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Task 1

Transform your chances

1.1 Introduction

This task required us to perform convolution on two given functions using the Fourier transforms and output the final convolved function.

1.2 Theory

1.2.1 Convolution

The convolution [1] of two functions f_1, f_2 is given by:

$$f_1 * f_2 = \int_{-\infty}^{\infty} f_1(k) f_2(x - k) dk$$

1.2.2 Fourier Transform

The Fourier transform [3] of a function $f(x)$ is given by:

$$\mathcal{F}\{f\}(k) = \int_{-\infty}^{\infty} f(x) e^{-2\pi i k x} dx$$

1.2.3 The Convolution Theorem

The convolution theorem [2] states that the Fourier transform of the convolution of two functions is equal to the point-wise product of the Fourier transforms of the functions. Mathematically,

$$\mathcal{F}\{f_1 * f_2\}(k) = \mathcal{F}\{f_1\}(k) \cdot \mathcal{F}\{f_2\}(k)$$

In simpler terms, we can say that the convolution of two functions f_1, f_2 is given by:

$$f_1 * f_2 = \mathcal{F}^{-1}\{\mathcal{F}\{f_1\} \cdot \mathcal{F}\{f_2\}\}$$

1.3 Code

1.3.1 Reading the functions

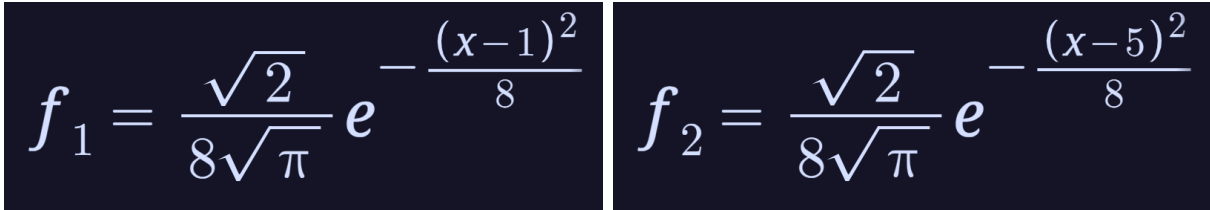
The first step was to read the functions from the given file. The functions were given in latex form. I converted them to sympy expressions using the `latex2sympy2` module.

Listing 1.1: Reading the functions

```
from latex2sympy2 import latex2sympy as l2s

with open(sys.argv[1], "r") as file:
    f1, f2 = (l2s(line) for line in file.readlines())
```

The functions f_1, f_2 are:



(a) f_1 (b) f_2

Figure 1.1: Functions f_1 and f_2

1.3.2 Fourier Transform

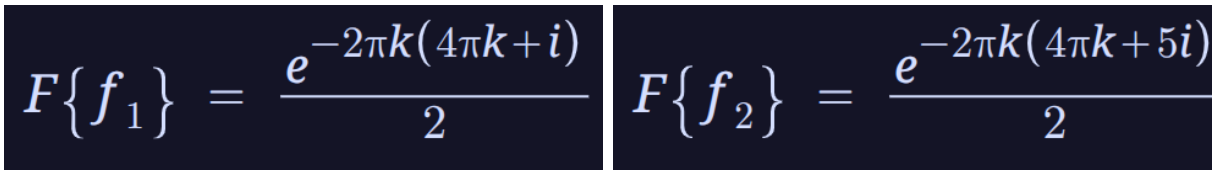
The next step was to find the Fourier transforms of the functions f_1, f_2 . This was done using the `sympy` module, which has in-built functions for Fourier Transforms.

Listing 1.2: Finding the Fourier transforms of the functions

```
x, k = sp.symbols("x_k")

F1 = sp.fourier_transform(f1, x, k)
F2 = sp.fourier_transform(f2, x, k)
```

The values of F1 and F2 are:



(a) $\mathcal{F}\{f_1\}$ (b) $\mathcal{F}\{f_2\}$

Figure 1.2: Fourier Transforms of f_1 and f_2

1.3.3 Convolution

Finally, the convolution of the two functions was found by taking the inverse Fourier transform of the product of the Fourier transforms of the functions.

