

# Anisotropic Poisson Problem

An anisotropic Poisson problem in a two-dimensional domain  $\Omega$  is given by the strong form

$$\begin{aligned} -\nabla \cdot (A \nabla m) + m &= f && \text{in } \Omega, \\ A \nabla m \cdot \mathbf{n} &= 0 && \text{on } \partial\Omega, \end{aligned}$$

where the conductivity tensor  $A(\mathbf{x}) \in \mathbb{R}^{2 \times 2}$  is assumed to be symmetric and positive definite for all  $\mathbf{x}$ ,  $f(\mathbf{x})$  is a given distributed source, and  $\mathbf{n}$  is the unit outward normal vector to  $\partial\Omega$ .

## Solve BVP problem

Choose  $\Omega$  to be a unit square  $[-1, 1]^2$  and take the source terms to be

$$f = \exp(-100(x^2 + y^2)).$$

Use conductivity tensors  $A(x)$  given by

$$A_1 = \begin{pmatrix} 10 & 0 \\ 0 & 10 \end{pmatrix} \text{ and } A_2 = \begin{pmatrix} 1 & -0.5 \\ -0.5 & 10 \end{pmatrix}$$

## Part b

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In [1]: import matplotlib.pyplot as plt
        %matplotlib inline

import dolfin as dl

import math
import numpy as np
import logging

logging.getLogger('FFC').setLevel(logging.WARNING)
logging.getLogger('UFL').setLevel(logging.WARNING)
dl.set_log_active(False)

mesh = dl.RectangleMesh(dl.Point(-1,-1), dl.Point(1,1), 128, 128)

V = dl.FunctionSpace(mesh, "Lagrange", 2)
print ("Number of unknowns = ", V.dim() )

f = dl.Expression("exp(-100*(x[0]*x[0] + x[1]*x[1]))", degree = 5)

A1 = dl.Constant(((10., 0.), (0., 10.0)))
A2 = dl.Constant(((1., -.5), (-0.5, 10.0)))

m_hat = dl.TrialFunction(V)
m_tilde = dl.TestFunction(V)

a1 = dl.inner(A1*dl.grad(m_hat), dl.grad(m_tilde))*dl.dx + dl.inner(m_hat, m_tilde)*dl.dx
a2 = dl.inner(A2*dl.grad(m_hat), dl.grad(m_tilde))*dl.dx + dl.inner(m_hat, m_tilde)*dl.dx

b = dl.inner(f,m_tilde)*dl.dx

m1 = dl.Function(V)
dl.solve(a1 == b, m1)

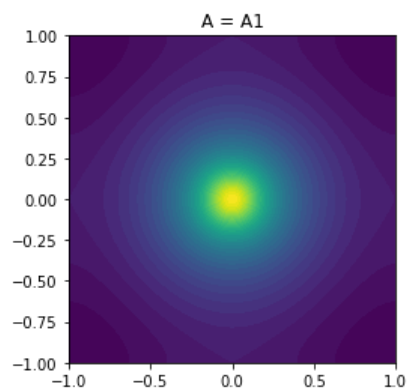
m2 = dl.Function(V)
dl.solve(a2 == b, m2)

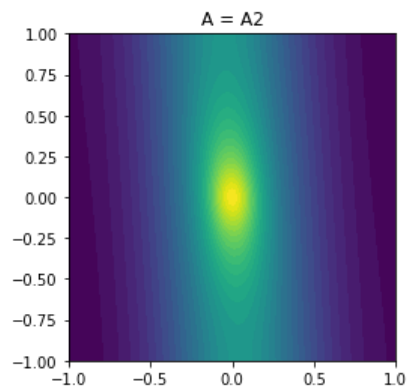
dl.plot(m1, title="A = A1")
plt.show()
dl.plot(m2, title="A = A2")
plt.show()

```

Number of unknowns = 66049

--- Instant: compiling ---





## Part c

When the diffusion matrix of the system,  $A(x) = A_1$ , the solution  $m$  seems to be an independent gaussian distribution. We comment on the fact that the diffusion matrix is a diagonal matrix with equal values on the diagonal. When  $A(x) = A_2$ , the solution  $m$  is "stretched" and seems to yield a dependent gaussian distribution. We note that the off-diagonals are not 0, and the diagonal values are not equal. It seems as though the diffusion matrix,  $A(x)$  is acting as a covariance matrix for the displayed gaussian distribution in the image.