar curvature and T_{ij} the energy-momentum tensor. It is a system of 10 partial differential equations.

9. Equations of relativistic quantum mechanics:

$$(-\partial_t^2 + \Delta) \psi = m^2 \psi$$

This is the Klein-Gordon equation with the mass m .

$$\mathbf{i} \gamma^j \partial_j \psi = m \psi$$

This is the Dirac equation, a system of 4 complex-valued or 8 real-valued partial differential equations for a particle with spin $\frac{1}{2}$.

10. Water waves can be described by the Korteweg-de Vries equation:

$$\partial_t u + u \partial_x u + \partial_x^3 u = 0$$

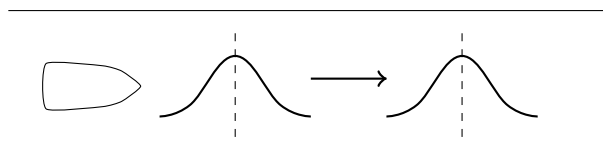


Figure 1.1: Solitons (discovered by John Russel in 1834): When the ship suddenly stops, the water flows on along the channel. This wave moves with a constant speed and its shape stays the same.

11. Shock waves: Burger’s equation

$$\partial_t u + u \partial_x u = 0$$

is hyperbolic.

12. Turbulence can be described by the incompressible Navier-Stokes equations for the velocity $v : \Omega \subseteq \mathbb{R}^3 \rightarrow \mathbb{R}^3$.

$$\begin{array}{l} \operatorname{div}(v) = 0 \\ \rho \partial_t v^j + \rho v^i \partial_i v^j - \eta \Delta v^j = -\partial_j P \end{array}$$

Here ϱ is the gas density, P the pressure and η the viscosity.

1.3 Classification

I) The *order* of a partial differential equation is the highest order of the derivatives in it.

$$\begin{array}{ll} \Delta u = f & \text{second order} \\ \partial_t \varphi = \Delta \varphi & \text{second order} \\ \partial_t u + u \partial_x u + \partial_x^3 u = 0 & \text{third order} \end{array}$$

II) Algebraic classification:

a) *Linear* equations: The unknown function u and its derivatives appear only linearly.

$$\begin{array}{ll} \partial_t u = u & \text{linear} \\ \partial_t u + u \partial_x u = 0 & \text{non-linear} \end{array}$$

b) Linear *homogeneous* equations: If u is a solution, then λu for $\lambda \in \mathbb{R}$ is also a solution.

$$\begin{array}{ll} \Delta u = 0 & \text{linear homogeneous} \\ \Delta u = \varrho & \text{linear inhomogeneous} \\ \Delta \setminus u = 0 & \text{linear homogeneous} \end{array}$$

c) Linear with *constant coefficients*:

$$\begin{array}{ll} \Delta u = \rho & \text{linear with constant coefficients} \\ \Delta \setminus u = \varrho & \text{in general non-constant coefficients} \end{array}$$

III) Classification by type: elliptic, parabolic, hyperbolic

Here we only consider scalar second order equations with $x \in \Omega \subseteq \mathbb{R}^n$.

$$\begin{array}{l} F(x, u, Du, D^2u) = 0 \\ F : \Omega \times \mathbb{R} \times \mathbb{R}^n \times \mathbb{R}^{n \times n} \rightarrow \mathbb{R} \end{array}$$

$$A_{ij} := \frac{\partial F(x, u, p_i, p_{ij})}{\partial p_{ij}}$$

is a symmetric $n \times n$ matrix.

- If A is positive definite, the equation is called *elliptic*.
 - If A has $n - 1$ positive and one negative eigenvalue, the equation is called *hyperbolic*.
 - If A has $n - 1$ positive eigenvalues and a non-trivial kernel, the equation is called *parabolic*.
 - If all eigenvalues are negative or $n - 1$ are negative, then we replace F by $-F$.
- All other case of *mixed type* are difficult and we do not consider them in this lecture.

1.4 Examples

Consider the Poisson equation:

$$\Delta u = \varrho$$

$$\begin{aligned}
 F(x, u, Du, D^2u) &= -\varrho(x) + \delta^{ij} \partial_{ij} u \\
 F(x, u, p_i, p_{ij}) &= -\varrho(x) + \delta^{ij} p_{ij} \\
 A_{ij} &= \frac{\partial F(x, u, p_i, p_{ij})}{\partial p_{ij}} = \delta_{ij}
 \end{aligned}$$

So we have $A = \mathbb{1}$ and thus the equation is elliptic.

Now consider the inhomogeneous wave equation:

$$(\partial_t^2 - \Delta) \phi(t, x_1, x_2, x_3) = \varrho$$

$$F(x, u, p_i, p_{ij}) = \varrho(x) + \eta^{ij} p_{ij} \quad \eta = \text{diag}(-1, 1, 1, 1)$$

So $A = \eta$ has one negative and three positive eigenvalues which means that the equation is hyperbolic.

$$\partial_t \phi = \Delta \phi$$

$$\begin{aligned}
 F(x, u, D\phi, D^2\phi) &= -\partial_0 \phi + \sum_{i,j=1}^3 \delta^{ij} \partial_{ij} \phi \\
 A_{ij} &= \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}
 \end{aligned}$$

Therefore the equation is parabolic.

2 Distributions and Fourier Transform

Motivation

We want to solve partial differential equations with constant coefficients in $\Omega = \mathbb{R}^n$, for example:

$$(-\partial_t^2 + \Delta) \phi = 0$$

Now we make a “plane wave ansatz” with $t, \omega \in \mathbb{R}$ and $k, x \in \mathbb{R}^{n-1}$:

$$\begin{aligned}\phi(t, x) &= e^{-i\omega t + i\langle k, x \rangle} \\ \partial_t \phi(t, x) &= -i\omega \phi(t, x) \\ \partial_j \phi(t, x) &= ik_j \phi(t, x)\end{aligned}$$

This gives an algebraic equation:

$$\begin{aligned}(-(-i\omega)^2 + (ik)^2) \phi &= 0 \\ \Leftrightarrow \omega^2 &= k^2\end{aligned}$$

We also want to differentiate non-smooth functions, e.g.:

$$\Delta_{\mathbb{R}^3} \frac{1}{|x|} = -4\pi \delta(x)$$

$\delta(x)$ is called Dirac δ -distribution.

2.1 The Schwartz Space and Distributions

Laurent Schwartz was the first to investigate distributions systematically. He was awarded the fields medal for his research.

2.1.1 Definition (Multi-Index)

For \mathbb{R}^n we denote indices by $i, j, k \in \{1, \dots, n\}$. We call $\alpha = (i_1, \dots, i_k)$ with $i_l \in \{1, \dots, n\}$ a *multi-index*. $|\alpha| := k$ is called the *order* or *absolute value* of the multi-index.

With this we can write differentials of order k as

$$D^\alpha := \frac{\partial}{\partial x^{i_1}} \cdots \frac{\partial}{\partial x^{i_k}} \quad (2.1)$$

and homogeneous polynomials of degree k in the components of a vector $x = (x^1, \dots, x^n)$ as:

$$x^\alpha := x^{i_1} \cdots x^{i_k} \quad (2.2)$$

For $f \in C^\infty(\mathbb{R}^n)$ and $r, s \in \mathbb{N}$ we define the *Schwartz norm*:

$$\|f\|_{r,s} := \sum_{\alpha, |\alpha| \leq r} \sum_{\beta, |\beta| \leq s} \sup_{x \in \mathbb{R}^n} |x^\alpha D^\beta f(x)| \quad (2.3)$$

For example for $r = 0 = s$ we have:

$$\|f\|_{0,0} = \sup_{x \in \mathbb{R}^n} |f(x)| = \|f\|_{C^0}$$

2.1.2 Definition (Schwartz Space)

The *Schwartz space* $\mathcal{S}(\mathbb{R}^n)$ is the vector space of all $f \in C^\infty(\mathbb{R}^n)$ for which all Schwartz norms are finite, i.e. for all $r, s \in \mathbb{N}$ holds:

$$\|f\|_{r,s} < \infty$$

This space is an infinite-dimensional vector space.

On a normed space $(E, \|\cdot\|)$, the topology is given by the open sets.

$\Omega \subseteq E$ is defined as *open* if holds:

$$\forall_{x \in \Omega} \exists_{\varepsilon > 0} : B_\varepsilon(x) \subseteq \Omega$$

A subset $\Omega \subseteq \mathcal{S}(\mathbb{R}^n)$ is called *open* if for every $f \in \Omega$ there is a $\varepsilon > 0$ and $r, s \in \mathbb{N}$ such that holds:

$$\left\{ g \in \mathcal{S} \mid \|g - f\|_{r,s} < \varepsilon \right\} \subseteq \Omega \quad (2.4)$$

Note: This topology is fine, because it involves many open sets, since the condition for open only involves the statement “there are $r, s \in \mathbb{N}$ ”.

Convergence $f_n \rightarrow f$ in \mathcal{S} means that *every* open neighborhood U of f contains almost all f_n . For a finer topology, the condition for a sequence to converge is stronger.

2.1.3 Theorem (Criterion for Convergence)

Convergence $f_n \rightarrow f$ in \mathcal{S} is equivalent to the convergence $\|f_n - f\|_{r,s} \rightarrow 0$ for all $r, s \in \mathbb{N}$.

Proof

“ \Rightarrow ”: Suppose that $f_n \rightarrow f$ converges. By definition of the convergence, every open neighborhood of f contains almost all f_n . For all $r, s \in \mathbb{N}$ the sets $U_\varepsilon^{r,s} := \{g \mid \|g - f\|_{r,s} < \varepsilon\}$ are open by definition. So the inequality

$$\|f_n - f\|_{r,s} < \varepsilon$$

holds for almost all f_n and thus converges $\|f_n - f\|_{r,s} \rightarrow 0$.

“ \Leftarrow ”: Assume that $\|f_n - f\|_{r,s} \rightarrow 0$ converges for all $r, s \in \mathbb{N}$. Let A be an open neighborhood of f . This means by definition that there exist $r, s \in \mathbb{N}$ and $\varepsilon \in \mathbb{R}_{>0}$ with $U_\varepsilon^{r,s} \subseteq A$. For this (r, s) we know that $\|f_n - f\|_{r,s} \rightarrow 0$ converges. Hence there exists a $N \in \mathbb{N}$ such that $\|f_n - f\|_{r,s} < \varepsilon$ holds for all $n \in \mathbb{N}_{>N}$, in other words $f_n \in U_\varepsilon^{r,s} \subseteq A$. So $f_n \rightarrow f$ converges in \mathcal{S} . $\square_{2.1.3}$

A vector space with a topology generated by a family of norms or semi-norms is a *uniform space* and is called *topological vector space*.

2.1.4 Definition (Tempered Distribution)

Let $\mathcal{S}^*(\mathbb{R}^n)$ be the dual space of $\mathcal{S}(\mathbb{R}^n)$. It is called the space of *tempered distributions* (temperierte Distributionen).

In linear algebra for a finite-dimensional vector space V , the dual space $V^* = L(V, \mathbb{R})$ is the space of linear functionals. V^* is again a vector space with $\dim(V^*) = \dim(V)$.

Here $\mathcal{S}(\mathbb{R}^n)$ is an infinite-dimensional vector space with a topology. $\mathcal{S}^*(\mathbb{R}^n)$ is the space of all *continuous* linear functionals.

In a Banach space $(E, \|\cdot\|)$ holds: A linear functional $A : E \rightarrow \mathbb{R}$ is continuous if and only if A is bounded, i.e. $|Au| \leq c \|U\|$ for all $u \in E$.

2.1.5 Lemma (Criterion for Continuity)

A linear functional $T : \mathcal{S}(\mathbb{R}^n) \rightarrow \mathbb{R}$ is continuous if and only if there are $r, s \in \mathbb{N}$ and a $c \in \mathbb{R}_{>0}$ such that holds:

$$|Tf| \leq c \|f\|_{r,s} \quad (2.5)$$

Proof

“ \Leftarrow ”: Assume that (2.5) holds for some $r, s \in \mathbb{N}$. We want to show that T is continuous. To this end, let $f_n \rightarrow f$ be a convergent series in \mathcal{S} . Our task is to show that $Tf_n \rightarrow Tf$ converges.

The convergence $f_n \rightarrow f$ implies $\|f_n - f\|_{r',s'} \rightarrow 0$ for all $r', s' \in \mathbb{N}$ and thus in particular for r, s satisfying the inequality (2.5). By linearity follows:

$$|Tf_n - Tf| = |T(f_n - f)| \leq c \|f_n - f\|_{r,s} \xrightarrow{n \rightarrow \infty} 0$$

So T maps convergent sequences to convergent sequences and is thus continuous.

“ \Rightarrow ”: Assume that T is continuous. Then the preimage of open sets is open, in particular $T^{-1}(B_1(0)) \subseteq \mathcal{S}(\mathbb{R}^n)$ is open. So there exist $r, s \in \mathbb{N}$ and a $\varepsilon \in \mathbb{R}_{>0}$ such that holds:

$$T^{-1}(B_1(0)) \supseteq U_{\varepsilon}^{r,s} := \{g \mid \|g\|_{r,s} < \varepsilon\}$$

This implies:

$$\|g\|_{r,s} < \varepsilon \quad \Rightarrow \quad g \in T^{-1}(B_1(0))$$

Now $g \in T^{-1}(B_1(0))$ means $|Tg| < 1$. For any $f \in \mathcal{S}(\mathbb{R}^n)$ apply this to $g = \frac{f}{\lambda}$ with $\lambda \in \mathbb{R}_{>0}$.

$$\begin{aligned} \frac{1}{\lambda} \|f\|_{r,s} < \varepsilon &\Rightarrow \frac{1}{\lambda} |Tf| < 1 & / \cdot \lambda \\ \|f\|_{r,s} < \lambda \varepsilon &\Rightarrow |Tf| < \lambda \end{aligned}$$

Now choose $\lambda = \frac{2}{\varepsilon} \|f\|_{r,s}$, so the left side holds, which implies:

$$|Tf| < \frac{2}{\varepsilon} \|f\|_{r,s} \quad \forall_{f \in \mathcal{S}(\mathbb{R}^n)}$$

□_{2.1.5}

2.1.6 Example (δ -Distribution)

a) Consider the following functional:

$$\begin{aligned}\delta : \mathcal{S}(\mathbb{R}) &\rightarrow \mathbb{R} \\ f &\mapsto f(0)\end{aligned}\tag{2.6}$$

This is obviously linear.

$$|\delta(f)| = |f(0)| \leq \sup_{\mathbb{R}} |f| = \|f\|_{0,0}$$

Hence δ is continuous, which means that $\delta \in \mathcal{S}^*(\mathbb{R})$ is a tempered distribution. A convenient *notation* with $f \in \mathcal{S}(\mathbb{R})$ is:

$$\delta(f) = \int_{\mathbb{R}} f(x) \delta(x) dx$$

b) In higher dimension $n \in \mathbb{N}$ we define:

$$\begin{aligned}\delta : \mathcal{S}(\mathbb{R}^n) &\rightarrow \mathbb{R} \\ f &\mapsto f(0)\end{aligned}$$

Again holds $|\delta(f)| \leq \|f\|_{0,0}$. The physicists' notation for this is:

$$\begin{aligned}\delta(f) &= \int_{\mathbb{R}^n} f(x) \delta^{(n)}(x) dx \\ \delta^{(n)}(x) &= \delta(x^1) \cdots \delta(x^n)\end{aligned}$$

Remark

δ can also be introduced as a *measure* on \mathbb{R}^n , the *Dirac measure*. For $A \subseteq \mathbb{R}^n$ define:

$$\delta(x) = \begin{cases} 1 & \text{if } 0 \in A \\ 0 & \text{otherwise} \end{cases}\tag{2.7}$$

Then for $f \in C^0(\mathbb{R}^n)$ the expression

$$\int_{\mathbb{R}^n} f(x) d\delta(x) = f(0)$$

makes mathematical sense as an integral.

