UQ

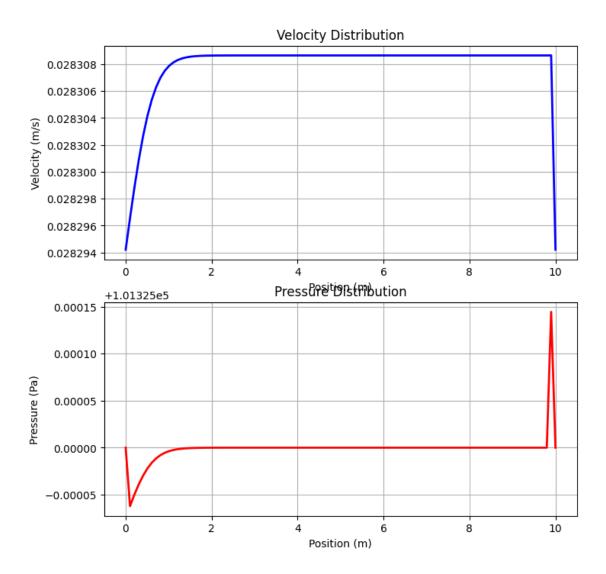
September 18, 2023

Fluid simulation

```
[]: import numpy as np
              import matplotlib.pyplot as plt
              import math
              def calculate_friction_factor(Re, pipe_roughness, pipe_diameter):
                          # Hagen-Poiseuille equaiton for luminar
                         if Re < 2000:
                                     f = 64 / Re
                          # Swamee-Jain equation for turbulance
                         else:
                                     f = 0.25 / (math.log10((pipe_roughness / (3.7 * pipe_diameter)) + (5.74 / pipe_roughness)) + (5.74 / pipe_roughness) + (
                 → Re**0.9)))**2
                         return f
              def simulate_1D_pipe_flow(pipe_length, pipe_diameter, pipe_roughness,_
                 →inlet_flow_rate, inlet_pressure, outlet_pressure, fluid_density,
                 →fluid_viscosity, total_time, num_points):
                         # Calculate pipe cross-sectional area
                         area = np.pi * (pipe_diameter / 2)**2
                         # Calculate the initial velocity
                         initial_velocity = inlet_flow_rate / area
                         velocity = np.ones(num_points) * initial_velocity
                         pressure = np.linspace(inlet_pressure, outlet_pressure, num_points)
                         dx = pipe_length / (num_points - 1)
                          # Calculate the maximum velocity (for time step estimation)
                         max_velocity = 1.2 * initial_velocity
                         # Estimate the time step
                         C = 0.1 # Courant number
                         time_step = C * min(dx / max_velocity, dx**2 / fluid_viscosity)
                         num_time_steps = int(total_time / time_step)
```

```
for t in range(num_time_steps):
        for i in range(1, num_points - 1):
            Re = fluid_density * abs(velocity[i]) * pipe_diameter /__
 \hookrightarrowfluid_viscosity
            f = calculate_friction_factor(Re, pipe_roughness, pipe_diameter)
            # Crank-Nicolson method with 2nd-order upwind scheme
            a = fluid_viscosity * time_step / (2 * dx**2)
            b = f * velocity[i] * abs(velocity[i]) * time_step / (4 *_
→pipe_diameter)
            c = fluid_density * dx / (2 * time_step)
            if velocity[i] >= 0:
                A = c - a + 0.5 * fluid_density * velocity[i] * dx
                B = c + a
                C = -a - b
                D = -a + b
            else:
                A = c - a
                B = c + a + 0.5 * fluid_density * velocity[i] * dx
                C = -a - b
                D = -a + b
            velocity[i] = (A * velocity[i - 1] + B * velocity[i] + C *_
\rightarrowvelocity[i + 1]) / (A + B + C + D)
        pressure[0] = inlet_pressure
        pressure[-1] = outlet_pressure
        pressure[1:-1] = pressure[1:-1] - fluid_density * velocity[1:-1] *__
→(velocity[2:] - velocity[:-2]) * dx
    return velocity, pressure
# Define pipe parameters
pipe_length = 10.0 #m
pipe_diameter = 0.03 #m
pipe_roughness = 0.0001
inlet_flow_rate = 0.00002 \# m^3/s
inlet_pressure = 101325.0 # Pa
outlet_pressure = 101325.0 # Pa
fluid_density = 1000.0 \# kq/m^3
fluid_viscosity = 0.001 # Pa*s
total_time = 2.0 # s
num_points = 100
# Run the simulation
```

```
velocity, pressure = simulate_1D_pipe_flow(pipe_length, pipe_diameter,_u
→pipe_roughness, inlet_flow_rate, inlet_pressure, outlet_pressure,
→fluid_density, fluid_viscosity, total_time, num_points)
# Calculate the position of each point
x = np.linspace(0, pipe_length, num_points)
# Calculate the flow rate at each point
flow_rate = velocity * np.pi * (pipe_diameter / 2)**2
# Plot the results
fig, axs = plt.subplots(2, figsize=(8, 8))
axs[0].plot(x, velocity, 'b-', linewidth=2)
axs[0].set_xlabel('Position (m)')
axs[0].set_ylabel('Velocity (m/s)')
axs[0].set_title('Velocity Distribution')
axs[0].grid(True)
axs[1].plot(x, pressure, 'r-', linewidth=2)
axs[1].set_xlabel('Position (m)')
axs[1].set_ylabel('Pressure (Pa)')
axs[1].set_title('Pressure Distribution')
axs[1].grid(True)
plt.show()
avg_velocity = np.mean(velocity)
print("Average Velocity: {:.8f} m/s".format(avg_velocity))
```