Listing 1.3: Finding the Convolution

```
F = sp.inverse_fourier_transform(F1 * F2, k, x)
```

The final convolved function is:

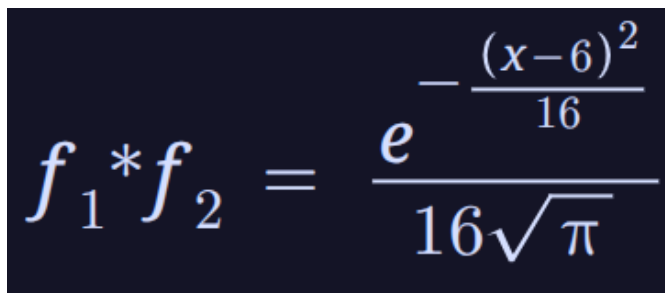

$$f_1 * f_2 = \frac{e^{-\frac{(x-6)^2}{16}}}{16\sqrt{\pi}}$$

Figure 1.3: Convolved function $f_1 * f_2$

1.4 Observations

Here are a few things that I observed, while working on the Task

- The **sympy** has a lot of in-built functions that made it very easy to perform the Fourier Transforms and Inverse Fourier Transforms.
- While researching, I found that **sympy** has an in-built function for Convolution as well. However, that is only for discrete convolution, and not for continuous convolution.
- To compare the performance of convolution using Fourier Transforms vs traditional convolution, I implemented the traditional convolution as well. The implementation is shown below:

```
F = sp.integrate(f1.subs(x, k) * f2.subs(x, x - k), (k, -sp.oo, sp.oo))
```

I used the time interval of execution to compare the two methods. The Fourier Transform method had an average execution time of **0.913** seconds, while the traditional convolution method had an average execution time of **0.312**. I believe that the traditional convolution method is faster because it doesn't involve any complex mathematical operations such as Fourier Transforms, but rather just a simple integral.

- Graphically, the functions f_1 , f_2 and their convolved function look as shown below:

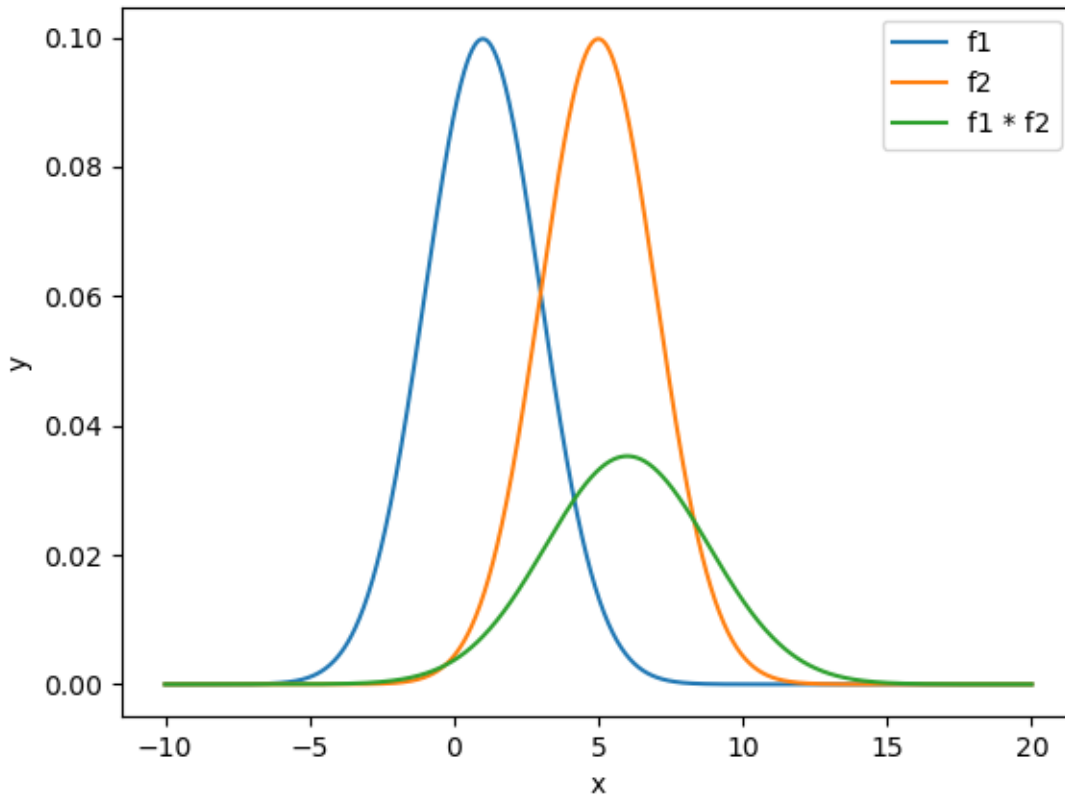


Figure 1.4: Graphs of $f_1, f_2, f_1 * f_2$

- In the example above, the functions f_1, f_2 are gaussian functions. We can observe that the convolved function is also a gaussian function. This shows that the convolution of two gaussian functions is also a gaussian function.

1.5 Conclusion

- This was a very interesting task. I learnt the basics of Fourier Transforms and Convolution and the relationship between them.
- `sympy` made this task pretty easy to implement. It allowed me to easily convert the latex functions to sympy expressions and do the necessary mathematical operations.

Task 2

Poised for Poiseuille flow

2.1 Introduction

In this task, we are required to solve the Navier-Stokes equation [4]. The problem is simplified using the following assumptions:

- Incompressible flow (ρ is constant) $\implies \partial\rho = 0$
- Fully Developed Flow (z-velocity not dependent on z) $\implies \frac{\partial v_z}{\partial z} = 0$
- θ - symmetric flow (θ components and their changes can be neglected) $\implies v_\theta = 0$
- Impenetrable wall (Zero radial velocity at a radius equal to pipe radius) $\implies v_r(r = R) = 0$
- Continuous and Smooth (Differentiable) flow profile.

2.2 Theory

2.2.1 Navier-Stokes Equation

Beginning with the general form of the Navier-Stokes equation:

$$\frac{\partial \rho}{\partial t} + \frac{1}{r} \frac{\partial(\rho r v_r)}{\partial \rho} + \frac{\partial(\rho v_z)}{\partial z} + \frac{1}{r} \frac{\partial \rho v_\theta}{\partial \theta} = 0$$

We can simplify this equation using the assumptions given above to get:

$$\begin{aligned} \frac{1}{r} \frac{\partial(\rho r v_r)}{\partial \rho} &= 0 \\ r v_r &= \text{constant} \\ v_r &= \frac{c}{r} \end{aligned}$$

At $r=0$, v_r is finite. Therefore, $c = 0$ and $v_r = 0$.

Using this, we can now assume that $\vec{v} = v_z \hat{e}_z$. The equation now becomes:

$$\frac{\partial(\rho \vec{v})}{\partial t} + (\vec{v} \cdot \nabla) \vec{v} = \rho \vec{g} - \nabla P + \mu \nabla^2 \vec{v}$$

This is simplified to:

$$\frac{\partial P}{\partial z} = \mu \nabla^2 v_z$$

$$\frac{\partial P}{\partial z} = \mu \left(\frac{\partial^2 v_z}{\partial r^2} + \frac{1}{r} \frac{\partial v_z}{\partial r} \right)$$

From product rule, we get that:

$$\frac{1}{r} \frac{\partial \left(r \frac{\partial v_z}{\partial r} \right)}{\partial r} = \mu \left(\frac{\partial^2 v_z}{\partial r^2} + \frac{1}{r} \frac{\partial v_z}{\partial r} \right)$$

The final equation to solve is:

$$\frac{\partial P}{\partial z} = \frac{1}{r} \frac{\partial \left(r \frac{\partial v_z}{\partial r} \right)}{\partial r}$$