This is useful because convergence theorems and so on from measure theory are available. The problem is, that this does not work for every distribution and thus is not general enough for most purposes, e.g. the derivative $\delta'(x)$ is a distribution, but cannot be written as a measure.

2.1.7 Example (Integral Operator)

a) Consider $g \in C^\infty(\mathbb{R}^n)$ with at most polynomial growth, i.e. there are $c \in \mathbb{R}_{>0}$ and $r \in \mathbb{N}$ such that holds:

$$|g(x)| \leq c(1 + |x|^r)$$

Now define:

$$T_g : \mathcal{S}(\mathbb{R}^n) \rightarrow \mathbb{R}$$

$$f \mapsto \int_{\mathbb{R}^n} g(x) f(x) d^n x$$

The integral here is just the Lebesgue integral and it exists:

For $f \in \mathcal{S}(\mathbb{R}^n)$ holds $\|f\|_{r,s} < \infty$ for all $r, s \in \mathbb{N}$.

$$\sup_{\mathbb{R}} |f| + \sup_{x \in \mathbb{R}} (|x|^{\tilde{r}} \cdot |f(x)|) \leq \|f\|_{\tilde{r},0} < \infty$$

$$\Rightarrow \sup_{x \in \mathbb{R}} \left((1 + |x|^{\tilde{r}}) |f(x)| \right) \leq \|f\|_{\tilde{r},0}$$

$$\Rightarrow |f(x)| \leq \frac{\|f\|_{\tilde{r},0}}{1 + |x|^{\tilde{r}}} \quad \forall \tilde{r} \in \mathbb{N}$$

So we get:

$$T_g f = \int g(x) f(x) d^n x \leq \int c(1 + |x|^r) \frac{\|f\|_{\tilde{r},0}}{(1 + |x|^{\tilde{r}})} d^n x =$$

$$\stackrel{\substack{\text{polar coordinates} \\ \rho := |x|}}{=} c \|f\|_{\tilde{r},0} \underbrace{\mu(S^{n-1})}_{\text{volume of unit sphere}} \int_0^\infty \rho^{n-1} \frac{1 + \rho^r}{1 + \rho^{\tilde{r}}} d\rho \stackrel{\tilde{r} > r+n}{<} \infty$$

This is finite if and only if the integrand decays faster than ρ^{-1} , i.e. $n - 1 + r - \tilde{r} < -1$ and thus $\tilde{r} > n + r$. Since $\tilde{r} \in \mathbb{N}$ is arbitrary, the integral exists.

Continuity: The previous estimate implies with $\tilde{r} = n + r + 1$:

$$|T_g f| \leq C(g, n) \|f\|_{\tilde{r},0}$$

Thus $T_g \in \mathcal{S}^*(\mathbb{R}^n)$ is a tempered distribution.

b) Chose $g(x) = e^x$ and define:

$$T_g f := \int_{-\infty}^{\infty} f(x) e^x dx$$

This is *not* a well-defined tempered distribution. Namely, choose:

$$f(x) = \frac{1}{\cosh\left(\frac{x}{2}\right)}$$

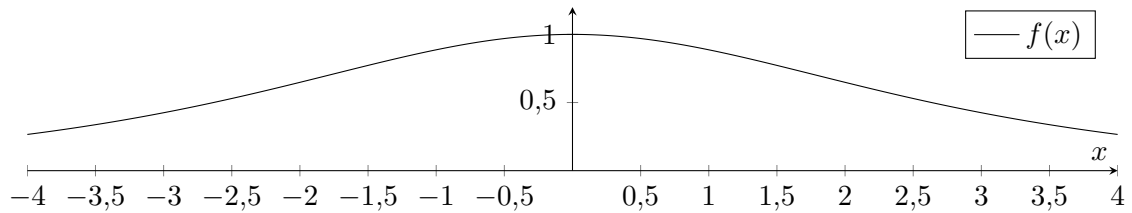


Figure 2.1: $f(x)$ decays rapidly.

$f(x)$ and all its derivatives decay rapidly (exponentially fast $\sim e^{-\frac{x}{2}}$) at $\pm\infty$, so $f \in \mathcal{S}$ is a Schwartz function. But $T_g f$ diverges:

$$T_g f = \int_{-\infty}^{\infty} \frac{e^x}{\cosh\left(\frac{x}{2}\right)} dx = +\infty$$

2.1.8 Remark (Schwartz Functions as Distributions)

The mapping

$$\begin{aligned} T : \mathcal{S}(\mathbb{R}^n) &\rightarrow \mathcal{S}^*(\mathbb{R}^n) \\ f &\mapsto T_f \end{aligned}$$

is injective.

Proof

If T was injective, there were $f_1, f_2 \in \mathcal{S}(\mathbb{R}^n)$ with $T_{f_1} = T_{f_2}$ and $f_1 \neq f_2$. By linearity this would imply $T_g = 0$ with $g = f_1 - f_2 \neq 0$ and we could choose a $y \in \mathbb{R}^n$ with $g(y) \neq 0$. By continuity follows $g > 0$ or $g < 0$ in a neighborhood U of y . Now choose a test function h with $\text{supp}(h) \subseteq U$ and $h \geq 0$. Then follows the contradiction:

$$0 = T_g(h) = \int_{\mathbb{R}^n} g(x) h(x) \, d^n x \neq 0$$

□_{2.1.8}

Thus we can regard distributions as “generalized functions”. Namely we identify a function g with T_g . (Later on many people often do not distinguish between g and T_g .)

Operations on Schwartz Functions and Distributions

- $\mathcal{S}^*(\mathbb{R}^n)$ is a vector space with addition $T+S$ and scalar multiplication $\alpha \cdot f$ for distributions $T, S \in \mathcal{S}^*(\mathbb{R}^n)$ and $\alpha \in \mathbb{R}$.
- Multiplication of a distribution by a Schwartz function is defined for $T \in \mathcal{S}^*(\mathbb{R}^n)$ and $f, g \in \mathcal{S}(\mathbb{R}^n)$ as:

$$(fT)(g) := T(f \cdot g) \tag{2.8}$$

This is well defined, because $f \cdot g$ is again a Schwartz function and for $h \in \mathcal{S}(\mathbb{R}^n)$ holds:

$$\begin{aligned} (fT_h)(g) &\stackrel{\text{definition of } fT_h}{=} T_h(f \cdot g) \stackrel{\text{definition of } T_h}{=} \int_{\mathbb{R}^n} h(x) (f \cdot g)(x) \, d^n x = \\ &= \int_{\mathbb{R}^n} (f \cdot h)(x) g(x) \, d^n x = T_{fh}(g) \end{aligned}$$

So this definition extends the multiplication of Schwartz functions to distributions. But we still have to show, that this operation gives a continuous functional.

2.1.9 Definition (regular/singular distribution)

A tempered distribution T is called *regular* distribution if there is a $g \in L^1_{\text{loc}}(\mathbb{R}^n)$ (locally integrable, i.e. integrable on every compact interval) with $T = T_g$. Otherwise, T is called *singular*.

For example $\delta(x)$ is singular.

2.1.10 Lemma (Multiplication of a Distribution by a Schwartz Function)

Let $f \in \mathcal{S}(\mathbb{R}^n)$ be a Schwartz function and $T \in \mathcal{S}^*(\mathbb{R}^n)$ a distribution. Then fT is a *continuous* linear functional on $\mathcal{S}(\mathbb{R}^n)$, in other words $fT \in \mathcal{S}^*(\mathbb{R}^n)$ is also a distribution.

Proof

According to Lemma 2.1.5, our task is to show that there are $r, s \in \mathbb{N}$ and a $C \in \mathbb{R}_{>0}$ with:

$$|(fT)(g)| \leq C \|g\|_{r,s}$$

Since T is continuous, there exist $r, s \in \mathbb{N}$ and a $\tilde{C} \in \mathbb{R}_{>0}$ with:

$$|T(fg)| \leq \tilde{C} \|fg\|_{r,s}$$

Thus it remains to show that there is a $c(r, s) \in \mathbb{R}_{>0}$ such that for all $f, g \in \mathcal{S}(\mathbb{R}^n)$ holds:

$$\|fg\|_{r,s} \leq c(r, s) \|f\|_{r,s} \cdot \|g\|_{r,s}$$

This inequality can be proven by induction in s .

Induction basis $s = 0$:

$$\begin{aligned} \|fg\|_{r,0} &= \sum_{|\alpha| \leq r} \sup_{x \in \mathbb{R}^n} (|x^\alpha f(x) g(x)|) \leq \\ &\leq \sup_{y \in \mathbb{R}^n} |g(y)| \sum_{|\alpha| \leq r} \sup_{x \in \mathbb{R}^n} (|x^\alpha f(x)|) = \\ &= \|g\|_{0,0} \cdot \|f\|_{r,0} \leq \|g\|_{r,0} \cdot \|f\|_{r,0} \end{aligned}$$

Induction step $s \rightsquigarrow s+1$: Assume that the statement holds for a $s \in \mathbb{N}$ for all $r \in \mathbb{N}$. Let β be a multi-index with $|\beta| = s+1$, i.e. $\beta = (i_1, \dots, i_{s+1})$. Now set:

$$\hat{\beta} := (i_1, \dots, i_s) \qquad j := i_{s+1} \qquad D^\beta = D^{\hat{\beta}} \frac{\partial}{\partial x^j}$$

It holds:

$$D^\beta(fg) = D^{\hat{\beta}} \frac{\partial}{\partial x^j} (fg) = D^{\hat{\beta}} \left(\left(\frac{\partial}{\partial x^j} f \right) g + f \left(\frac{\partial}{\partial x^j} g \right) \right)$$

$$\begin{aligned} \|fg\|_{r,s+1} &= \|f \cdot g\|_{r,s} + \sum_{\substack{|\alpha| \leq r \\ |\beta| = s+1}} \sup_{x \in \mathbb{R}} \left| x^\alpha D^\beta (f \cdot g)(x) \right| = \\ &= \|fg\|_{r,s} + \sum_{\substack{|\alpha| \leq r \\ |\hat{\beta}| = s}} \sum_{j=1}^n \sup_{x \in \mathbb{R}} \left| x^\alpha D^{\hat{\beta}} \left(\left(\frac{\partial}{\partial x^j} f \right) g + f \left(\frac{\partial}{\partial x^j} g \right) \right) \right| \leq \\ &\stackrel{\substack{\text{induction} \\ \leq \\ \text{hypothesis}}}{\leq} c(r, s) \|f\|_{r,s} \|g\|_{r,s} + \sum_{j=1}^n c(r, s) \left(\left\| \frac{\partial}{\partial x^j} f \right\|_{r,s} \|g\|_{r,s} + \|f\|_{r,s} \left\| \frac{\partial}{\partial x^j} g \right\|_{r,s} \right) \leq \\ &\leq c(r, s) \|f\|_{r,s} \|g\|_{r,s} + n \cdot c(r, s) \left(\|f\|_{r,s+1} \|g\|_{r,s} + \|f\|_{r,s} \|g\|_{r,s+1} \right) \leq \\ &\leq \underbrace{(2n+1) c(r, s)}_{=c(r,s+1)} \|f\|_{r,s+1} \|g\|_{r,s+1} \end{aligned}$$

□2.1.10

2.1.11 Example (Derivative of the δ -Distribution)

We make a formal computation:

$$\int_{\mathbb{R}} \delta'(x) f(x) dx = \int_{\mathbb{R}} \left(\frac{d}{dx} \delta(x) \right) f(x) dx \stackrel{\text{integration by parts}}{=} - \int_{\mathbb{R}} \delta(x) f'(x) dx = -f'(0)$$

This motivates us to *define*:

$$\begin{aligned} \delta' : \mathcal{S}(\mathbb{R}) &\rightarrow \mathbb{R} \\ f &\mapsto -f'(0) \end{aligned} \quad (2.9)$$

This is obviously linear and it is continuous, because for all $f \in \mathcal{S}(\mathbb{R})$ holds:

$$|\delta'(f)| = |f'(0)| \leq \|f\|_{0,1}$$

Hence we have $\delta' \in \mathcal{S}^*(\mathbb{R})$.

Remark

In contrast to δ , the distribution δ' cannot be introduced as a measure:

$$\begin{aligned} \delta(\Omega) &= \begin{cases} 1 & \text{if } 0 \in \Omega \\ 0 & \text{otherwise} \end{cases} \\ \delta'(\Omega) &=? \end{aligned}$$

2.1.12 Definition (Distributional Derivative, Convolution)

- For a tempered distribution $T \in \mathcal{S}^*(\mathbb{R}^n)$ we define the *distributional derivative* $D^\alpha T$ by:

$$(D^\alpha T)(f) := (-1)^{|\alpha|} T(D^\alpha f) \quad (2.10)$$

$D^\alpha T$ is a distribution, since it is a continuous functional:

$$|(D^\alpha T)(f)| = |T(D^\alpha f)| \stackrel{T \text{ continuous}}{\leq} C \|D^\alpha f\|_{r,s} \leq C \|f\|_{r,s+|\alpha|}$$

So we have a mapping $D^\alpha : \mathcal{S}^*(\mathbb{R}^n) \rightarrow \mathcal{S}^*(\mathbb{R}^n)$.

- The convolution (Faltung) for $f, g \in \mathcal{S}(\mathbb{R}^n)$ is defined as:

$$(f * g)(x) := \int_{\mathbb{R}^n} f(x-y) g(y) d^n y \quad (2.11)$$

2.1.13 Lemma (Commutativity and Associativity of the Convolution)

The convolution is commutative and associative, i.e. for $f, g, h \in \mathcal{S}(\mathbb{R}^n)$ holds:

$$\begin{aligned} f * g &= g * f & f * (g * h) &= (f * g) * h \end{aligned} \quad (2.12)$$

Proof

$$(f * g)(x) = \int_{\mathbb{R}^n} f(x-y) g(y) d^n y \stackrel{z:=x-y}{\underset{d^n z = d^n y}{=}} \int_{\mathbb{R}^n} f(z) g(x-z) d^n z = (g * f)(x)$$

Associativity follows analogously using Fubini's theorem, which can be applied, since the function $(y, z) \mapsto f(x-y) g(y-z) h(z)$ is an element of $\mathcal{S}(\mathbb{R}^{2n}) \subseteq L^1(\mathbb{R}^{2n})$:

$$\begin{aligned} f * (g * h)(x) &= \int_{\mathbb{R}^n} f(x-y) (g * h)(y) d^n y = \\ &= \int_{\mathbb{R}^n} f(x-y) \left(\int_{\mathbb{R}^n} g(y-z) h(z) d^n z \right) d^n y = \\ &\stackrel{\tilde{y}:=y-z}{\underset{d^n \tilde{y} = d^n y}{=}} \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} f(x-\tilde{y}-z) g(\tilde{y}) h(z) d^n \tilde{y} d^n z = \\ &= \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^n} f(x-z-\tilde{y}) g(\tilde{y}) d^n \tilde{y} \right) h(z) d^n z = \\ &= \int_{\mathbb{R}^n} (f * g)(x-z) h(z) d^n z = \\ &= (f * g) * h(x) \end{aligned}$$

□_{2.1.13}

2.1.14 Proposition (Convolution is Continuous)

The convolution $*$: $\mathcal{S} \times \mathcal{S} \rightarrow \mathcal{S}$ is continuous.