Average Velocity: 0.02830785 m/s

```
[]: import numpy as np
from scipy.stats import lognorm
import matplotlib.pyplot as plt

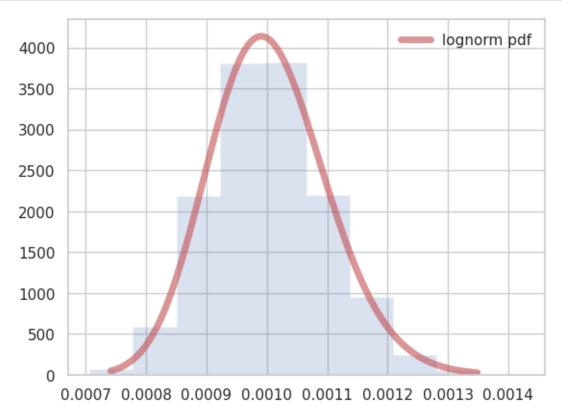
# Paremeters setting
mu = -6.907755 # mean value
sigma = 0.096809 # standard deviation

lognorm_dist = lognorm(s=sigma, scale=np.exp(mu))

samples = lognorm_dist.rvs(size=10000)

# Draw pdf
```

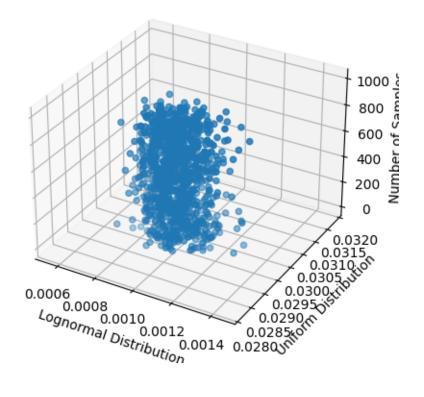
```
fig, ax = plt.subplots(1, 1)
x = np.linspace(lognorm_dist.ppf(0.001), lognorm_dist.ppf(0.999), 100)
ax.plot(x, lognorm_dist.pdf(x), 'r-', lw=5, alpha=0.6, label='lognorm pdf')
ax.hist(samples, density=True, histtype='stepfilled', alpha=0.2)
ax.legend(loc='best', frameon=False)
plt.show()
```



```
[]: import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D

# Setting
mu = -6.907755 # Mean value
sigma = 0.096809 # Standard deviation
num_samples = 1000 # Number of samples

# Produce the samples
fluid_viscosity_samples = np.random.lognormal(mu, sigma, num_samples)
pipe_diameter_samples = np.random.uniform(0.029, 0.03, num_samples)
# Draw
```



