

# Calculus in Infinite Dimensions

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# Finite vs Infinite Dimensional Space

**Finite Dimensional**

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$$\mathbb{R}, \mathbb{R}^2, \mathbb{R}^3, \dots$$

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- Lebesgue & Sobolev spaces

# Finite Dimensional Calculus

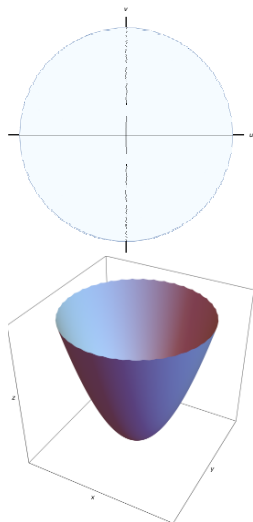
Consider a **function**  $f : B \rightarrow \mathbb{R}^3$ , with  $B$  the unit ball in 2D, given by

$$f(u, v) = \begin{pmatrix} u + v \\ u - v \\ u^2 + v^2 \end{pmatrix}$$

Then the **gradient** is given by

$$\nabla f(u, v) = \begin{pmatrix} +1 & +1 \\ +1 & -1 \\ 2u & 2v \end{pmatrix}$$

We can use this to investigate stationary points, minima, maxima etc.



# Infinite Dimensional Calculus

Now consider a **functional**  $f : \mathbb{R}^{\mathbb{N}} \rightarrow \mathbb{R}$ , given by

$$f(x) = x_1$$

Then the **gradient** is a sequence  $\nabla f(x)$  with entries given by

$$(\nabla f(x))_i = \delta_{1,i} = \begin{cases} 1 & i = 1 \\ 0 & i > 1 \end{cases}$$

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We shall mainly focus on functionals defined on spaces of functions.

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Taylor expand  $f$ :

$$f(x_0 + \epsilon v) - f(x_0) = \epsilon \nabla f(x_0)[v] + \epsilon^2 \nabla^2 f(x_0)[v, v] + o(\epsilon^2) \geq 0$$

for any vector  $v$ .

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for any vector  $v$ . Then

$$\nabla f(x_0) = 0$$

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is sufficient. However, only  $\nabla^2 f(x_0) \geq 0$  is necessary.



# Local Minima in Infinite Dimensions

Consider a functional

$$F[x] = \int_0^1 f(t, x(t), x'(t)) dt \quad x \in C^2((0, 1), \mathbb{R}^d)$$

For now, assume  $f \in C^1$ .

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We want this to be positive for all  $\varphi \in C_c^\infty(0, 1)$  and small  $\epsilon$ .

# Local Minima in Infinite Dimensions

If we want

$$\int_0^1 (f_x(t, x_0, x'_0) - f_{x'}(t, x_0, x'_0)') \varphi \, dt = 0 \quad \forall \varphi \in C_c^\infty(0, 1)$$

we must have

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This is known as the **Euler-Lagrange** equation. It is a second order differential equation to be solved for  $x_0(t)$ , provided we specify initial/boundary conditions.

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$$\begin{aligned} \frac{\delta L}{\delta \mathbf{r}} = 0 - \frac{d}{dt} \frac{\mathbf{r}'(t)}{|\mathbf{r}'(t)|_2} &= - \frac{\mathbf{r}''(t) |\mathbf{r}'(t)|_2 - \mathbf{r}'(t) \frac{\mathbf{r}'(t)}{|\mathbf{r}'(t)|_2} \cdot \mathbf{r}''(t)}{|\mathbf{r}'(t)|_2^2} \\ &= \left( \mathbf{r}'(t) \otimes \mathbf{r}'(t) - |\mathbf{r}'(t)|_2^2 I \right) \frac{\mathbf{r}''(t)}{|\mathbf{r}'(t)|_2^3} \end{aligned}$$

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This gives

$$\mathbf{r}(t) = (\mathbf{b} - \mathbf{a})t + \mathbf{a}$$

# Projectiles

The action of a particle of mass  $m$  travelling along a path  $\mathbf{r}(t)$ , under the influence of gravity, is

$$S[\mathbf{r}] = \int_0^1 \frac{m}{2} |\mathbf{r}'(t)|_2^2 + m\mathbf{g} \cdot \mathbf{r}(t) dt \quad \mathbf{g} = (0, -g)$$

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If we launch a particle from a cannon at the origin and it lands at  $(1, 0)$ . The trajectory must solve

$$\mathbf{r}''(t) = \mathbf{g} \quad \mathbf{r}(0) = (0, 0), \quad \mathbf{r}(1) = (1, 0)$$

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The time taken for the ball to reach its destination is

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The solution to this is an example of a **cycloid** and can be written as

$$x(\omega\theta) = \alpha(\theta - \sin(\theta))$$

$$y(\omega\theta) = \alpha(1 - \cos(\theta))$$

for an appropriate choice of  $\alpha > 0$  and  $\omega$ .