Proof

$$\|f * g\|_{r,s} = \sum_{\substack{|\alpha| \leq r \\ |\beta| \leq s}} \sup_{x \in \mathbb{R}^n} \left| x^\alpha D_x^\beta \int_{\mathbb{R}^n} f(x-y) g(y) d^n y \right|$$

The derivative may be commuted with the integral:

$$\begin{aligned} \frac{\partial}{\partial x_j} (f * g)(x) &= \lim_{\varepsilon \searrow 0} \frac{1}{\varepsilon} ((f * g)(x + \varepsilon e_j) - (f * g)(x)) = \\ &= \lim_{\varepsilon \searrow 0} \int_{\mathbb{R}^n} \frac{f(x + \varepsilon e_j - y) - f(x - y)}{\varepsilon} g(y) d^n y \end{aligned}$$

Using the estimate

$$\begin{aligned} \frac{f(x + \varepsilon e_j - y) - f(x - y)}{\varepsilon} &= \frac{1}{\varepsilon} \int_0^1 \frac{d}{d\tau} (f(x + \varepsilon \tau e_j - y)) d\tau = \\ &= \frac{1}{\varepsilon} \int_0^1 (\partial_j f)(x + \varepsilon \tau e_j - y) \varepsilon d\tau \\ \Rightarrow \left| \frac{f(x + \varepsilon e_j - y) - f(x - y)}{\varepsilon} \right| &\leq \sup_{\mathbb{R}^n} |\partial_j f| \leq \|f\|_{0,1} \end{aligned}$$

we get:

$$\left| \frac{f(x + \varepsilon e_j - y) - f(x - y)}{\varepsilon} g(y) \right| \leq \|f\|_{0,1} \cdot |g(y)| \in L^1(\mathbb{R}^n)$$

Thus the dominated convergence theorem implies:

$$\frac{\partial}{\partial x_j} (f * g)(x) = \int_{\mathbb{R}^n} \frac{\partial}{\partial x_j} f(x - y) g(y) d^n y$$

By induction follows:

$$x^\alpha D^\beta \int_{\mathbb{R}^n} f(x - y) g(y) d^n y = x^\alpha \int_{\mathbb{R}^n} (D^\beta f)(x - y) g(y) d^n y$$

Now we treat the x^α :

$$x^\alpha = ((x - y) + y)^\alpha = \sum_{\gamma, \delta \text{ with } \alpha = \gamma + \delta} c_{\gamma\delta} (x - y)^\gamma \cdot y^\delta$$

Here holds $|\gamma|, |\delta| \leq |\alpha|$. Now we can estimate:

$$\left| (x - y)^\gamma D^\beta f(x) \right| \leq \|f\|_{r,s}$$

Hence we get:

$$\begin{aligned} \|f * g\|_{r,s} &\leq c(r, s) \|f\|_{r,s} \sum_{\delta} \int_{\mathbb{R}^n} |y^\delta g(y)| \cdot \frac{(1 + |y|)^{n+1}}{(1 + |y|)^{n+1}} d^n y \leq \\ &\leq c(r, s) \|f\|_{r,s} \|g\|_{n+1+r,0} \underbrace{\int_{\mathbb{R}^n} \frac{d^n y}{(1 + |y|)^{n+1}}}_{< \infty} \end{aligned}$$

□_{2.1.14}

2.1.15 Definition (Convolution with Distribution)

How can we define the convolution of $T \in \mathcal{S}^*(\mathbb{R}^n)$ with $f \in \mathcal{S}(\mathbb{R}^n)$? $f * T_g$ should be equal to T_{f*g} . For $h \in \mathcal{S}(\mathbb{R}^n)$ holds:

$$\begin{aligned} T_{f*g}(h) &= \int_{\mathbb{R}^n} (f * g)(x) \cdot h(x) d^n x = \\ &= \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^n} f(x - y) g(y) d^n y \right) \cdot h(x) d^n x \end{aligned}$$

By Fubini's theorem we may interchange the order of integration to get:

$$\begin{aligned} T_{f*g}(h) &= \int_{\mathbb{R}^n} g(y) \left(\int_{\mathbb{R}^n} f(x - y) \cdot h(x) d^n x \right) d^n y = \\ &\stackrel{\tilde{f}(z) := f(-z)}{=} \int_{\mathbb{R}^n} g(y) \left(\int_{\mathbb{R}^n} \tilde{f}(y - x) \cdot h(x) d^n x \right) d^n y = \\ &= \int_{\mathbb{R}^n} g(y) (\tilde{f} * h)(y) d^n y = T_g(\tilde{f} * h) \end{aligned}$$

So for a distribution $T \in \mathcal{S}^*(\mathbb{R}^n)$ we define the *convolution* as:

$$\begin{aligned} * : \mathcal{S}(\mathbb{R}^n) \times \mathcal{S}^*(\mathbb{R}^n) &\rightarrow \mathcal{S}^*(\mathbb{R}^n) \\ (f * T)(h) &:= T(\tilde{f} * h) \end{aligned} \tag{2.13}$$

For $S, T \in \mathcal{S}^*(\mathbb{R}^n)$, $S * T$ and $S \cdot T$ are ill-defined in general. For example $\delta(x) \cdot \delta(x)$ makes no sense, as well as $T_f * T_f$ for $f = 1$.

2.2 The Fourier Transform

First consider the Fourier transform on $\mathcal{S}(\mathbb{R}^n)$ and later on $\mathcal{S}'(\mathbb{R}^n)$.

2.2.1 Definition (Fourier Transform)

Define linear functionals \mathcal{F} and $\overline{\mathcal{F}}$ on \mathcal{S} :

$$(\mathcal{F}f)(p) := \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{-ipx} f(x) \, d^n x \quad (2.14)$$

$$(\overline{\mathcal{F}}f)(x) := \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{ipx} f(p) \, d^n p \quad (2.15)$$

The integrals are well-defined and finite, because f has suitable decay properties at infinity. An alternative convention, which is not convenient here, because it has less symmetry, is:

$$(\mathcal{F}f)(p) := \int_{\mathbb{R}^n} e^{-ipx} f(x) \, d^n x$$

$$(\overline{\mathcal{F}}f)(x) := \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{ipx} f(p) \, d^n p$$

2.2.2 Proposition (Fourier Transform)

\mathcal{F} and $\overline{\mathcal{F}}$ are well-defined linear operators from $\mathcal{S}(\mathbb{R}^n)$ to $\mathcal{S}(\mathbb{R}^n)$.

Proof

The linearity is clear. We still have to show, that all norms $\|\mathcal{F}f\|_{r,s}$ are finite. First consider the norm $\|\cdot\|_{0,0}$:

$$\begin{aligned} |(\mathcal{F}f)(p)| &\leq \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} |f(x)| \cdot \frac{(1+|x|)^{n+1}}{(1+|x|)^{n+1}} \, d^n x \leq \\ &\leq \frac{1}{(2\pi)^{\frac{n}{2}}} \|f\|_{n+1,0} \underbrace{\int_{\mathbb{R}^n} \frac{d^n x}{(1+|x|)^{n+1}}}_{<\infty} \leq c \|f\|_{n+1,0} \end{aligned}$$

Now we consider $|p^\alpha D^\beta (\mathcal{F}f)(p)|$.

$$\begin{aligned} \frac{\partial}{\partial p^j} (\mathcal{F}f)(p) &= \frac{\partial}{\partial p^j} \int_{\mathbb{R}^n} e^{-ipx} f(x) \, d^n x = \dots = \int_{\mathbb{R}^n} \left(\frac{\partial}{\partial p^j} e^{-ipx} \right) f(x) \, d^n x = \\ &= \int_{\mathbb{R}^n} (-ix^j) e^{-ipx} f(x) \, d^n x \end{aligned}$$

That the derivative and the integral can be interchanged is shown as follows:

$$\begin{aligned} \frac{\partial}{\partial p^j} \int_{\mathbb{R}^n} e^{-ipx} f(x) \, d^n x &= \lim_{\varepsilon \searrow 0} \int_{\mathbb{R}^n} \frac{e^{-i(p+\varepsilon e_j)x} - e^{-ipx}}{\varepsilon} f(x) \, d^n x \\ \frac{e^{-i(p+\varepsilon e_j)x} - e^{-ipx}}{\varepsilon} &= \frac{1}{\varepsilon} \int_0^1 \frac{d}{d\tau} e^{-i(p+\varepsilon \tau e_j)x} \, d\tau = -ie_j x \int_0^1 e^{-i(p+\varepsilon \tau e_j)x} \, d\tau \end{aligned}$$

$$\Rightarrow \left| \frac{e^{-i(p+\varepsilon e_j)x} - e^{-ipx}}{\varepsilon} \right| \leq \|x\| \cdot \underbrace{\int_0^1 \left| e^{-i(p+\varepsilon \tau e_j)x} \right| d\tau}_{=1} = \|x\|$$

$$\left| \frac{e^{-i(p+\varepsilon e_j)x} - e^{-ipx}}{\varepsilon} f(x) \right| \leq \|x\| \cdot |f(x)| \leq \frac{\|x\|}{1 + \|x\|^{n+2}} \|f\|_{n+2,0} =: h(x) \in L^1(\mathbb{R}^n)$$

This allows us to apply the dominated convergence theorem to take the limit $\varepsilon \rightarrow 0$ inside the integral. Iteration of this process gives:

$$\begin{aligned} D^\beta(\mathcal{F}f)(p) &= \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} (-i)^{|\beta|} x^\beta f(x) e^{-ipx} d^n x \\ p^\alpha D^\beta(\mathcal{F}f)(p) &= \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} (-i)^{|\beta|} x^\beta f(x) p^\alpha e^{-ipx} d^n x \\ p^j e^{-ipx} &= i \frac{\partial}{\partial x^j} e^{-ipx} \\ \Rightarrow p^\alpha e^{-ipx} &= i^{|\alpha|} D_x^\alpha e^{-ipx} \\ p^\alpha D^\beta(\mathcal{F}f)(p) &= \frac{(-i)^{|\beta|} i^{|\alpha|}}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \underbrace{(x^\beta f(x))}_{\text{rapid decay}} (D_x^\alpha e^{-ipx}) d^n x = \\ &\stackrel{\substack{\text{integration} \\ \text{by parts}}}{=} \frac{(-i)^{|\beta|} i^{|\alpha|}}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} (-1)^{|\alpha|} D_x^\alpha (x^\beta f(x)) e^{-ipx} d^n x = \\ &= \frac{(-i)^{|\alpha|+|\beta|}}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} D_x^\alpha (x^\beta f(x)) e^{-ipx} d^n x \end{aligned}$$

From the computation we did earlier in the proof we know:

$$|D_x^\alpha(\mathcal{F}f)(p)| \leq C \left\| D_x^\alpha (x^\beta f) \right\|_{n+1,0} \leq \tilde{C}(\alpha, \beta) \|f\|_{|\beta|+n+1,|\alpha|}$$

$$\|\mathcal{F}f\|_{r,s} \leq \tilde{c}(s, r, n) \|f\|_{s+n+1,r}$$

Therefore $\mathcal{F} : \mathcal{S}(\mathbb{R}^n) \rightarrow \mathcal{S}(\mathbb{R}^n)$ is well-defined. The same follows analogously for $\overline{\mathcal{F}}$. $\square_{2.2.2}$

So we have the following correspondence:

$$-ix^j \leftrightarrow \frac{\partial}{\partial p^j} \tag{2.16}$$

$$-i \frac{\partial}{\partial x^j} \leftrightarrow p^j \tag{2.17}$$

Here the derivatives always act on f or $\mathcal{F}f$ and not on e^{-ipx} .

$$x^\alpha D^\beta f \leftrightarrow i^{|\alpha|+|\beta|} D_p^\alpha p^\beta (\mathcal{F}f)(p) = i^{|\alpha|+|\beta|} p^\beta D^\alpha (\mathcal{F}f)(p) + \text{lower order terms}$$

Suppose we had worked with $\|f\|_{0,k} = \|f\|_{C^k}$ as family of norms. Then the norms of the Fourier transform of a function with finite norms would not necessarily be finite.

2.2.3 Theorem (Plancherel, Convergence Generating Factor)

\mathcal{F} and $\overline{\mathcal{F}}$ are inverse to each other:

$$\overline{\mathcal{F}}\mathcal{F} = \mathcal{F}\overline{\mathcal{F}} = \mathbb{1} : \mathcal{S}(\mathbb{R}^n) \rightarrow \mathcal{S}(\mathbb{R}^n) \tag{2.18}$$

Proof

$$\begin{aligned}
 (\mathcal{F}\overline{\mathcal{F}}f)(p) &= \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} e^{-ipx} \underbrace{\left(\int_{\mathbb{R}^n} e^{iqx} f(q) d^n q \right)}_{(\overline{\mathcal{F}}f)(x)} d^n x \stackrel{?}{=} f(p) \\
 &\stackrel{?}{=} \frac{1}{(2\pi)^n} \int_{\mathbb{R}^n} f(q) \left(\int_{\mathbb{R}^n} e^{-i(p-q)x} d^n x \right) d^n q
 \end{aligned}$$

The problem here is, that $e^{-i(p-q)x}$ does not decay at infinity, so the integral is not well-defined. Instead we have to introduce a *convergence generating factor* $e^{-\varepsilon x^2}$ and, after integrating, calculate the limes $\varepsilon \rightarrow 0$.

$$\begin{aligned}
 (\mathcal{F}\overline{\mathcal{F}}f)(p) &= \frac{1}{(2\pi)^n} \lim_{\varepsilon \searrow 0} \int_{\mathbb{R}^n} e^{-ipx} e^{-\varepsilon x^2} \left(\int_{\mathbb{R}^n} e^{iqx} f(q) d^n q \right) d^n x = \\
 &\stackrel{\text{Fubini}}{=} \frac{1}{(2\pi)^n} \lim_{\varepsilon \searrow 0} \int_{\mathbb{R}^n} f(q) \left(\int_{\mathbb{R}^n} e^{-i(p-q)x} e^{-\varepsilon x^2} d^n x \right) d^n q
 \end{aligned}$$

The resulting Gaussian integral can be computed in closed form. In one dimension it is:

$$\begin{aligned}
 \int_{\mathbb{R}} e^{-i\lambda x} e^{-\varepsilon x^2} dx &= \int_{\mathbb{R}} e^{-\varepsilon \left(x + \frac{i\lambda}{2\varepsilon}\right)^2 - \frac{\lambda^2}{4\varepsilon}} dx = e^{-\frac{\lambda^2}{4\varepsilon}} \int_{\mathbb{R}} e^{-\varepsilon \left(x + \frac{i\lambda}{2\varepsilon}\right)^2} dx = \\
 &\stackrel{z = \sqrt{\varepsilon} \left(x + \frac{i\lambda}{2\varepsilon}\right)}{=} e^{-\frac{\lambda^2}{4\varepsilon}} \int_{\mathbb{R} + \frac{i\lambda}{2\sqrt{\varepsilon}}} e^{-z^2} \frac{dz}{\sqrt{\varepsilon}} = \\
 &\stackrel{\text{contour deformation}}{=} \frac{e^{-\frac{\lambda^2}{4\varepsilon}}}{\sqrt{\varepsilon}} \underbrace{\int_{\mathbb{R}} e^{-z^2} dz}_{=\sqrt{\pi}} = \sqrt{\frac{\pi}{\varepsilon}} e^{-\frac{\lambda^2}{4\varepsilon}}
 \end{aligned}$$