Monte Carlo Solver

```
[]: import numpy as np
import matplotlib.pyplot as plt
import math
from tqdm import tqdm
import csv

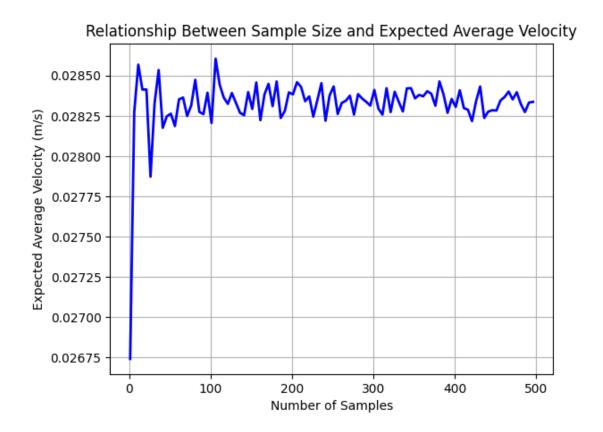
def calculate_friction_factor(Re, pipe_roughness, pipe_diameter):
```

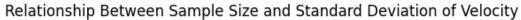
```
# Hagen-Poiseuille equaiton for luminar
         if Re < 2000:
                   f = 64 / Re
          # Swamee-Jain equation for turbulance
                   f = 0.25 / (math.log10((pipe_roughness / (3.7 * pipe_diameter)) + (5.74 / pipe_roughness)) + (5.74 / pipe_roughness) + (
  \rightarrow Re**0.9)))**2
         return f
def simulate_1D_pipe_flow(pipe_length, pipe_diameter, pipe_roughness,_
  →inlet_flow_rate, inlet_pressure, outlet_pressure, fluid_density, 
  →fluid_viscosity, total_time, num_points):
         # Calculate pipe cross-sectional area
         area = np.pi * (pipe_diameter / 2)**2
         # Calculate the initial velocity
         initial_velocity = inlet_flow_rate / area
         velocity = np.ones(num_points) * initial_velocity
         pressure = np.linspace(inlet_pressure, outlet_pressure, num_points)
         dx = pipe_length / (num_points - 1)
         # Calculate the maximum velocity (for time step estimation)
         max_velocity = 1.2 * initial_velocity
         # Estimate the time step
         C = 0.1 \# Courant number
         time_step = C * min(dx / max_velocity, dx**2 / fluid_viscosity)
         num_time_steps = int(total_time / time_step)
         for t in range(num_time_steps):
                   for i in range(1, num_points - 1):
                            Re = fluid_density * abs(velocity[i]) * pipe_diameter /_
  →fluid_viscosity
                            f = calculate_friction_factor(Re, pipe_roughness, pipe_diameter)
                            # Crank-Nicolson method with 2nd-order upwind scheme
                            a = fluid_viscosity * time_step / (2 * dx**2)
                            b = f * velocity[i] * abs(velocity[i]) * time_step / (4 *_
  →pipe_diameter)
                            c = fluid_density * dx / (2 * time_step)
                            if velocity[i] >= 0:
                                      A = c - a + 0.5 * fluid_density * velocity[i] * dx
                                     B = c + a
                                      C = -a - b
```