To solve this equation:

$$\frac{\partial P}{\partial z} r \partial r = \partial \left(r \frac{\partial v_z}{\partial r} \right)$$

Integrating both sides, we get:

$$\frac{\partial P}{\partial z} \frac{r^2}{2} = r \frac{\partial v_z}{\partial r} + C_2$$

$$\frac{\partial P}{\partial z} \frac{r}{2} \partial r = \partial v_z + \frac{C_2}{r} \partial r$$

Integrating once again, we get:

$$v_z = C_1 + C_2 \cdot \log(r) + \frac{\left(\frac{\partial P}{\partial z} \right) r^2}{4}$$

The differential equation can also be solved using the `sp.dsolve` function in `sympy`, to get the same result.

The boundary conditions are:

- $v_z(0) \neq \pm\infty$
- $v_z(R) = 0$

Applying the boundary conditions, we get:

$$v_z = \frac{1}{4} \frac{\partial P}{\partial z} (r^2 - 1)$$

2.3 Code

2.3.1 Reading the function

The first step was to read the given function and convert it from latex to a sympy expression.

Listing 2.1: Reading the function

```
import sympy as sp
from latex2sympy2 import latex2sympy as l2s

with open(sys.argv[1], "r") as file:
    P = l2s(file.readline())
```

2.3.2 Solving the equation

To solve the Navier-Stokes equation, I used `sp.dsolve` function, which solves the given differential equation.

$$\nabla P = \frac{\mu}{r} \frac{\partial(r \frac{\partial v_z}{\partial r})}{\partial r}$$

Listing 2.2: Solving the equation

```
navier_stokes = sp.Eq(
    sp.Derivative(rho * vz, z) + vz * sp.Derivative(vz, z),
    rho * gz
    - sp.Derivative(P, z)
    + mu * (sp.Derivative((r * sp.Derivative(vz, r)), r)) / r,
)

solution = sp.dsolve(navier_stokes, vz)
```

The general solution obtained from the `sp.dsolve` function is:

$$v_z = C_1 + C_2 \cdot \log(r) + \frac{(\nabla P)r^2}{4}$$

2.3.3 Applying the boundary conditions

The boundary conditions are applied to the general solution to get the final solution.

Listing 2.3: Applying the boundary conditions

```
general_solution = solution.rhs
C1, C2 = sp.symbols("C1_C2")

boundary_conditions = [
    sp.Eq(general_solution.subs(r, 0), C1),
    sp.Eq(general_solution.subs(r, R), 0),
```

```

]

constants = sp.solve(boundary_conditions, (C1, C2))
particular_solution = general_solution.subs(constants)

assumptions = {
    "g_z": 0,
    "mu": 1,
    "R": 1,
}

final_solution = particular_solution.subs(assumptions)

```

2.3.4 Generating the .cpp file

Finally, the solution is written to a .cpp file using the `sp.ccode` function, which converts the sympy expression to a C++ expression. I assumed that the executable's name is `vel.out`.

```

print(
    """#include <iostream>
#include <cmath>
int main(int argc, char *argv[]) {{
    double r = std::stod(argv[1]);
    std::cout << std::abs({}) << "\\n";
}}""".format(
    str(sp.ccode(final_solution))
),
    file=open("vel.cpp", "w"),
)

import subprocess

subprocess.run(["g++", "-O2", "vel.cpp", "-o", "vel.out"])

```

2.4 Observations

- The Navier-Stokes equation is a very complex equation to solve. The assumptions used in this task simplified the equation to a great extent.
- Once again, the `sympy` module impressed me with its capabilities. It was able to solve the differential equations and also apply the boundary conditions to get the final solution. Moreover, it has an in-built function `sp.ccode` to convert the sympy expression to a C++ expression, which, I believe is pretty cool.

2.5 Conclusion

- I learnt a lot about the Navier-Stokes equation and the basics of fluid dynamics through working on this task. Overall, it was another interesting task, and I especially liked it, when I found out about the `sp.ccode` function.

Bibliography

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