So we get:

$$(\mathcal{F}\overline{\mathcal{F}}f)(p) = \frac{1}{(2\pi)^n} \lim_{\varepsilon \searrow 0} \int_{\mathbb{R}^n} f(q) \left(\frac{\pi}{\varepsilon} \right)^{\frac{n}{2}} e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q$$

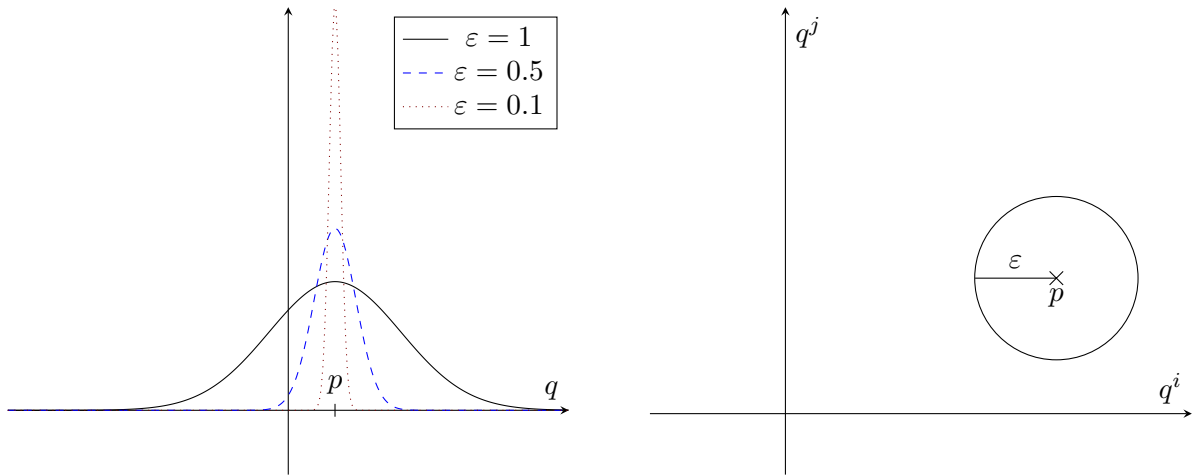


Figure 2.2: The Gaussian gets very narrow and very high as ε decreases.

Estimate the integral as follows:

$$(\mathcal{F}\overline{\mathcal{F}}f)(p) = \lim_{\varepsilon \searrow 0} \frac{1}{(4\pi\varepsilon)^{\frac{n}{2}}} \left(\int_{\mathbb{R}^n} f(p) e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q + \int_{\mathbb{R}^n} (f(q) - f(p)) e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q \right)$$

The first integral gives:

$$\int_{\mathbb{R}^n} e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q \stackrel{\substack{z=\frac{p-q}{2\sqrt{\varepsilon}} \\ d^n z = \frac{d^n q}{(4\varepsilon)^{\frac{n}{2}}}}}{=} (4\varepsilon)^{\frac{n}{2}} \underbrace{\int_{\mathbb{R}^n} e^{-z^2} d^n q}_{=\pi^{\frac{n}{2}}} = (4\pi\varepsilon)^{\frac{n}{2}}$$

So we get:

$$(\mathcal{F}\overline{\mathcal{F}}f)(p) = f(p) + \lim_{\varepsilon \searrow 0} \frac{1}{(4\pi\varepsilon)^{\frac{n}{2}}} \int_{\mathbb{R}^n} (f(q) - f(p)) e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q$$

It remains to show that the second summand goes to zero for $\varepsilon \rightarrow 0$. We use the following scaling argument:

$$\begin{aligned} \frac{1}{\varepsilon^{\frac{n}{2}}} \int_{\mathbb{R}^n} (f(q) - f(p)) e^{-\frac{(p-q)^2}{4\varepsilon}} d^n q &\stackrel{\substack{u=\frac{p-q}{\sqrt{\varepsilon}} \\ d^n q = \varepsilon^{\frac{n}{2}} d^n u}}{=} \frac{1}{\varepsilon^{\frac{n}{2}}} \int_{\mathbb{R}^n} (f(p - \sqrt{\varepsilon}u) - f(p)) e^{-\frac{u^2}{4}} \varepsilon^{\frac{n}{2}} d^n u = \\ &= \int_{\mathbb{R}^n} \underbrace{(f(p - \sqrt{\varepsilon}u) - f(p)) e^{-\frac{u^2}{4}}}_{\xrightarrow{\varepsilon \searrow 0} 0 \text{ pointwise}} d^n u \end{aligned}$$

For the integrand holds:

$$\underbrace{(f(p - \sqrt{\varepsilon}u) - f(p)) e^{-\frac{u^2}{4}}}_{\xrightarrow{\varepsilon \searrow 0} 0 \text{ pointwise}} \leq \|f\|_{0,0} e^{-\frac{u^2}{4}} \in L^1(\mathbb{R}^n)$$

So the dominated convergence theorem can be applied to get:

$$(\mathcal{F}\overline{\mathcal{F}}f)(p) = f(p) + \lim_{\varepsilon \searrow 0} \frac{1}{(4\pi)^{\frac{n}{2}}} \underbrace{\int_{\mathbb{R}^n} \lim_{\varepsilon \searrow 0} (f(p - \sqrt{\varepsilon}u) - f(p)) e^{-\frac{u^2}{4}} d^n u}_{=0} = f(p)$$

$\overline{\mathcal{F}}\mathcal{F} = \mathbb{1}$ follows analogously. □_{2.2.3}

We want to generalize the Fourier transform to $\mathcal{S}^*(\mathbb{R}^n)$. We begin with the case T_g with $g \in \mathcal{S}(\mathbb{R}^n)$. We want:

$$\mathcal{F}(T_g) = T_{\mathcal{F}g}$$

$$T_{\mathcal{F}g}(f) = \int_{\mathbb{R}^n} (\mathcal{F}g)(p) f(p) d^n p = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^n} e^{-ipx} g(x) d^n x \right) f(p) d^n p$$

Fubini's theorem allows us to interchange the order of integration:

$$T_{\mathcal{F}g}(f) = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} g(x) \underbrace{\left(\int_{\mathbb{R}^n} e^{-ipx} f(p) d^n p \right)}_{=\mathcal{F}f(x)} d^n x = \int_{\mathbb{R}^n} g(x) (\mathcal{F}f)(x) d^n x = T_g(\mathcal{F}f)$$

Since we want $(\mathcal{F}T_g)(f) = T_{\mathcal{F}g} = T_g(\mathcal{F}f)$, this motivates the following general definition:

2.2.4 Definition (Fourier Transform of Distributions)

$\mathcal{F}, \overline{\mathcal{F}} : \mathcal{S}^*(\mathbb{R}^n) \rightarrow \mathcal{S}^*(\mathbb{R}^n)$ are defined by their action on a test function $f \in \mathcal{S}(\mathbb{R}^n)$:

$$(\mathcal{F}T)(f) := T(\mathcal{F}f) \quad (2.19)$$

$$(\overline{\mathcal{F}}T)(f) := T(\overline{\mathcal{F}}f) \quad (2.20)$$

It holds:

$$|(\mathcal{F}T)(f)| = |T(\mathcal{F}f)| \stackrel{T \in \mathcal{S}^*(\mathbb{R}^n)}{\leq} \underset{\Rightarrow \exists r, s \in \mathbb{N}, c \in \mathbb{R}_{>0}}{c} \|\mathcal{F}f\|_{r,s} \leq \tilde{c} \|f\|_{s+n+1,r}$$

Thus $\mathcal{F}T$ is indeed a tempered distribution.

2.2.5 Theorem (Plancherel for Distributions)

Plancherel's theorem holds on $\mathcal{S}^*(\mathbb{R}^n)$ as well:

$$\mathcal{F}\overline{\mathcal{F}} = \overline{\mathcal{F}}\mathcal{F} = \mathbb{1}_{\mathcal{S}^*(\mathbb{R}^n)} \quad (2.21)$$

Proof

$$(\mathcal{F}\overline{\mathcal{F}}T)(f) \stackrel{\text{Definition 2.2.4}}{=} (\overline{\mathcal{F}}T)(\mathcal{F}f) = T(\overline{\mathcal{F}}\mathcal{F}f) \stackrel{\text{Plancherel 2.2.3}}{=} T(f)$$

Since this holds for all $f \in \mathcal{S}(\mathbb{R}^n)$ and all $T \in \mathcal{S}^*(\mathbb{R}^n)$ it follows:

$$\mathcal{F}\overline{\mathcal{F}} = \mathbb{1}_{\mathcal{S}^*(\mathbb{R}^n)}$$

The same follows for $\overline{\mathcal{F}}\mathcal{F}$.

□_{2.2.5}

2.2.6 Examples

1. The Fourier transform of the δ -Distribution can be calculated as follows:

$$\begin{aligned} (\mathcal{F}\delta)(f) &= \delta(\mathcal{F}f) = (\mathcal{F}f)(0) = \\ &= \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{-ipx} f(x) \, dx \Big|_{p=0} = \int_{\mathbb{R}^n} \frac{1}{(2\pi)^{\frac{n}{2}}} f(x) \, dx = T_{(2\pi)^{-\frac{n}{2}}}(f) \end{aligned}$$

This means:

$$\mathcal{F}\delta = T_{(2\pi)^{-\frac{n}{2}}}$$

Or, in a more computational manner, one can write this as:

$$\mathcal{F}\delta = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{-ipx} \delta(x) \, dx = \frac{1}{(2\pi)^{\frac{n}{2}}}$$

This is not satisfying from a mathematical point of view, because one does not know, that the usual formula for the Fourier transform also works for distributions.

2. Consider T_f with $f(p) = e^{ipy}$ for a given $y \in \mathbb{R}^n$.

$$\begin{aligned} (\mathcal{F}T_f)(h) &= T_f(\mathcal{F}h) = T_f\left(\frac{1}{(2\pi)^{\frac{n}{2}}} \int e^{-ipx} h(x) \, d^n x\right) = \\ &= \frac{1}{(2\pi)^{\frac{n}{2}}} \int e^{ipy} \left(\int e^{-ipx} h(x) \, d^n x\right) d^n p = \\ &= (2\pi)^{\frac{n}{2}} (\overline{\mathcal{F}\mathcal{F}}) h(y) = (2\pi)^{\frac{n}{2}} h(y) \end{aligned}$$

So we get:

$$\begin{aligned} \mathcal{F}T_f h &= (2\pi)^{\frac{n}{2}} h(y) \\ (\mathcal{F}T_f)(x) &= (2\pi)^{\frac{n}{2}} \delta^{(n)}(x - y) \end{aligned}$$

Formally one can write:

$$(\mathcal{F}f)(x) = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{-ipx} e^{ipy} d^n p = \frac{1}{(2\pi)^{\frac{n}{2}}} \int_{\mathbb{R}^n} e^{-ip(x-y)} d^n p$$

This is ill-defined, but physicists use the formal relation:

$$\int_{\mathbb{R}^n} e^{-ip(x-y)} d^n p = (2\pi)^n \delta^{(n)}(x - y)$$

3. Consider $T = T_g$ with $g(p) = p^2 e^{ipy}$ for a given $y \in \mathbb{R}$.

$$\begin{aligned} (\mathcal{F}T_g)(f) &= T_g(\mathcal{F}f) = \int_{-\infty}^{\infty} g(p) (\mathcal{F}f)(p) \, dp = \\ &= \int g(p) \left(\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-ipx} \, dx \right) dp \end{aligned}$$

One cannot interchange the integrals, because the integral

$$\int_{-\infty}^{\infty} \underbrace{g(p) e^{-ip\alpha}}_{\notin L^1} \, dp$$

does not exist. Therefore we work again with a convergence generating factor, which we can due to the dominated convergence theorem:

$$\begin{aligned} (\mathcal{F}T_g)(f) &= \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} g(p) e^{-\varepsilon|p|} \left(\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(x) e^{-ipx} \, dx \right) dp = \\ &= \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \left(\int_{-\infty}^{\infty} g(p) e^{-\varepsilon|p|} e^{-ipx} \, dp \right) dx = \\ &= \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \left(\int_{-\infty}^{\infty} p^2 e^{ipy} e^{-\varepsilon|p|} e^{-ipx} \, dp \right) dx = \\ &= \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \left(\int_{-\infty}^{\infty} (\mathbf{i}\partial_x)^2 e^{-ip(x-y)-\varepsilon|p|} \, dp \right) dx = \\ &\stackrel{\text{Lebesgue theorem}}{=} \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) (\mathbf{i}\partial_x)^2 \left(\int_{-\infty}^{\infty} e^{-ip(x-y)-\varepsilon|p|} \, dp \right) dx \end{aligned}$$

Now one can decompose the integral into integrals from $-\infty$ to 0 and from 0 to ∞ and calculate the result.

$$\begin{aligned}
 (\mathcal{F}T_g)(f) &= \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) (\mathbf{i}\partial_x)^2 \left(\int_{-\infty}^0 e^{-\mathbf{i}p(x-y)+\varepsilon p} dp + \int_0^{\infty} e^{-\mathbf{i}p(x-y)-\varepsilon p} dp \right) dx = \\
 &= \frac{-1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \partial_x^2 \left(\frac{1}{-\mathbf{i}(x-y)+\varepsilon} - \frac{1}{-\mathbf{i}(x-y)-\varepsilon} \right) dx = \\
 &= \frac{-1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \partial_x^2 \left(\frac{-\mathbf{i}(x-y)-\varepsilon - (-\mathbf{i}(x-y)+\varepsilon)}{-(x-y)^2 - \varepsilon^2} \right) dx = \\
 &= \frac{1}{\sqrt{2\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f(x) \partial_x^2 \left(\frac{2\varepsilon}{(x-y)^2 + \varepsilon^2} \right) dx = \\
 &\stackrel{\text{integration by parts}}{=} \sqrt{\frac{2}{\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f''(x) \cdot \frac{\varepsilon}{(x-y)^2 + \varepsilon^2} dx = \\
 &\stackrel{z:=\frac{x-y}{\varepsilon}}{=} \sqrt{\frac{2}{\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f''(\varepsilon z + y) \cdot \frac{\varepsilon}{\varepsilon^2 z^2 + \varepsilon^2} \varepsilon dz = \\
 &= \sqrt{\frac{2}{\pi}} \lim_{\varepsilon \searrow 0} \int_{-\infty}^{\infty} f''(\varepsilon z + y) \cdot \frac{1}{z^2 + 1} dz = \\
 &= \sqrt{\frac{2}{\pi}} f''(y) \underbrace{\int_{-\infty}^{\infty} \frac{1}{z^2 + 1} dz}_{=\pi} = \sqrt{2\pi} f''(y)
 \end{aligned}$$

So we get:

$$(\mathcal{F}T_g)(x) = \sqrt{2\pi} \delta''(x - y)$$

4. Consider the following Hilbert space:

$$L^2(\mathbb{R}^n) = \left\{ f : \mathbb{R}^n \rightarrow \mathbb{R} \mid f \text{ measurable with } \int |f|^2 d^n x < \infty \right\}$$

For $f \in L^2(\mathbb{R}^n)$ define:

$$T_f(g) := \int_{\mathbb{R}^n} f(x) g(x) d^n x$$

$$|T_f(g)| \stackrel{\text{Schwarz}}{\leq} \|f\|_{L^2(\mathbb{R}^n)} \|g\|_{L^2(\mathbb{R}^n)} \leq \|f\|_{L^2(\mathbb{R}^n)} \cdot c(n) \|g\|_{n+1,0}$$

Here we used:

$$|g(x)| \leq \tilde{c}(n) \frac{\|g\|_{n+1,0}}{(1+|x|)^{n+1}}$$

So we have $T_f \in \mathcal{S}^*(\mathbb{R}^n)$.

$$\begin{aligned}
 L^2(\mathbb{R}^n) &\hookrightarrow \mathcal{S}^*(\mathbb{R}^n) \xrightarrow{\mathcal{F}} \mathcal{S}^*(\mathbb{R}^n) \\
 f &\mapsto T_f \mapsto \mathcal{F}T_f
 \end{aligned}$$

2.2.7 Theorem (Fourier Transform is isometry)

The mappings $\mathcal{F}, \overline{\mathcal{F}} : L^2(\mathbb{R}^n) \rightarrow L^2(\mathbb{R}^n)$ are isometries, i.e.:

$$\|f\|_{L^2(\mathbb{R}^n)} = \|\mathcal{F}f\|_{L^2(\mathbb{R}^n)} \quad (2.22)$$

Due to $\mathcal{F}\overline{\mathcal{F}} = 1$ they are even unitary transformations.