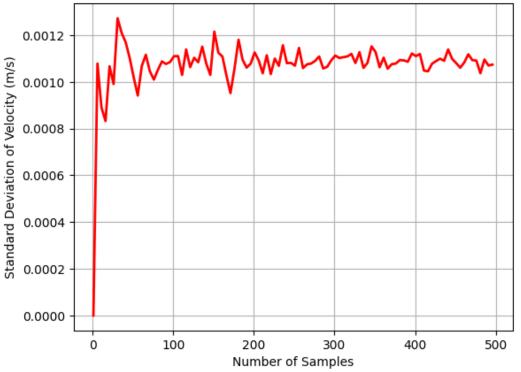
```
D = -a + b
            else:
                A = c - a
                B = c + a + 0.5 * fluid_density * velocity[i] * dx
                C = -a - b
                D = -a + b
            velocity[i] = (A * velocity[i - 1] + B * velocity[i] + C *_
\rightarrowvelocity[i + 1]) / (A + B + C + D)
        pressure[0] = inlet_pressure
        pressure[-1] = outlet_pressure
        pressure[1:-1] = pressure[1:-1] - fluid_density * velocity[1:-1] *__
 \hookrightarrow (velocity[2:] - velocity[:-2]) * dx
    return velocity, pressure
# Define pipe parameters
pipe_length = 10.0 #m
pipe_roughness = 0.0001
inlet_flow_rate = 0.00002 \# m^3/s
inlet_pressure = 101325.0 # Pa
outlet_pressure = 101325.0 # Pa
fluid_density = 1000.0 \# kq/m^3
total_time = 2.0 # s
num_points = 100
# Define the log-normal distribution of fluid viscosity
mu = -6.907755 # Log-normal distribution mean
sigma = 0.096809 # Log-normal distribution standard deviation
# Define the range of sample sizes to consider
min_samples = 1
max_samples = 500
step = 5
# Initialize arrays to store results
sample_sizes = np.arange(min_samples, max_samples + 1, step)
expected_avg_velocities = np.zeros(len(sample_sizes))
velocity_std_devs = np.zeros(len(sample_sizes))
# Loop over different sample sizes and calculate the expected average velocity\Box
\rightarrow for each
for i, num_samples in enumerate(tqdm(sample_sizes, desc='Processing samples')):
    # Generate samples of fluid viscosity using Monte Carlo simulation
    fluid_viscosity_samples = np.random.lognormal(mu, sigma, num_samples)
    pipe_diameter_samples = np.random.uniform(0.029, 0.031, num_samples)
```

```
# Calculate the average velocity for each sample of fluid viscosity
    avg_velocities = []
    for j in range(num_samples):
        _, pressure = simulate_1D_pipe_flow(pipe_length,_
 →pipe_diameter_samples[j], pipe_roughness, inlet_flow_rate, inlet_pressure,_
 →outlet_pressure, fluid_density, fluid_viscosity_samples[j], total_time,_
 →num_points)
        avg_velocities.append(np.mean(_))
    # Calculate the expected value of the average velocity
    expected_avg_velocities[i] = np.mean(avg_velocities)
    velocity_std_devs[i] = np.std(avg_velocities)
# Save the results to a CSV file
with open('results.csv', 'w', newline='') as csvfile:
    csv_writer = csv.writer(csvfile)
    csv_writer.writerow(['Number of Samples', 'Expected Average Velocity (m/s)', __
 →'Standard Deviation of Velocity (m/s)'])
    for i in range(len(sample_sizes)):
        csv_writer.writerow([sample_sizes[i], expected_avg_velocities[i],__
 →velocity_std_devs[i]])
# Plot the relationship between sample size and expected average velocity
plt.figure(1)
plt.plot(sample_sizes, expected_avg_velocities, 'b-', linewidth=2)
plt.xlabel('Number of Samples')
plt.ylabel('Expected Average Velocity (m/s)')
plt.title('Relationship Between Sample Size and Expected Average Velocity')
plt.grid(True)
# Plot the relationship between sample size and standard deviation of velocity
plt.figure(2)
plt.plot(sample_sizes, velocity_std_devs, 'r-', linewidth=2)
plt.xlabel('Number of Samples')
plt.ylabel('Standard Deviation of Velocity (m/s)')
plt.title('Relationship Between Sample Size and Standard Deviation of Velocity')
plt.grid(True)
plt.show()
```

Processing samples: 100%|| 100/100 [00:57<00:00, 1.75it/s]