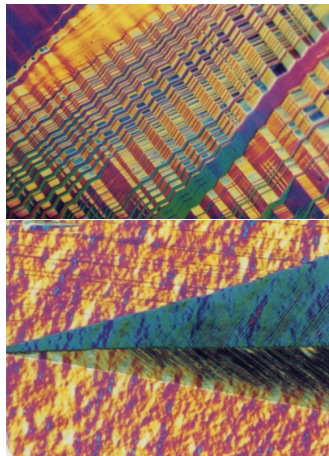
# Elasticity & SME

Deformations of materials  $\Omega$  are often investigated by studying functionals:

$$E[u] = \int_{\Omega} \varphi(\nabla u, \theta) dV$$

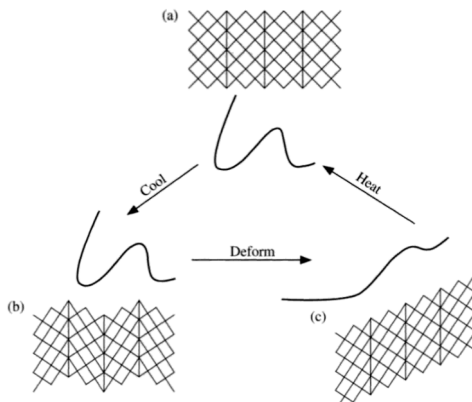
where

- $E$  is the total energy of the material.
- $\varphi$  is the stored energy density.
- $u$  is an admissible deformation.
- $\theta$  is the temperature.



# Shape Memory Effect

Changing the temperature changes the functional and thus changes the minimisers (what we observe).



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# Dirichlet Energy Functional

The Dirichlet energy functional  $\mathbb{D}(u)$  can be used to describe the strain energy density under a deformation  $u$ .

$$\mathbb{D}(u) = \int_B |\nabla u|_{\mathbb{F}}^2 \, dx \quad u \in u_0 + W_0^{1,2}(B; \mathbb{R}^2)$$

with  $B$  the unit ball and  $u_0$  is the **double-covering map** [1].

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where, in particular, we have

$$E(u) = E_f(u) := \int_B |\nabla u|_F^2 + f(x) \det \nabla u \, dx$$

with  $f(u) = 3 \log |x|_2$  (often called a 'pressure' function) [1].

# A 1D Family of Polyconvex Functions

The functional

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We observe that:

- $E_\lambda(0) = 0$ .
- $E_\lambda$  is either unbounded below or bounded below by zero.

Hence  $E_\lambda$  has a global minimum iff  $E_\lambda \geq 0$ .

# Pointwise Hadamard

We can use of the following theorem

## Hadamard's Inequality [3]

For any matrix  $A \in \mathbb{R}^{2 \times 2}$ , we have  $|A|_F^2 \geq 2 |\det A|$ .

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to show

$$\begin{aligned} E_\lambda(u) &\geq \int_B |\nabla u(x)|_F^2 - \frac{\lambda}{2} |x|_2 |\nabla u(x)|_F^2 \, dx \\ &= \int_B \left(1 - \frac{\lambda}{2} |x|_2\right) |\nabla u(x)|_F^2 \, dx \geq 0 \end{aligned}$$

for  $\lambda \leq 2$ .

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for  $\lambda \leq 2$ . Can we do better?

## Fourier Series &amp; Poincaré Inequality

Consider a Fourier expansion of  $u$  as

$$u = \sum_{j \geq 0} u^{(j)} = \frac{1}{2} \mathbf{A}_0(r) + \sum_{j > 0} \mathbf{A}_j(r) \cos(j\theta) + \mathbf{B}_j(r) \sin(j\theta)$$

with  $\mathbf{A}_j(1) = \mathbf{B}_j(1) = 0$  for each  $j \geq 0$  [1].

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Then we make use of a Poincaré inequality for the Fourier modes:

## Poincaré Inequality for Fourier Modes

Let  $j \geq 1$ . Then

$$\int_B |u^{(j)}|_2^2 dx \leq \frac{1}{j_0^2} \int_B |u_{,r}^{(j)}|_2^2 dx$$

where  $j_0$  is the first zero of a Bessel function.