Proof

Consider first $f, g \in \mathcal{S}(\mathbb{R}^n) \subseteq L^2(\mathbb{R}^n)$.

$$\begin{aligned} \langle \mathcal{F}f, \mathcal{F}g \rangle_{L^2(\mathbb{R}^n)} &= \int_{\mathbb{R}^n} \overline{(\mathcal{F}f)(x)} (\mathcal{F}g)(x) \, dx \stackrel{?}{=} \int \bar{f}(x) g(x) \, dx \\ &= T_{\bar{\mathcal{F}}\bar{f}}(\mathcal{F}g) = \left(\bar{\mathcal{F}} T_{\bar{f}} \right) (\mathcal{F}g) = T_{\bar{f}}(\bar{\mathcal{F}}\mathcal{F}g) = \\ &\bar{\mathcal{F}}^{\mathcal{F}=1} T_{\bar{f}}(g) = \int \bar{f}(x) g(x) \, dx \end{aligned}$$

$\mathcal{S}(\mathbb{R}^n)$ is dense in $L^2(\mathbb{R}^n)$. Because \mathcal{F} is continuous, \mathcal{F} is also isometric on $L^2(\mathbb{R}^n)$. For $f \in L^2(\mathbb{R}^n)$ choose $f_n \in \mathcal{S}(\mathbb{R}^n)$ with $f_n \rightarrow f$ converging in $L^2(\mathbb{R}^n)$. Then holds for all $n \in \mathbb{N}$:

$$\|f_n\|_{L^2} = \|\mathcal{F}f_n\|$$

In the limit $n \rightarrow \infty$ we get, since \mathcal{F} is continuous:

$$\|f\| = \|\mathcal{F}f\|$$

□_{2.2.7}

2.3 Applications to Partial Differential Equations with Constant Coefficients

Consider as example the Poisson equation

$$\Delta u = f$$

in \mathbb{R}^n with a given f and assume for simplicity $f \in \mathcal{S}(\mathbb{R}^n)$. After a Fourier transform and defining $\hat{u} := \mathcal{F}u$ and $\hat{f} = \mathcal{F}f$ we get:

$$\begin{aligned} \left(-\|p\|^2 \right) \hat{u}(p) &= \hat{f}(p) \\ \Rightarrow \hat{u}(p) &= -\frac{\hat{f}(p)}{\|p\|^2} \end{aligned}$$

Then $\bar{\mathcal{F}}\hat{u}$ is a solution of the Poisson equation.

- For $\|p\|^{-2} \hat{f}(p) \in \mathcal{S}(\mathbb{R}^n)$ the method works directly and one gets a $u \in \mathcal{S}(\mathbb{R}^n)$.
- In the case of $n \geq 3$, $\hat{u} = -\|p\|^{-2} \hat{f}(p)$ is a regular distribution. (If $n < 3$ the integral does not necessarily converge.)

$$(T_{\hat{u}})(g) := \int \left(-\frac{\hat{f}(p)}{\|p\|^2} \right) g(p) \, d^n p$$

Therefore $u := \bar{\mathcal{F}}T_{\hat{u}}$ is a distributional solution of the Poisson equation, so we get $\Delta u = T_f$.

Problem: The distributional solution is not unique, because e.g.

$$\hat{u}(p) = -\frac{\hat{f}(p)}{\|p\|^2} + c\delta(p)$$

is also a solution. Therefore we have to specify the behavior of $u(x)$ in the limit $\|x\| \rightarrow \infty$.

3 The Laplace Equation in $\Omega \subseteq \mathbb{R}^n$

In this chapter we always consider an open subset $\Omega \subseteq \mathbb{R}^n$ with boundary $\partial\Omega$. With the Laplacian

$$\Delta = \frac{\partial^2}{\partial x_1^2} + \dots + \frac{\partial^2}{\partial x_n^2} \quad (3.1)$$

the Laplace equation can be written as:

$$\Delta u = 0$$

The inhomogeneous Laplace equation is called Poisson equation:

$$\Delta u = f$$

For both one needs to specify boundary conditions on $\partial\Omega$ to get a unique solution.

3.0 Reminder

3.0.1 Theorem (Gauss's Theorem)

If Y is a smooth vector field on the closure $\overline{\Omega}$ and $\partial\Omega$ is smooth with outer normal ν , then holds:

$$\int_{\Omega} \operatorname{div}(Y) \underbrace{\mathrm{d}^n x}_{=\mathrm{d}\mu} = \int_{\partial\Omega} \langle Y, \nu \rangle \mathrm{d}\mu_{\partial\Omega} \quad (3.2)$$

(Proof omitted)

3.0.2 Theorem (Green's Identities)

For $u, w \in C^\infty(\overline{\Omega})$ holds:

$$\int_{\Omega} w \Delta u \mathrm{d}\mu_{\Omega} = \int_{\partial\Omega} w (\nabla_{\nu} u) \mathrm{d}\mu_{\partial\Omega} - \int_{\Omega} \langle \nabla w, \nabla u \rangle \mathrm{d}\mu_{\Omega} \quad (3.3)$$

$$\int_{\Omega} (w (\Delta u) - (\Delta w) u) \mathrm{d}\mu_{\Omega} = \int_{\partial\Omega} (w (\nabla_{\nu} u) - (\nabla_{\nu} w) u) \mathrm{d}\mu_{\partial\Omega} \quad (3.4)$$

Proof

We use $\Delta = \operatorname{div} \operatorname{grad}$ and integrate by parts:

$$\int_{\Omega} w \Delta u \, d\mu_{\Omega} = \int_{\Omega} w \cdot \operatorname{div} (\nabla u) \, d\mu_{\Omega} = \int_{\Omega} \operatorname{div} (w \nabla u) - \langle \nabla w, \nabla u \rangle \, d\mu_{\Omega}$$

Then one can use Gauss's theorem to get the first identity. Now one subtracts the identity with w and u commuted:

$$\begin{aligned} w \Delta u - u \Delta w &= \operatorname{div} (w \nabla u) - \langle \nabla w, \nabla u \rangle - (\operatorname{div} (u \nabla w) - \langle \nabla u, \nabla w \rangle) = \\ &\stackrel{\langle \cdot, \cdot \rangle \text{ symmetric}}{=} \operatorname{div} (w \nabla u) - \operatorname{div} (u \nabla w) \end{aligned}$$

Using Gauss's theorem the second identity follows. □_{3.0.2}

3.1 Representation Formulas for Harmonic Functions

3.1.1 Definition (Harmonic Functions)

A function $u \in C^2(\overline{\Omega})$ is called *harmonic*, if the Laplacian vanishes:

$$\Delta u = 0$$

The harmonic functions form a vector space.

Examples

- Constant or linear functions
- $u(x_1, x_2) = x_1^2 - x_2^2$ is harmonic on \mathbb{R}^2 .
- Holomorphic functions on $\Omega \subseteq \mathbb{C} \cong \mathbb{R}^2$

Consider now *spherically symmetric harmonic functions*. For this we choose polar coordinates (r, ϑ, φ) in \mathbb{R}^3 :

$$\begin{aligned} x &= r \cos(\vartheta) \\ y &= r \sin(\vartheta) \cos(\varphi) \\ z &= r \sin(\vartheta) \sin(\varphi) \end{aligned}$$

More general in \mathbb{R}^n with $n \in \mathbb{N}_{\geq 2}$ we choose $r = \|x\|$ and $\omega \in S^{n-1}$. Regard \mathbb{R}^n with the Euclidian metric as a Riemannian manifold (M, g) . Polar coordinates give a special chart on $\Omega \subseteq M$ with $0 \notin \Omega$. Then we can calculate $\Delta = \Delta_{\omega}$ as Laplace-Beltrami operator in polar coordinates. The metric is:

$$g = \begin{pmatrix} 1 & 0 \\ 0 & r^2 g_{S^{n-1}} \end{pmatrix} \qquad g^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & r^{-2} g_{S^{n-1}}^{-1} \end{pmatrix}$$

$$\begin{aligned} \det(g) &= r^{2(n-1)} g_{S^{n-1}} \\ \sqrt{\det(g)} &= r^{n-1} \sqrt{\det(g_{S^{n-1}})} \end{aligned}$$

Now the Laplace-Beltrami operator can be calculated with the *Koszul formula*:

$$\begin{aligned}
 \Delta \backslash u &= \nabla_j \nabla^j u = \frac{1}{\sqrt{\det(g)}} \frac{\partial}{\partial x^j} \left(\sqrt{\det(g)} g^{jk} \frac{\partial}{\partial x^k} u \right) = \\
 &= \frac{1}{r^{n-1} \sqrt{\det(g_{S^{n-1}})}} \frac{\partial}{\partial x^j} \left(r^{n-1} \sqrt{\det(g_{S^{n-1}})} g^{jk} \frac{\partial}{\partial x^k} u \right) = \\
 &= \frac{1}{r^{n-1} \sqrt{\det(g_{S^{n-1}})}} \frac{\partial}{\partial r} \left(r^{n-1} \sqrt{\det(g_{S^{n-1}})} \frac{\partial}{\partial r} u \right) + \\
 &\quad + \sum_{j,k=2}^n \frac{1}{r^{n-1} \sqrt{\det(g_{S^{n-1}})}} \frac{\partial}{\partial x^j} \left(r^{n-1} \sqrt{\det(g_{S^{n-1}})} \frac{1}{r^2} (g_{S^{n-1}})^{jk} \frac{\partial}{\partial x^k} u \right) = \\
 &= \frac{1}{r^{n-1}} \frac{\partial}{\partial r} \left(r^{n-1} \frac{\partial}{\partial r} u \right) + \frac{1}{r^2} \Delta \backslash_{S^{n-1}} u
 \end{aligned} \tag{3.5}$$

The important formula is:

$$\Delta \backslash = \frac{1}{r^{n-1}} \frac{\partial}{\partial r} \left(r^{n-1} \frac{\partial}{\partial r} \cdot \right) + \frac{1}{r^2} \Delta \backslash_{S^{n-1}} \tag{3.6}$$

For spherically symmetric solutions $\Gamma \in C^\infty(\mathbb{R}^n \setminus \{0\})$, the Laplace equation reads:

$$\Delta \backslash \Gamma = \frac{1}{r^{n-1}} \frac{\partial}{\partial r} \left(r^{n-1} \frac{\partial}{\partial r} \Gamma(r) \right) = 0$$

This gives:

$$\begin{aligned}
 r^{n-1} \frac{\partial}{\partial r} \Gamma(r) &= c \\
 \frac{\partial}{\partial r} \Gamma(r) &= cr^{1-n}
 \end{aligned}$$

$$\Gamma(r) = a + C \int^r \tau^{1-n} d\tau = \begin{cases} a + \tilde{c} \ln(r) & \text{if } n = 2 \\ a + \frac{c}{r^{n-2}} & \text{if } n > 2 \end{cases}$$

Now we choose specific values for a and c or \tilde{c} .

3.1.2 Definition (Fundamental Solution of Laplace Equation)

The *fundamental solution* Γ of the Laplace equation in \mathbb{R}^n is defined by:

$$\Gamma(x, y) = \Gamma(\|x - y\|) := \begin{cases} \frac{1}{2\pi} \ln(\|x - y\|) & \text{if } n = 2 \\ \frac{1}{n(2-n)\omega_n} \|x - y\|^{2-n} & \text{if } n > 2 \end{cases} \tag{3.7}$$

Here $\omega_n := \mu(B_1(0))$ is the Lebesgue measure of the unit ball.

For example for $n = 3$ we have $\omega_3 = \frac{4\pi}{3}$ and thus:

$$\Gamma(x, y) = \frac{1}{3(-1)\frac{4\pi}{3}} \cdot \frac{1}{\|x - y\|} = -\frac{1}{4\pi} \frac{1}{\|x - y\|} \tag{3.8}$$

3.1.3 Theorem (Green's representation)

Let $\Omega \subseteq \mathbb{R}^n$ be open with smooth boundary and $u \in C^2(\overline{\Omega})$. Then for any $y \in \Omega$ holds:

$$u(y) = \int_{\partial\Omega} (u(x) \nabla_\nu \Gamma(x, y) - \Gamma(x, y) \nabla_\nu u(x)) d\mu_{\partial\Omega}(x) + \int_{\Omega} \Gamma(x, y) (\Delta u)(x) d\mu(x) \quad (3.9)$$

For harmonic functions with $\Delta u = 0$, this simplifies to:

$$u(y) = \int_{\partial\Omega} (u(x) \nabla_\nu \Gamma(x, y) - \Gamma(x, y) \nabla_\nu u(x)) d\mu_{\partial\Omega}(x) \quad (3.10)$$

Thus u has an explicit representation in terms of its boundary values on $\partial\Omega$.