GMs(Polynominal Chaos Expansion)

```
[]: import numpy as np
     import matplotlib.pyplot as plt
     import chaospy as cp
     from tqdm import tqdm
     import math
     import csv
     def calculate_friction_factor(Re, pipe_roughness, pipe_diameter):
          if Re < 2000:
              f = 64 / Re
          else:
              f = 0.25 / (math.log10((pipe_roughness / (3.7 * pipe_diameter)) + (5.74 / pipe_diameter)) + (5.74 / pipe_diameter)) + (5.74 / pipe_diameter))
      → Re**0.9)))**2
          return f
     def simulate_1D_pipe_flow(pipe_length, pipe_diameter, pipe_roughness,_
      →inlet_flow_rate, inlet_pressure, outlet_pressure, fluid_density, __
      →fluid_viscosity, total_time, num_points):
          area = np.pi * (pipe_diameter / 2)**2
          initial_velocity = inlet_flow_rate / area
```

```
velocity = np.ones(num_points) * initial_velocity
    pressure = np.linspace(inlet_pressure, outlet_pressure, num_points)
    dx = pipe_length / (num_points - 1)
    max_velocity = 1.2 * initial_velocity
    C = 0.1
    time_step = C * min(dx / max_velocity, dx**2 / fluid_viscosity)
    num_time_steps = int(total_time / time_step)
    for t in range(num_time_steps):
        for i in range(1, num_points - 1):
            Re = fluid_density * abs(velocity[i]) * pipe_diameter /__
→fluid_viscosity
            f = calculate_friction_factor(Re, pipe_roughness, pipe_diameter)
            a = fluid_viscosity * time_step / (2 * dx**2)
            b = f * velocity[i] * abs(velocity[i]) * time_step / (4 *_
→pipe_diameter)
            c = fluid_density * dx / (2 * time_step)
            if velocity[i] >= 0:
                A = c - a + 0.5 * fluid_density * velocity[i] * dx
                B = c + a
                C = -a - b
                D = -a + b
            else:
                A = c - a
                B = c + a + 0.5 * fluid_density * velocity[i] * dx
                C = -a - b
                D = -a + b
            velocity[i] = (A * velocity[i - 1] + B * velocity[i] + C *_
\rightarrowvelocity[i + 1]) / (A + B + C + D)
        pressure[0] = inlet_pressure
        pressure[-1] = outlet_pressure
        pressure[1:-1] = pressure[1:-1] - fluid_density * velocity[1:-1] *__
\hookrightarrow (velocity[2:] - velocity[:-2]) * dx
    return velocity, pressure
# Pipe parameters
pipe_length = 10.0
pipe_roughness = 0.0001
inlet_flow_rate = 0.00002
inlet_pressure = 101325.0
outlet_pressure = 101325.0
```