# Applying the Inequality

Applying this inequality, we find

$$\begin{aligned}
 E_\lambda(u) &= \sum_{j \geq 0} \int_B \left| u_{,r}^{(j)} \right|_2^2 + \left| u_{,\tau}^{(j)} \right|_2^2 + \frac{\lambda}{2} \left\langle u^{(j)}, Ju_{,\tau}^{(j)} \right\rangle_2 dx \\
 &\geq \sum_{j \geq 0} \int_B j_0^2 \left| u^{(j)} \right|_2^2 + \left| u_{,\tau}^{(j)} \right|_2^2 - \frac{\lambda}{2} \left| u^{(j)} \right|_2 \left| u_{,\tau}^{(j)} \right|_2 dx
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 &\geq \sum_{j \geq 0} \int_B j_0^2 \left| u^{(j)} \right|_2^2 + \left| u_{,\tau}^{(j)} \right|_2^2 - \frac{\lambda}{2} \left| u^{(j)} \right|_2 \left| u_{,\tau}^{(j)} \right|_2 dx \\
 &= \sum_{j \geq 0} \int_B \mathbf{u}_j^\top M_\lambda \mathbf{u}_j dx \quad \mathbf{u}_j = \begin{pmatrix} \left| u^{(j)} \right|_2^2 \\ \left| u_{,\tau}^{(j)} \right|_2^2 \end{pmatrix} \quad M_\lambda = \begin{pmatrix} j_0^2 & -\frac{\lambda}{4} \\ -\frac{\lambda}{4} & 1 \end{pmatrix}
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# Applying the Inequality

Applying this inequality, we find

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 E_\lambda(u) &= \sum_{j \geq 0} \int_B \left| u_{,r}^{(j)} \right|_2^2 + \left| u_{,\tau}^{(j)} \right|_2^2 + \frac{\lambda}{2} \left\langle u^{(j)}, Ju_{,\tau}^{(j)} \right\rangle_2 dx \\
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Then we have that  $\lambda \leq 4j_0 \Rightarrow M_\lambda \geq 0 \Rightarrow E_\lambda \geq 0$ .

Here  $4j_0 \approx 9.619$  is a significant improvement on the previous bound of  $\lambda \leq 2$ .

# The Approach

We shall first construct a solution  $u^* \in W_0^{1,2}$  to

$$\nabla u^*(x) \in O(2) \quad \text{a.e. } x \in Q := [-2, +2]^2$$

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A necessary condition can then be derived:

$$\lambda \leq \frac{2(\pi - |K|)}{|\mathcal{I}_0|}$$

# Partial Differential Inclusions

The PDI

$$\nabla u \in O(2)$$

is an example of the **problem of potential wells**:

The Problem of Potential Wells [2, 4, 5]

$$\nabla u \in \bigcup_i \text{SO}(d)A_i \quad A_i \in \mathbb{R}^{d \times d}$$



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The following is known for this PDI:

Theorem (Dacorogna and Marcellini) [3]

There exists a dense set of solutions in  $W^{1,\infty}(Q; \mathbb{R}^2)$ .

Theorem (Liouville) [4]

All solutions are piecewise affine.

## Constructed Solution

A solution has been constructed for the 3D equivalent of this PDI [2].

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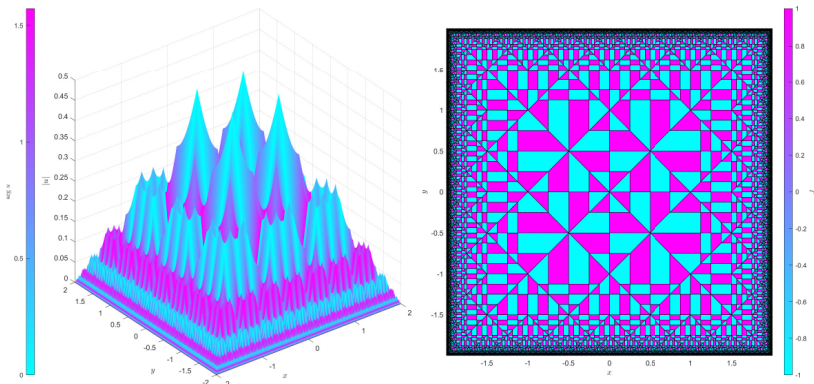


Figure: Constructed solution  $u^*$  and its Jacobian  $\det \nabla u^*$

# Tiling the Solution

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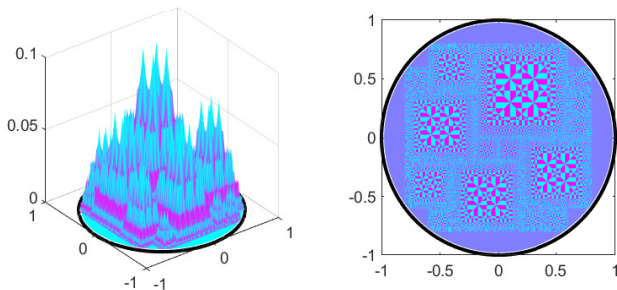


Figure: Tiled Solution with 15 Squares

# Numerical Results

We will make use of a simple tiling of 2 squares with

$$c_1 = \begin{pmatrix} +0.1 \\ +0.3 \end{pmatrix} \quad w_1 = 1.0 \quad c_2 = \begin{pmatrix} -0.3 \\ -0.5 \end{pmatrix} \quad w_2 = 0.6$$

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This leads to a bound of  $\lambda \leq 148.446$ .

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With further optimisations, we could try to refine these bounds.

We could try to find  $\lambda_{\text{crit}}$  such that  $E_\lambda \geq 0$  iff  $\lambda \leq \lambda_{\text{crit}}$ .

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