Proof

Choose $\varepsilon \in \mathbb{R}_{>0}$ such that $B_\varepsilon(y) \subseteq \Omega$.

$$\int_{\Omega \setminus B_\varepsilon(y)} \underbrace{\Gamma(x, y)}_{\text{smooth}} \underbrace{(\Delta u)(x)}_{\text{continuous}} d\mu(x)$$

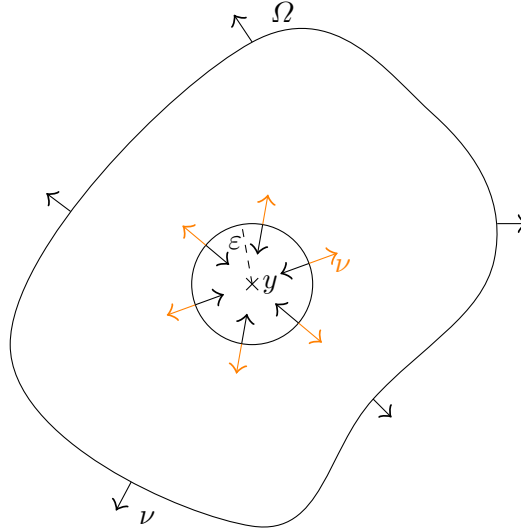


Figure 3.1: Outer normal ν of $\Omega \setminus B_\varepsilon(y)$ (black) and of $B_\varepsilon(y)$ (orange)

We apply the second Green's identity to obtain with $v(x) := \Gamma(x, y)$:

$$\begin{aligned} \int_{\Omega \setminus B_\varepsilon(y)} (v \cdot \Delta u - \underbrace{(\Delta v)}_{=0} u) d\mu &= \int_{\partial(\Omega \setminus B_\varepsilon(y))} (v(\nabla_\nu u) - (\nabla_\nu v) u) d\mu_{\partial\Omega} = \\ &= \int_{\partial\Omega} (\Gamma(x, y) \nabla_\nu u(x) - (\nabla_\nu \Gamma)(x, y) u(x)) d\mu_{\partial\Omega} + \\ &\quad - \int_{\partial B_\varepsilon(y)} (\Gamma(x, y) \nabla_\nu u(x) - \nabla_\nu \Gamma(x, y) u(x)) d\mu_{\partial B_\varepsilon(y)} \end{aligned}$$

The minus in front of $\int_{\partial B_\varepsilon(y)}$ comes from the fact, that the outer normal of $\partial(\Omega \setminus B_\varepsilon(y))$ shows in the opposite direction of the outer normal of $\partial B_\varepsilon(y)$. The left side of the integral gives in the limit $\varepsilon \searrow 0$:

$$\lim_{\varepsilon \searrow 0} \int_{\Omega \setminus B_\varepsilon(y)} v \cdot \Delta u d\mu = \int_{\Omega} v \cdot \Delta u d\mu = \int_{\Omega} \Gamma(x, y) \cdot (\Delta u)(x) d\mu(x)$$

We estimate the integral over $\partial B_\varepsilon(y)$ in the limit $\varepsilon \searrow 0$:

$$\begin{aligned} \left| \int_{\partial B_\varepsilon(y)} \Gamma(x, y) \nabla_\nu u(x) d\mu_{\partial B_\varepsilon(y)} \right| &\leq \Gamma(\varepsilon) \sup_{B_\varepsilon(y)} \|\nabla u\| \cdot \underbrace{n\omega_n}_{\text{surface area of the unit sphere}} \varepsilon^{n-1} \sim \\ &\sim \begin{cases} \varepsilon & \text{if } n > 2 \\ \varepsilon \ln(\varepsilon) & \text{if } n = 2 \end{cases} \xrightarrow{\varepsilon \searrow 0} 0 \end{aligned}$$

Now we expand the second part around $\varepsilon = 0$:

$$\begin{aligned} \int_{\partial B_\varepsilon(y)} (\nabla_\nu \Gamma)(x, y) u(x) d\mu_{\partial B_\varepsilon(y)} &= \underbrace{\frac{\partial}{\partial \varepsilon} \Gamma(\varepsilon)}_{=\frac{1}{n\omega_n} \varepsilon^{1-n}} \underbrace{\int_{\partial B_\varepsilon(y)} u(x) d\mu_{B_\varepsilon(y)}}_{=u(y)n\omega_n \varepsilon^{n-1} + o_0(\varepsilon^{n-1})} = \\ &= u(y) + o_0(\varepsilon^0) \xrightarrow{\varepsilon \searrow 0} u(y) \end{aligned}$$

This gives:

$$\begin{aligned} \int_{\Omega} \Gamma(x, y) (\Delta u)(x) d\mu(x) &= \int_{\partial \Omega} (\Gamma(x, y) \nabla_\nu u(x) - (\nabla_\nu \Gamma)(x, y) u(x)) d\mu_{\partial \Omega} + u(y) \\ \Rightarrow u(y) &= \int_{\partial \Omega} (u(x) \nabla_\nu \Gamma(x, y) - \Gamma(x, y) \nabla_\nu u(x)) d\mu_{\partial \Omega}(x) + \int_{\Omega} \Gamma(x, y) (\Delta u)(x) d\mu(x) \end{aligned}$$

□_{3.1.3}

3.1.4 Corollary (Laplacian of Fundamental Solution is Delta Distribution)

For any $\varphi \in C_0^\infty(\Omega)$ holds:

$$\varphi(y) = \int_{\Omega} \Gamma(x, y) (\Delta \varphi)(x) d\mu(x) \quad (3.11)$$

This can also be expressed in terms of distributions as:

$$\Delta_x \Gamma(x, y) = \delta^{(n)}(x - y) \quad (3.12)$$

More correctly, for fixed $y \in \mathbb{R}^n$, $T(x) := T_{\Gamma(x, y)}$ defines a regular distribution. Equation (3.11) means, that for all $\varphi \in C_0^\infty(\Omega) \subseteq \mathcal{S}(\mathbb{R}^n)$:

$$\begin{aligned} (\Delta T)(\varphi) &= \varphi(y) \\ \Leftrightarrow \Delta T &= \delta_y \end{aligned}$$

Proof

Since the support of φ lies inside of Ω , the first term in Green's representation vanishes:

$$\begin{aligned}\varphi(y) &= \int_{\partial\Omega} (\varphi(x) \nabla_\nu \Gamma(x, y) - \Gamma(x, y) \nabla_\nu \varphi(x)) d\mu_{\partial\Omega}(x) + \int_{\Omega} \Gamma(x, y) (\Delta\varphi)(x) d\mu(x) = \\ &= \int_{\Omega} \Gamma(x, y) (\Delta\varphi)(x) d\mu(x)\end{aligned}$$

□_{3.1.4}

Next we investigate the existence of solutions.

$$\Delta u = 0 \quad u|_{\partial\Omega} = u_0 \quad \text{Dirichlet problem} \quad (3.13)$$

$$\Delta u = 0 \quad \nabla_\nu u|_{\partial\Omega} = u_1 \quad \text{Neumann problem} \quad (3.14)$$

The problem is, that the representation

$$u(y) = \int_{\partial\Omega} (u(x) \nabla_\nu \Gamma(x, y) - \Gamma(x, y) \nabla_\nu u(x)) d\mu_{\partial\Omega}(x)$$

needs both u and $\nabla_\nu u$ on the boundary. Suppose we want to solve the Dirichlet problem. Then $u_0 = u|_{\partial\Omega}$ is given, but $\nabla_\nu u|_{\partial\Omega}$ is unknown.

3.2 The Green's Function

Consider the Dirichlet problem for the Poisson equation:

$$\Delta u(x) = f(x) \quad \forall_{x \in \Omega} \quad u|_{\partial\Omega} = \varphi \quad (3.15)$$

Assume $u \in C^2(\overline{\Omega})$, $f \in C^0(\overline{\Omega})$ and $\varphi \in C^2(\partial\Omega)$.

3.2.1 Definition (Green's function)

A function $G(x, y)$ defined for $x, y \in \overline{\Omega}$ with $x \neq y$ is called *Green's function* for the domain Ω if the following conditions are satisfied:

- i) For all $x \in \partial\Omega$ holds $G(x, y) = 0$.
- ii) $h(x, y) := G(x, y) - \Gamma(x, y)$ is in $C^2(\Omega)$, even for $x = y$, and harmonic in $x \in \Omega$.

3.2.2 Proposition (Solution of Dirichlet Problem)

For a solution u of the Dirichlet problem for the Poisson equation holds:

$$u(y) = \int_{\partial\Omega} u(x) \nabla_\nu G(x, y) d\mu_{\partial\Omega}(x) + \int_{\Omega} f(x) G(x, y) d\mu(x) \quad (3.16)$$

Proof

$$\begin{aligned} \int_{\Omega} h(x, y) (\Delta u)(x) d\mu(x) &= \int_{\Omega} h(x, y) f(x) d\mu(x) = \\ &\stackrel{\substack{2. \text{ Green's} \\ \text{identity}}}{=} \int_{\partial\Omega} (h(x, y) (\nabla_{\nu} u)(x) - (\nabla_{\nu} h(x, y)) \cdot u(x)) d\mu_{\partial\Omega}(x) \end{aligned}$$

Now we add the Green's representation

$$\int_{\Omega} \Gamma(x, y) f(x) d\mu(x) = \int_{\partial\Omega} (\Gamma(x, y) (\nabla_{\nu} u)(x) - (\nabla_{\nu} \Gamma(x, y)) u(x)) d\mu_{\partial\Omega}(x) + u(y)$$

to get:

$$\int_{\Omega} G(x, y) f(x) d\mu(x) = \int_{\partial\Omega} (G(x, y) (\nabla_{\nu} u)(x) - (\nabla_{\nu} G(x, y)) u(x)) d\mu_{\partial\Omega}(x) + u(y)$$

Since $G(x, y) = 0$ on $\partial\Omega$, the proposition follows. $\square_{3.2.2}$

3.2.3 Theorem (Symmetry of the Green's Function)

For all $x, y \in \Omega$ with $x \neq y$ holds:

$$G(x, y) = G(y, x) \quad (3.17)$$

Proof

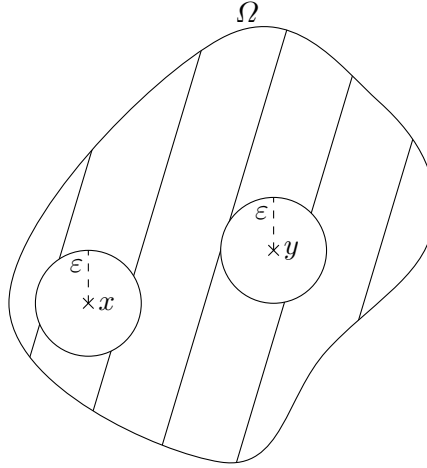


Figure 3.2: $B_{\varepsilon}(x) \subseteq \Omega$, $B_{\varepsilon}(y) \subseteq \Omega$, $B_{\varepsilon}(x) \cap B_{\varepsilon}(y) = \emptyset$

Choose $\varepsilon \in \mathbb{R}_{>0}$ such that $B_{\varepsilon}(x)$ and $B_{\varepsilon}(y)$ are disjoint subsets of Ω .

$$u(z) := G(z, x) \qquad v(z) := G(z, y)$$

It holds $u, v \in C^2(\Omega \setminus (B_{\varepsilon}(x) \cup B_{\varepsilon}(y)))$ and u, v are harmonic. Moreover for $z \in \partial\Omega$ holds:

$$u(z) = G(z, x) = 0 \qquad v(z) = G(z, y) = 0$$

Apply again the second Green's identity:

$$\begin{aligned} 0 &= \int_{\Omega \setminus (B_\varepsilon(x) \cup B_\varepsilon(y))} \underbrace{(v(\Delta u))}_{=0} - \underbrace{(\Delta v)u}_{=0} d\mu = \\ &= \int_{\partial\Omega \setminus (B_\varepsilon(x) \cup B_\varepsilon(y))} (v(\nabla_\nu u) - (\nabla_\nu v)u) d\mu_{\partial\Omega \setminus (B_\varepsilon(x) \cup B_\varepsilon(y))} \end{aligned}$$

Moreover, the boundary values on $\partial\Omega$ vanish. We conclude:

$$0 = \int_{\partial B_\varepsilon(x) \cup \partial B_\varepsilon(y)} (v(\nabla_\nu u) - (\nabla_\nu v)u) d\mu_{\partial B_\varepsilon(x) \cup \partial B_\varepsilon(y)}$$

Consider the integral over $\partial B_\varepsilon(x)$ first. Since $G = \Gamma + h$ and $h \in C^2$ is bounded, we know for all $z \in \partial B_\varepsilon(x)$:

$$u(z) = G(z, x) \sim \Gamma(z, x) \sim \begin{cases} \ln(\varepsilon) & \text{if } n = 2 \\ \varepsilon^{2-n} & \text{if } n > 2 \end{cases}$$

Since $\nabla_\nu v$ is also bounded due to $v \in C^2(B_\varepsilon(x))$, we get:

$$\int_{\partial B_\varepsilon(x)} (\nabla_\nu v)u d\mu_{\partial B_\varepsilon(x)} \xrightarrow{\varepsilon \searrow 0} 0$$

The other term gives:

$$\int_{\partial B_\varepsilon(x)} v(\nabla_\nu u) d\mu_{\partial B_\varepsilon(x)} \stackrel{\nabla_\nu u = \frac{\partial}{\partial \varepsilon} \Gamma(\varepsilon) + o_0(\varepsilon^0) = \tilde{c}\varepsilon^{1-n} + o_0(\varepsilon^0)}{=} cv(x)\varepsilon^{1-n} \underbrace{\int_{\partial B_\varepsilon(x)} 1 d\mu_{\partial B_\varepsilon(x)}}_{=n\omega_n\varepsilon^{n-1}} + o_0(\varepsilon) \xrightarrow{\varepsilon \searrow 0} cv(x)$$

Here the constant $c \neq 0$ does not vanish. Now follows:

$$\begin{aligned} &\int_{\partial B_\varepsilon(x)} ((\nabla_\nu v)u - v(\nabla_\nu u)) d\mu_{\partial B_\varepsilon(x)} \xrightarrow{\varepsilon \searrow 0} cv(x) \\ &\int_{\partial B_\varepsilon(y)} ((\nabla_\nu v)u - v(\nabla_\nu u)) d\mu_{\partial B_\varepsilon(y)} \xrightarrow{\varepsilon \searrow 0} cu(y) \end{aligned}$$

Adding these two integrals gives:

$$\begin{aligned} 0 &= c(v(x) - u(y)) \\ \Rightarrow G(x, y) &= v(x) = u(y) = G(y, x) \end{aligned}$$

□_{3.2.3}

Any solution u of the Dirichlet problem

$$\Delta u = f \quad u|_{\partial\Omega} = \varphi$$

has the representation:

$$u(y) = \int_{\partial\Omega} u(x) \nabla_\nu G(x, y) d\mu_{\partial\Omega}(x) + \int_{\Omega} G(x, y) f(x) d\mu(x)$$

We define:

$$u(y) := \int_{\partial\Omega} \varphi(x) \nabla_\nu G(x, y) d\mu_{\partial\Omega}(x) + \int_{\Omega} G(x, y) f(x) d\mu(x)$$

Since G is symmetric, $G(x, y) = G(y, x)$, the equation $\Delta_x G(x, y) = \delta^{(n)}(x - y)$ implies $\Delta_y G(x, y) = \delta^{(n)}(x - y)$ as well. As a consequence, a formal computation gives:

$$\Delta_y u(y) = \int_{\partial\Omega} \varphi(x) \nabla_\nu \delta^{(n)}(x - y) d\mu_{\partial\Omega}(x) + \int_{\Omega} \delta^{(n)}(x - y) f(x) d\mu(x)$$

For $y \in \Omega$ the first term vanishes, so we get:

$$\Delta_y u(y) = f(y)$$

This formal calculation can be made rigorous, once an explicit Green's function is given. Then one can also check, whether the boundary conditions are satisfied.

3.2.4 Example (Green's function for $B_R(0)$)

Now we want to construction the Green's function for $\Omega := B_R(0) \subseteq \mathbb{R}^n$. Consider an electric charge inside an earthed, electrically conducting sphere, that screens the electric field, so that it vanishes outside the sphere. The electric field inside the sphere can be calculated using the concept of a mirror charge, which ensures, that the electric field is perpendicular to the sphere.