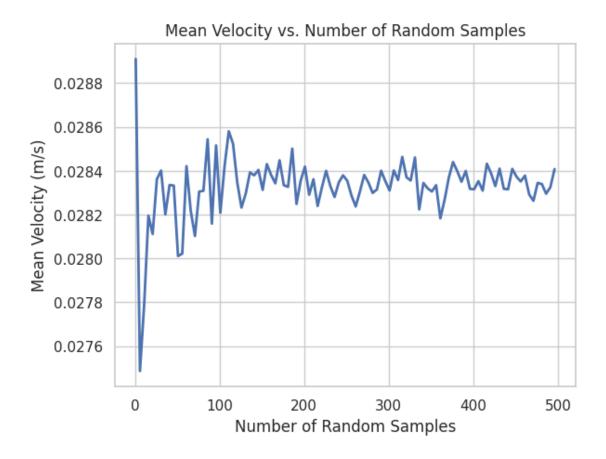
```
fluid_density = 1000.0
total_time = 2.0
num_points = 100
def uncertainty_analysis(fluid_viscosity_distribution,_
→pipe_diameter_distribution, num_samples):
    joint_distribution = cp.J(cp.LogNormal(mu=-6.907755, sigma=0.096809), cp.
\rightarrowUniform(0.029, 0.031))
    samples = joint_distribution.sample(num_samples)
    avg_velocities = np.array([simulate_1D_pipe_flow(pipe_length, d,__
 →pipe_roughness, inlet_flow_rate, inlet_pressure, outlet_pressure, __
 →fluid_density, v, total_time, num_points)[0].mean() for v, d in samples.T])
    pce_order = 2
    basis_polynomials = cp.expansion.stieltjes(pce_order, joint_distribution)
    pce_coeffs = cp.fit_regression(basis_polynomials, samples, avg_velocities)
    return pce_coeffs, basis_polynomials
# Run the uncertainty analysis using 100 random samples
num_samples = 100
pce_coeffs, basis_polynomials = uncertainty_analysis(cp.LogNormal(mu=-6.907755,_
→sigma=0.096809), cp.Uniform(0.029, 0.031), num_samples)
pce_expansion = cp.sum(pce_coeffs * basis_polynomials)
print("PCE coefficients:", pce_coeffs)
# Define the range of sample sizes to consider
min_samples = 1
max_samples = 500
step = 5
sample_sizes = np.arange(min_samples, max_samples + 1, step)
pce_avg_velocities = []
pce_std_devs = []
# Create a CSV file to store the results
with open('pce_results.csv', mode='w', newline='') as csvfile:
    csv_writer = csv.writer(csvfile)
    csv_writer.writerow(['Number of Samples', 'Mean Velocity', 'Standardu
→Deviation'])
    # Run the uncertainty analysis for different sample sizes
    for num_random_samples in tqdm(sample_sizes, desc='Processing samples'):
        random_samples = cp.J(cp.LogNormal(mu=-6.907755, sigma=0.096809), cp.

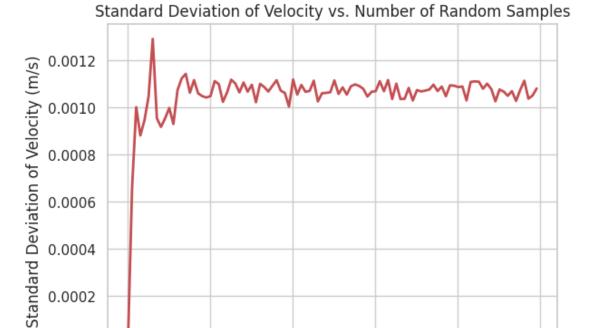
→Uniform(0.029, 0.031)).sample(num_random_samples)
```

```
avg_velocity = np.mean([pce_expansion(*sample) for sample in_
 →random_samples.T])
        std_dev = np.std([pce_expansion(*sample) for sample in random_samples.T])
        pce_avg_velocities.append(avg_velocity)
        pce_std_devs.append(std_dev)
        # Write the results to the CSV file
        csv_writer.writerow([num_random_samples, avg_velocity, std_dev])
# Plot the results
plt.figure(1)
plt.plot(sample_sizes, pce_avg_velocities, 'b-', linewidth=2)
plt.xlabel('Number of Random Samples')
plt.ylabel('Mean Velocity (m/s)')
plt.title('Mean Velocity vs. Number of Random Samples')
plt.grid(True)
plt.figure(2)
plt.plot(sample_sizes, pce_std_devs, 'r-', linewidth=2)
plt.xlabel('Number of Random Samples')
plt.ylabel('Standard Deviation of Velocity (m/s)')
plt.title('Standard Deviation of Velocity vs. Number of Random Samples')
plt.grid(True)
plt.show()
```

PCE coefficients: -5665.163709832621*q1**3+15.57300847153761*q0*q1**2+7184.423370636065*q0**2*q1-888.3674389939833*q0**3+605.9448976069962*q1**2-13.618935942079386*q0*q1-214.32567903768657*q0**2-22.94086298272593*q1+0.40818645752662946*q0+0.3241287090136312

Processing samples: 100%|| 100/100 [02:37<00:00, 1.58s/it]





Number of Random Samples

Comparision Between The Monte Carlo Method and PCE (different order).