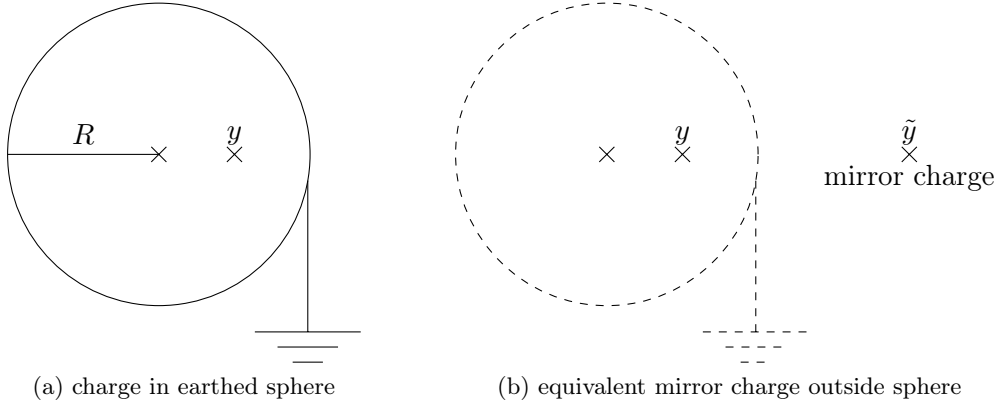


Figure 3.3: The mirror charge replicates the boundary conditions of the sphere.

The position of the mirror charge must be:

$$\begin{aligned} \tilde{y} &= \frac{R^2}{y^2} \cdot y \\ \Rightarrow \|y\| \cdot \|\tilde{y}\| &= R^2 \end{aligned}$$

Compare this with the inversion with respect to the unit circle (Spiegelung am Einheitskreis) in the complex plane:

$$z \mapsto \frac{1}{\bar{z}} = \frac{z}{|z|^2}$$

This motivates the ansatz:

$$G(x, y) = \begin{cases} \Gamma(\|x - y\|) - \Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) & \text{if } y \neq 0 \\ \Gamma(\|x\|) - \Gamma(R) & \text{if } y = 0 \end{cases}$$

Let us verify that G has all the required properties:

$$G(x, y) - \Gamma(x, y) = \begin{cases} -\Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) & \text{if } y \neq 0 \\ -\Gamma(R) & \text{if } y = 0 \end{cases}$$

For $x, y \in B_R(0)$ follows $\tilde{y} \notin B_R(0)$ and thus $x \neq \tilde{y}$. So $\Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right)$ is smooth if $y \neq 0$. To see smoothness in the case $y = 0$, we rewrite:

$$\begin{aligned} \Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) &= \Gamma\left(\sqrt{\frac{y^2}{R^2} (x^2 + \tilde{y}^2 - 2\langle x, \tilde{y} \rangle)}\right) = \\ &= \Gamma\left(\sqrt{\frac{y^2}{R^2} \left(x^2 + \frac{R^4}{y^2} - 2\frac{R^2}{y^2} \langle x, y \rangle\right)}\right) = \\ &= \Gamma\left(\sqrt{\frac{x^2 y^2}{R^2} + R^2 - 2\langle x, y \rangle}\right) \end{aligned} \quad (3.18)$$

The argument is smooth at $y = 0$ and it holds:

$$\lim_{y \rightarrow 0} \Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) = \Gamma(R)$$

So $G(x, y) - \Gamma(x, y)$ is in C^2 for $x, y \in \Omega$.

For $x \in \partial\Omega$, i.e. $\|x\| = R$, holds:

$$\begin{aligned} G(x, y) &= \Gamma(\|x - y\|) - \Gamma\left(\sqrt{\frac{\|x\|^2 \|y\|^2}{R^2} + R^2 - 2\langle x, y \rangle}\right) = \\ &\stackrel{\|x\|=R}{=} \Gamma(\|x - y\|) - \Gamma\left(\sqrt{\|y\|^2 + \|x\|^2 - 2\langle x, y \rangle}\right) = \\ &= \Gamma(\|x - y\|) - \Gamma\left(\sqrt{\|x - y\|^2}\right) = 0 \end{aligned}$$

Thus $G(x, y)$ satisfies the boundary condition $G(x, y) = 0$ for $x \in \partial\Omega$.

Now we show that $G(x, y) - \Gamma(x, y)$ is harmonic:

$$\begin{aligned} G(x, y) - \Gamma(x, y) &= -\Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) \\ \nabla \Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) &= \Gamma'\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) \frac{\|y\|}{R} \nabla \|x - \tilde{y}\| \\ \Delta \Gamma\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) &= \Gamma''\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) \left(\frac{\|y\|}{R}\right)^2 (\nabla \|x - \tilde{y}\|)^2 + \\ &\quad + \Gamma'\left(\frac{\|y\|}{R} \|x - \tilde{y}\|\right) \frac{\|y\|}{R} \Delta \|x - \tilde{y}\| \end{aligned}$$

We know:

$$\begin{aligned} 0 &= \Delta \Gamma = \frac{1}{r^{n-1}} \partial_r (r^{n-1} \partial_r \Gamma) \\ \Rightarrow \quad 0 &= \Gamma''(r) + \frac{n-1}{r} \Gamma'(r) \end{aligned}$$

With this follows:

$$\begin{aligned} \Delta \Gamma \left(\frac{\|y\|}{R} \|x - \tilde{y}\| \right) &= -(n-1) \frac{R}{\|y\| \|x - \tilde{y}\|} \Gamma' \left(\frac{\|y\|}{R} \|x - \tilde{y}\| \right) \left(\frac{\|y\|}{R} \right)^2 (\nabla \|x - \tilde{y}\|)^2 + \\ &\quad + \Gamma' \left(\frac{\|y\|}{R} \|x - \tilde{y}\| \right) \frac{\|y\|}{R} \Delta \|x - \tilde{y}\| = \\ &= \frac{\|y\|}{R} \Gamma' \left(\frac{\|y\|}{R} \|x - \tilde{y}\| \right) \left(\frac{-(n-1)}{\|x - \tilde{y}\|} (\nabla \|x - \tilde{y}\|)^2 + \Delta \|x - \tilde{y}\| \right) = \\ &= \frac{\|y\|}{R} \Gamma' \left(\frac{\|y\|}{R} \|x - \tilde{y}\| \right) \left(\frac{-(n-1)}{\|x - \tilde{y}\|} \left(\frac{x - \tilde{y}}{\|x - \tilde{y}\|} \right)^2 + \frac{n-1}{\|x - \tilde{y}\|} \right) = 0 \end{aligned}$$

Thus $G(x, y)$ is the desired Green's function. From (3.18) one sees explicitly:

$$G(x, y) = G(y, x)$$

We hope that the solution of the Dirichlet problem

$$\Delta u = f \quad u|_{\partial B_R(0)} = \varphi$$

is given by the Green's representation:

$$u(y) = \int_{B_R(0)} G(x, y) f(x) d\mu(x) + \int_{\partial B_R(0)} \nabla_\nu G(x, y) \varphi(x) d\mu_{\partial B_R(0)}(x)$$

Computing $\nabla_\nu G(x, y)$ for $x \in \partial B_R(0)$ gives:

$$\nabla_\nu G(x, y) = \frac{R^2 - y^2}{n\omega_n R} \cdot \frac{1}{\|x - y\|^n}$$

3.2.5 Theorem (Poisson representation)

The function

$$u(y) := \begin{cases} \frac{R^2 - y^2}{n\omega_n R} \int_{\partial B_R(0)} \frac{\varphi(x)}{\|x - y\|^n} d\mu_{\partial B_R(0)} & \text{if } y \in B_R(0) \\ \varphi(y) & \text{if } y \in \partial B_R(0) \end{cases} \quad (3.19)$$

with $\varphi \in C^0(\partial B_R(0))$ has the following properties:

- $u \in C^0(\overline{B_R(0)})$
- $u \in C^2(B_R(0))$
- u is harmonic in $B_R(0)$.

Proof

This can be shown using Green's representation, computing the boundary values and justifying that the y -derivative may be taken inside the integral. $\square_{3.2.5}$

3.3 The Mean Value Theorem and the Maximum Principle for Harmonic Functions

3.3.1 Theorem (Mean Value Formulas)

A continuous function $u : \Omega \rightarrow \mathbb{R}$ with $\Omega \subseteq \mathbb{R}^n$ is harmonic, i.e. $u \in C^2(\Omega)$ and $\Delta u = 0$, if and only if for all $x_0 \in \Omega$ and all balls $B_r(x_0) \subseteq \Omega$ one of the following mean value formulas holds:

$$u(x_0) = \frac{1}{n\omega_n r^{n-1}} \int_{\partial B_r(x_0)} u(x) d\mu_{\partial B_r}(x) \quad (\text{spherical mean}) \quad (3.20)$$

$$u(x_0) = \frac{1}{\omega_n r^n} \int_{B_r(x_0)} u(x) d\mu(x) \quad (\text{mean over ball}) \quad (3.21)$$

In this case, the other formula holds as well.

Proof

“ \Rightarrow ”: Let $u \in C^2(\Omega)$ be harmonic and $B_r(x_0) \subseteq \Omega$. Then $B_r(y) \subseteq \Omega$ holds also for all y in a neighborhood of x_0 . The function

$$H(x, y) = \Gamma(x, y) - \Gamma(r)$$

coincides up to a constant with the fundamental solution and it holds $H(x, y) = 0$ for $x \in \partial B_r(y)$. Using Green's representation gives:

$$\begin{aligned} u(y) &= \int_{B_r(y)} \underbrace{(\Delta u)(x)}_{=0} \cdot H(x, y) d\mu(x) + \\ &\quad + \int_{\partial B_r(y)} \left(u(x) \nabla_\nu H(x, y) - \underbrace{H(x, y)}_{=0} \nabla_\nu u(x) \right) d\mu_{\partial B_r(y)}(x) = \\ &= \int_{\partial B_r(y)} u(x) \nabla_\nu H(x, y) d\mu_{\partial B_r(y)}(x) \end{aligned}$$

$$\nabla_\nu H(x, y) = \nabla_{\nu, x} \Gamma(\|x - y\|) \stackrel{r:=\|x-y\|}{=} \frac{\partial}{\partial r} \Gamma(r) = \frac{1}{n\omega_n r^{n-1}}$$

For $y := x_0$ follows the spherical mean formula:

$$u(x_0) = \frac{1}{n\omega_n r^{n-1}} \int_{\partial B_r(x_0)} u(x) d\mu_{\partial B_r}(x)$$

Now follows:

$$\begin{aligned} \int_{B_r(y)} u(x) d\mu(x) &\stackrel{\text{Fubini}}{=} \int_0^r d\rho \int_{\partial B_\rho(y)} u(x) d\mu_{\partial B_\rho}(x) = \int_0^r u(y) n\omega_n \rho^{n-1} d\rho = \\ &= u(y) n\omega_n \cdot \frac{1}{n} r^n = \omega_n r^n u(y) \end{aligned} \quad (3.22)$$

“ \Leftarrow ”: Let $u \in C^0(\Omega)$ be continuous. If the formula for the spherical mean holds, the computation (3.22) gives the formula for means over balls. Let us show that the formula for means over balls implies the formula for spherical means. So assume for fixed y and all $r < r_0$:

$$u(y) = \frac{1}{\omega_n r^n} \int_{B_r(y)} u(x) d\mu(x)$$

Thus using Fubini's theorem follows:

$$r^n u(y) = \frac{1}{\omega_n} \int_0^r \overbrace{\rho \int_{\partial B_\rho(y)} u(x) d\mu_{\partial B_\rho(y)}(x)}^{C^1 \text{ in } r} \underbrace{d\rho}_{\text{continuous in } \rho}$$

Differentiation on both sides with respect to r gives the formula for spherical means:

$$nr^{n-1}u(y) = \frac{1}{\omega_n} \int_{\partial B_r(y)} u(x) d\mu_{\partial B_r(y)}(x)$$

Next we show, that u is smooth in Ω . The idea is to mollify by convolution. Choose:

$$\varrho(t) := \begin{cases} c_n e^{\frac{1}{t^2-1}} & \text{if } -1 < t < 1 \\ 0 & \text{otherwise} \end{cases} \in C_0^\infty(\mathbb{R})$$

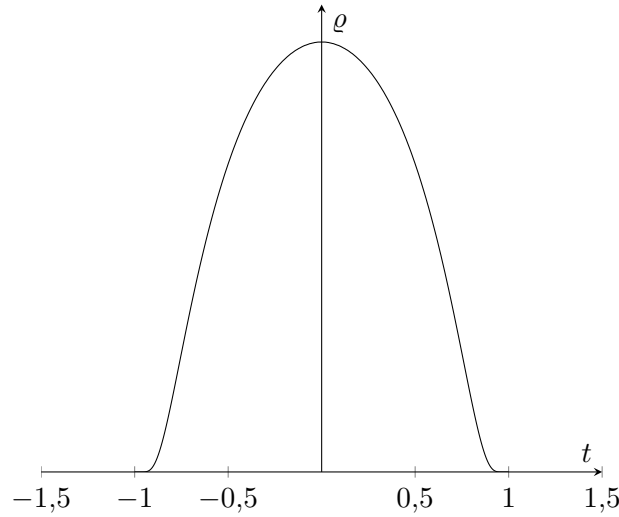


Figure 3.4: $\varrho(t)$ is smooth and has compact support $[-1, 1]$.

Then holds $\varrho(\|x\|) \in C_0^\infty(\mathbb{R}^n)$. Choose c_n such that holds:

$$1 = \int_{\mathbb{R}^n} \varrho(\|x\|) d^n x$$

Now for $\varepsilon \in \mathbb{R}_{>0}$ set:

$$\varrho_\varepsilon(x, y) := \frac{1}{\varepsilon^n} \varrho\left(\frac{\|x - y\|}{\varepsilon}\right)$$

Then still holds:

$$1 = \int_{\mathbb{R}^n} \varrho_\varepsilon(x, y) d^n x$$

Also we know $\varrho_\varepsilon(x, \cdot) \in C_0^\infty(B_{2\varepsilon}(x))$. Choose ε so small that $B_{2\varepsilon}(y) \subseteq \Omega$ and use ϱ_ε as our convolution kernel to define:

$$u_\varepsilon(y) := \int_{\Omega} \varrho_\varepsilon(x, y) u(x) d\mu(x)$$

Now u_ε is a smooth function, because $\varrho_\varepsilon(x, \cdot)$ is smooth and any derivative with respect to y can be exchanged with the integral, since the integration volume is compact.

$$u_\varepsilon(y) = \frac{1}{\varepsilon^n} \int_{\Omega} \varrho\left(\frac{\|x - y\|}{\varepsilon}\right) u(x) d\mu(x)$$

Choose polar coordinates around y and use Fubini to get:

$$\begin{aligned} u_\varepsilon(y) &= \frac{1}{\varepsilon^n} \int_0^{2\varepsilon} dr \int_{\partial B_r(y)} \varrho\left(\frac{r}{\varepsilon}\right) u(x) d\mu_{\partial B_r(y)}(x) = \\ &\stackrel{\text{spherical mean}}{=} \frac{1}{\varepsilon^n} \int_0^{2\varepsilon} dr \varrho\left(\frac{r}{\varepsilon}\right) u(y) n\omega_n r^{n-1} = \\ &= u(y) \frac{1}{\varepsilon^n} \int_0^{2\varepsilon} dr \int_{\partial B_r(y)} 1 \cdot \varrho\left(\frac{r}{\varepsilon}\right) d\mu_{\partial B_r(y)}(x) = \\ &= u(y) \int_{\mathbb{R}^n} \varrho_\varepsilon(x, y) d^n x = u(y) \end{aligned}$$

Thus u is smooth.

Compute Δu using the theorem of Gauss:

$$\int_{B_r(y)} (\Delta u(x)) d\mu(x) = \int_{\partial B_r(y)} \nabla_\nu u d\mu_{\partial B_r(y)}$$

With

$$\omega := \frac{x - y}{\|x - y\|} \in S^{n-1}$$

we get:

$$\begin{aligned} \int_{B_r(y)} (\Delta u(x)) d\mu(x) &= r^{n-1} \int_{\partial B_1(0)} \frac{\partial}{\partial r} u(y + r\omega) d\mu_{\partial B_1(0)}(\omega) = \\ &= r^{n-1} \frac{\partial}{\partial r} \left(\frac{1}{r^{n-1}} \int_{\partial B_r(y)} u(x) d\mu_{\partial B_r(y)}(x) \right) = \\ &\stackrel{\text{mean value property}}{=} r^{n-1} \frac{\partial}{\partial r} \left(\frac{1}{r^{n-1}} u(y) n\omega_n r^{n-1} \right) = 0 \end{aligned}$$

In the limit $r \rightarrow 0$ follows $\Delta u(y) = 0$ since u is smooth. Thus u is harmonic. □_{3.3.1}

3.3.2 Theorem (Strong Maximum Principle)

Let $u \in C^2(\Omega)$ be a harmonic function on an open and connected subset $\Omega \subseteq \mathbb{R}^n$. If a point $x_0 \in \Omega$ with

$$u(x_0) = \sup_{x \in \Omega} (u)$$

or

$$u(x_0) = \inf_{x \in \Omega} (u)$$

exists, then u is constant.

Proof

Since $-u$ is also harmonic, it suffices to consider the case $u(x_0) = \sup_{\Omega} (u)$, so assume

$$u(x_0) = \sup_{\Omega} (u) =: M$$

and define:

$$\Omega_M := \{x \in \Omega \mid u(x) = M\}$$

Clearly, Ω_M is not empty, because $x_0 \in \Omega_M$. By continuity, Ω_M is closed in Ω with respect to the relative topology. Thus it suffices to show that Ω_M is open. Consider $y \in \Omega_M$, i.e. $u(y) = M$, and choose $r \in \mathbb{R}_{>0}$ with $B_r(y) \subseteq \Omega$.

$$0 = u(y) - M \stackrel{\text{mean value property}}{=} \frac{1}{\omega_n r^n} \int_{B_r(y)} \underbrace{(u(x) - M)}_{\leq 0} d\mu(x)$$

Since $u(x) \leq M$, the integrand is non-positive and continuous. So $u(x) = M$ for all $x \in B_r(y)$ and thus follows $B_r(y) \subseteq \Omega_M$. This means that Ω_M is open. Since it is also non-empty, closed and connected, the only possibility is $\Omega_M = \Omega$. $\square_{3.3.2}$

3.3.3 Theorem (Weak Maximum Principle)

Let $\Omega \subseteq \mathbb{R}^n$ be open and bounded and $u \in C^0(\overline{\Omega})$ be harmonic. Then holds for all $x \in \Omega$:

$$\min_{y \in \partial\Omega} (u(y)) \leq u(x) \leq \max_{y \in \partial\Omega} (u(y))$$

Proof

If the statement were false, there would be an interior maximum or minimum at point $x_0 \in \Omega$, since $\overline{\Omega}$ is compact, which ensures that the maximum and minimum of u are attained in $\overline{\Omega}$. Let $\tilde{\Omega} \subseteq \Omega$ be the connected component of Ω which contains x_0 .

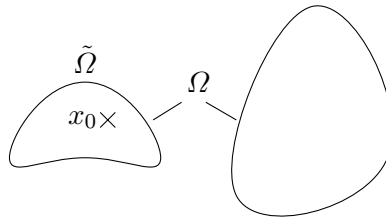


Figure 3.5: $\tilde{\Omega}$ is the connected component of Ω that contains x_0 .

The strong maximum principle implies $u = u(x_0)$ on $\tilde{\Omega}$. Thus follows:

$$\begin{aligned} \min_{\partial\tilde{\Omega}} (u) &= u(x_0) = \max_{\partial\tilde{\Omega}} (u) \\ \Rightarrow \min_{\partial\Omega} (u) &\leq \min_{\partial\tilde{\Omega}} (u) = u(x_0) = \max_{\partial\tilde{\Omega}} (u) \leq \max_{\partial\Omega} (u) \end{aligned}$$

Therefore x_0 cannot be an interior maximum or minimum in contradiction to our assumption. $\square_{3.3.3}$

3.3.4 Corollary (Uniqueness of Solutions of the Poisson Equation)

Let $u, v \in C^0(\overline{\Omega}) \cap C^2(\Omega)$ be solutions of the Poisson equation:

$$\Delta u = f = \Delta v$$

If moreover holds $u \leq v$ on $\partial\Omega$, then follows $u \leq v$ in Ω .

In particular, $u = v$ on $\partial\Omega$ implies $u = v$ in Ω .

Proof

The function $h := u - v$ is harmonic

$$\Delta h = \Delta u - \Delta v = f - f = 0$$

and on $\partial\Omega$ holds $h \leq 0$. The weak maximum principle gives $h \leq 0$ in Ω .

□_{3.3.4}

3.3.5 Corollary (Liouville's Theorem)

If $u : \mathbb{R}^n \rightarrow \mathbb{R}$ is harmonic and bounded, then u is constant.

Proof

Choose $y_1, y_2 \in \mathbb{R}^n$. Then for any $r \in \mathbb{R}_{>0}$ holds for $i \in \{1, 2\}$.

$$u(y_i) = \frac{1}{\omega_n r^n} \int_{B_r(y_i)} u(x) \, d\mu(x)$$

The difference is:

$$u(y_1) - u(y_2) = \frac{1}{\omega_n r^n} \left(\int_{B_r(y_1)} u(x) \, d\mu(x) - \int_{B_r(y_2)} u(x) \, d\mu(x) \right)$$

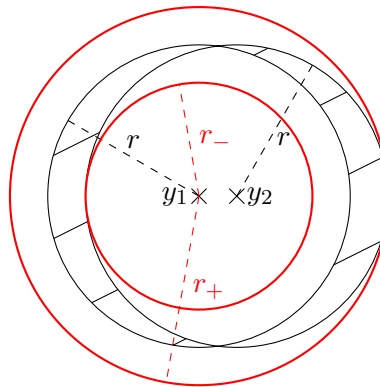


Figure 3.6: $r_- := r - \|y_1 - y_2\|$, $r_+ := r + \|y_1 - y_2\|$

So we get:

$$u(y_1) - u(y_2) \leq \frac{1}{\omega_n r^n} \underbrace{\sup_{\mathbb{R}^n} (|u|)}_{< \infty} \mu(B_r(y_1) \setminus B_r(y_2) \cup B_r(y_2) \setminus B_r(y_1)) \leq$$

$$\begin{aligned}
 &\leq \frac{1}{\omega_n r^n} \sup_{\mathbb{R}^n} (|u|) \left(\mu(B_{r+\|y_2-y_1\|}(y_1)) - \mu(B_{r-\|y_2-y_1\|}(y_1)) \right) \leq \\
 &\leq \frac{1}{\omega_n r^n} \sup_{\mathbb{R}^n} (|u|) (\omega_n (r + \|y_2 - y_1\|^n) - \omega_n (r - \|y_2 - y_1\|^n)) = \\
 &= \sup_{\mathbb{R}^n} (|u|) \frac{1}{r^n} (2n \|y_2 - y_1\| r^{n-1} + \dots) \xrightarrow{r \rightarrow \infty} 0
 \end{aligned}$$

This gives $u(y_1) = u(y_2)$ and thus u is constant. $\square_{3.3.5}$

3.4 The Harnack Inequality for Harmonic Functions

3.4.1 Theorem (Harnack inequality)

Let $\Omega \subseteq \mathbb{R}^n$ be open and $u : \Omega \rightarrow \mathbb{R}$ harmonic and non-negative. Then for every compact, (for simplicity) connected subset $\Omega' \subseteq \Omega$ there is a constant $c = c(\Omega, \Omega') \in \mathbb{R}$, which is independent of u , such that holds:

$$\sup_{\Omega'} (u) \leq c \cdot \inf_{\Omega'} (u)$$

Proof

We begin with the case $\Omega' = B_r(x_0)$ and $B_{4r}(x_0) \subseteq \overline{\Omega}$. Choose $y_1, y_2 \in \Omega'$.

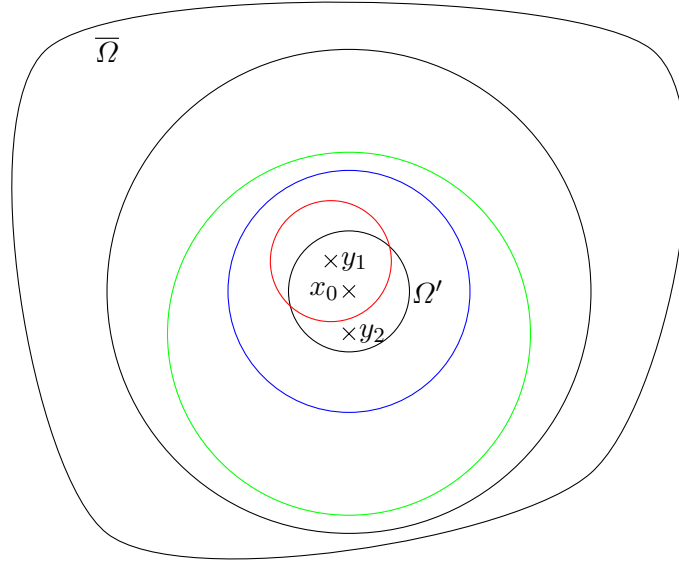


Figure 3.7: $B_r(y_1) \subseteq B_{2r}(x_0) \subseteq B_{3r}(y_2) \subseteq B_{4r}(x_0) \subseteq \overline{\Omega}$

It holds:

$$\begin{aligned}
 u(y_1) &= \frac{1}{\omega_n r^n} \int_{B_r(y_1)} u(x) d\mu(x) \stackrel{u \geq 0}{\leq} \frac{1}{\omega_n r^n} \int_{B_{2r}(x_0)} u(x) d\mu(x) \leq \\
 &\leq \frac{1}{\omega_n r^n} \int_{B_{3r}(y_2)} u(x) d\mu(x) \leq \frac{3^n}{\omega_n (3r)^n} \int_{B_{3r}(y_2)} u(x) d\mu(x) = 3^n u(y_2)
 \end{aligned}$$

Taking the supremum over y_1 and the infimum over y_2 we get:

$$\sup_{B_r(x_0)} (u) \leq 3^n \inf_{B_r(x_0)} (u)$$

Now consider a general compact subset $\Omega' \subseteq \Omega$. If Ω' is empty, we have nothing to show, so consider a non-empty Ω' . We want to find a $r \in \mathbb{R}_{>0}$ such that for every $x \in \Omega'$ holds $B_{4r}(x) \subseteq \Omega$. For $\Omega = \mathbb{R}^n$ we can choose any r . Otherwise $\partial\Omega \subseteq \mathbb{R}^n$ is non-empty and closed. Since Ω' is also non-empty and even compact, the distance $\text{dist}(\Omega', \partial\Omega)$ is attained. This must be larger than zero, because from $\partial\Omega \cap \Omega = \emptyset$ and $\Omega' \subseteq \Omega$ follows $\partial\Omega \cap \Omega' = \emptyset$. So we can choose any $r \in \mathbb{R}_{>0}$ with:

$$r < \frac{1}{4} \text{dist}(\Omega', \partial\Omega)$$

We cover Ω' by a finite number $\Omega_1, \dots, \Omega_N$ of balls of radius r . On each of the sets Ω_l holds:

$$\sup(u) \leq 3^n \inf(u)$$

One can proceed inductively as follows: Without loss of generality holds $\sup_{\Omega'}(u) \leq \sup_{\Omega_1}(u)$, otherwise change the numbering of the Ω_i . Since Ω' is connected, one can choose a finite number $\Omega_2, \dots, \Omega_k$ from the balls of the covering such that for $i \in \{1, \dots, k-1\}$ holds

$$\Omega_i \cap \Omega_{i+1} \neq \emptyset$$

and $\inf_{\Omega_k}(u) \leq \inf_{\Omega'}(u)$. If $\inf_{\Omega_1}(u) \leq \inf_{\Omega'}(u)$, we can choose $k = 1$. It holds:

$$\sup_{\Omega_i}(u) \leq 3^n \inf_{\Omega_i}(u) \stackrel{z_i \in \Omega_i \cap \Omega_{i+1}}{\leq} 3^n u(z_i) \leq 3^n \sup_{\Omega_{i+1}}(u)$$

Inductively one gets after $k-1$ steps:

$$\sup_{\Omega'}(u) \leq \sup_{\Omega_1}(u) \leq 3^{(k-1)n} \sup_{\Omega_k}(u) \leq 3^{kn} \inf_{\Omega_k}(u) \leq 3^{kn} \inf_{\Omega'}(u)$$

In the worst case of u , we have $k = N$ and therefore for any u holds:

$$\sup_{\Omega'}(u) \leq 3^{Nn} \inf_{\Omega'}(u)$$

Thus the theorem is proven. □_{3.4.1}

3.4.2 Corollary (Harnack's Convergence Theorem)

Let $u_n : \Omega \rightarrow \mathbb{R}^n$ be a monotonically increasing sequence of harmonic functions. Assume that there is a $y \in \Omega$, where $u_n(y)$ is a bounded sequence.

Then the functions u_n converge locally uniformly, i.e. uniformly in a small neighborhood of y , to a harmonic function u .

Proof

Since the sequence $u_n(y)$ is bounded and monotonically increasing, it is convergent, i.e. for all $\varepsilon \in \mathbb{R}_{>0}$ there exists a $N \in \mathbb{N}$ such that for all $k > m > N$ holds:

$$u_k(y) - u_m(y) < \varepsilon$$

The function $u_k - u_m$ is non-negative and harmonic. Now choose a bounded and connected neighborhood Ω' of y with $\overline{\Omega'} \subseteq \Omega$. According to the Harnack inequality holds:

$$\sup_{\overline{\Omega'}} (u_k - u_m) \leq c(\Omega, \overline{\Omega'}) \cdot \inf_{\overline{\Omega'}} (u_k - u_m) \leq c(u_k(y) - u_m(y)) \xrightarrow{k, m \rightarrow \infty} 0$$

Hence u_k converges *uniformly* in Ω' to a function $u \in C^0(\Omega')$. The function u is harmonic, because for all $x_0 \in \Omega'$ and $r \in \mathbb{R}_{>0}$ such that $B_r(x_0) \subseteq \Omega'$ holds:

$$u_k(x_0) = \frac{1}{\omega_n r^n} \int_{B_r(x_0)} u_k(x) \, d\mu(x)$$

From $u_k(x_0) \rightarrow u(x_0)$ and $u_k(x) \rightrightarrows u(x)$ follows:

$$u(x_0) = \frac{1}{\omega_n r^n} \int_{B_r(x_0)} u(x) \, d\mu(x)$$

So u has the mean-value property and is thus harmonic. □_{3.4.2}

4 The Perron Method of Sub- and Supersolution

Consider the Dirichlet problem on a bounded $\Omega \subseteq \mathbb{R}^n$:

$$\Delta u = 0 \qquad u|_{\partial\Omega} = \varphi$$

The Perron method, named after Oskar Perron (1880-1972), yields the existence of a solution if $\partial\Omega$ is sufficiently smooth. This is a classical method, in the sense that it works always with continuous functions and no weak solutions, but only classical solutions. The idea is to find the solution as a supremum of subsolutions:

$$u(x) := \sup_{v \text{ subsolution}} (v(x))$$

This works for any domain and gives a harmonic function, but $\partial\Omega$ has to be sufficiently smooth, for the boundary condition to be satisfied.

Appendix

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Andreas Völklein