0.0000

```
def simulate_1D_pipe_flow(pipe_length, pipe_diameter, pipe_roughness,_
→inlet_flow_rate, inlet_pressure, outlet_pressure, fluid_density,
→fluid_viscosity, total_time, num_points):
    area = np.pi * (pipe_diameter / 2)**2
    initial_velocity = inlet_flow_rate / area
    velocity = np.ones(num_points) * initial_velocity
    pressure = np.linspace(inlet_pressure, outlet_pressure, num_points)
    dx = pipe_length / (num_points - 1)
    max_velocity = 1.2 * initial_velocity
    C = 0.1
    time_step = C * min(dx / max_velocity, dx**2 / fluid_viscosity)
    num_time_steps = int(total_time / time_step)
    for t in range(num_time_steps):
        for i in range(1, num_points - 1):
            Re = fluid_density * abs(velocity[i]) * pipe_diameter /__
→fluid_viscosity
            f = calculate_friction_factor(Re, pipe_roughness, pipe_diameter)
            a = fluid_viscosity * time_step / (2 * dx**2)
            b = f * velocity[i] * abs(velocity[i]) * time_step / (4 *_
→pipe_diameter)
            c = fluid_density * dx / (2 * time_step)
            if velocity[i] >= 0:
                A = c - a + 0.5 * fluid_density * velocity[i] * dx
                B = c + a
                C = -a - b
                D = -a + b
            else:
                B = c + a + 0.5 * fluid_density * velocity[i] * dx
                C = -a - b
                D = -a + b
            velocity[i] = (A * velocity[i - 1] + B * velocity[i] + C *_
\rightarrowvelocity[i + 1]) / (A + B + C + D)
        pressure[0] = inlet_pressure
        pressure[-1] = outlet_pressure
        pressure[1:-1] = pressure[1:-1] - fluid_density * velocity[1:-1] *__
 \hookrightarrow (velocity[2:] - velocity[:-2]) * dx
    return velocity, pressure
# Pipe parameters
```

```
pipe_length = 10.0
pipe_roughness = 0.0001
inlet_flow_rate = 0.00002
inlet_pressure = 101325.0
outlet_pressure = 101325.0
fluid_density = 1000.0
total\_time = 2.0
num_points = 100
def uncertainty_analysis(fluid_viscosity_distribution,_
→pipe_diameter_distribution, num_samples, pce_order):
    joint_distribution = cp.J(cp.LogNormal(mu=-6.907755, sigma=0.096809), cp.
 \hookrightarrow Uniform(0.029, 0.031))
    samples = joint_distribution.sample(num_samples)
    avg_velocities = np.array([simulate_1D_pipe_flow(pipe_length, d,__
 →pipe_roughness, inlet_flow_rate, inlet_pressure, outlet_pressure, __
 →fluid_density, v, total_time, num_points)[0].mean() for v, d in samples.T])
    basis_polynomials = cp.expansion.stieltjes(pce_order, joint_distribution)
    pce_coeffs = cp.fit_regression(basis_polynomials, samples, avg_velocities)
    return pce_coeffs, basis_polynomials
def plot_pce_vs_mc(pce_order):
    pce_coeffs, basis_polynomials = uncertainty_analysis(cp.LogNormal(mu=-6.
→907755, sigma=0.096809), cp.Uniform(0.029, 0.031), num_samples, pce_order)
    pce_expansion = cp.sum(pce_coeffs * basis_polynomials)
    # Calculate average velocities using PCE method
    avg_velocities_pce = pce_expansion(*samples_mc)
    # Plot the probability density functions
    sns.kdeplot(avg_velocities_pce, label=f'PCE Order {pce_order}', linewidth=2)
# Run the uncertainty analysis using 100 random samples
num\_samples = 100
# Generate random samples
num_samples_mc = 10000
joint_distribution = cp.J(cp.LogNormal(mu=-6.907755, sigma=0.096809), cp.
\rightarrowUniform(0.029, 0.031))
samples_mc = joint_distribution.sample(num_samples_mc)
# Calculate average velocities using Monte Carlo method
```

Progress: 100%|| 10000/10000 [00:37<00:00, 267.75it/s]

