

Homework 2

April 27, 2020

Motor Design Problem / Heather Miller / Started: 4/22/20

```
[3]: import numpy as np
      from scipy.stats import norm
```

Motor Design problem : weight of motor must be <22kgs

```
[23]: # answers for each part will be stored here
      success_probabilities = []

      # number of samples to run
      samples = 100000

      #limit in kgs
      limit = 22
```

Given Constraints

```
[24]: # Motor Design Variables
      Lwa = 14.12 # armature wire length (m)
      Lwf = 309.45 # field wire length (m)
      ds = 0.00612 # slot depth (m)

      # Coupling Variables b, shared with control problem
      n = 122 # rotational speed (rev/s)
      v = 40 # design voltage (V)
      pmin = 3.94 # minimum required power (kW)
      ymin = 5.12e-3 # minimum required torque (kNm)

      # Parameter Vector a (constants)
      fi = 0.7 # pole arc to pole pitch ratio
      p = 2 # number of poles
      s = 27 # number of slots (teeth on rotor)
      rho = 1.8e-8 # resistivity (ohm-m) of copper at 25C

      # Derived parameters and constants
      mu0 = 4 * np.pi * 1e-7 # magnetic constant
      ap = p # parallel circuit paths (equals poles)
      eff = 0.85 # efficiency
```

```

bfc = 40e-3 # pole depth (m)
fcf = 0.55 # field coil spacing factor
Awa = 2.0074e-006 # cross sectional area of armature winding (m^2)
Awf = 0.2749e-6 # cross sectional area of field coil winding (m^2)

```

Functions used in this code

```

[25]: def calculate_weight(diameter, length, rho_cu, rho_fe):
    # calculate the weight of the motor
    weight = rho_cu * (Awa*Lwa + Awf*Lwf) + rho_fe * length * np.pi *
    pow(diameter + ds, 2)
    return weight

def prob_success_mc(weights, limit):
    # determine the probability of success that the weight of engine will be
    less than a limit
    # add 1 to limit_sum every time weight is under limit weight
    x = len(weights)
    limit_sum = 0
    for i in weights:
        if i < limit:
            limit_sum += 1
    return limit_sum/x

def first_derivative(variable_dict, variable_of_interest):
    # calculate the first derivatives of each of the variables
    h = 0.1 * variable_dict[variable_of_interest][1]
    # inputs [diameter, length, rho_cu, rho_fe]
    inputs = []
    # this loop will put two values into input for each value, if the variable
    selected matches the key it will
    # modify those values with h otherwise both inputs will be the same
    for key in variable_dict:
        if key == variable_of_interest:
            inputs.append([variable_dict[key][0] + h, variable_dict[key][0] - h])
        else:
            inputs.append([variable_dict[key][0], variable_dict[key][0]])
    # calculate the first derivative with the values in the input
    first_d = (calculate_weight(inputs[0][0], inputs[1][0], inputs[2][0],
    inputs[3][0]) -
    calculate_weight(inputs[0][1], inputs[1][1], inputs[2][1],
    inputs[3][1]))/2*h
    return first_d

def calculate_sigma(variable_sigmas, variable_derivatives, correlation_matrix):
    # calculate the sigma of the function

```

```

sigma_squared = 0
for i in range(len(variable_derivatives)):
    for j in range(len(variable_derivatives)):
        sigma_squared += variable_derivatives[i] * variable_derivatives[j] * \
→correlation_matrix[i][j] \
        * variable_sigmas[i] * variable_sigmas[j]
return np.sqrt(sigma_squared)

```

1 Using Monte Carlo and Normal Distribution

- $D \sim N(7.5, 0.5)$ %m rotor diameter
- $L \sim N(9.5, 0.5)$ %m rotor axial length
- $dcu \sim N(8.94e3, 100)$ %copper density density at 25C (kg/m³)
- $dfe \sim N(7.98e3, 100)$ %iron density density at 25C (kg/m³)

```

[26]: d_n = np.random.normal(0.075, 0.005, samples) # rotor diameter (cm)
l_n = np.random.normal(0.095, 0.005, samples) # rotor axial length (cm)
rho_cu_n = np.random.normal(8.94e3, 100, samples) # copper density density at
→25C (kg/m^3)
rho_fe_n = np.random.normal(7.98e3, 100, samples) # iron density density at 25C
→(kg/m^3)
weight_p1 = [calculate_weight(d_n[i], l_n[i], rho_cu_n[i], rho_fe_n[i]) for i in
→range(samples)]
success_probabilities.append(prob_success_mc(weight_p1, limit))

```

2 Using Monte Carlo and Uniform Distribution

- $D \sim \text{Uniform}(6.5, 8.5)$ %m rotor diameter
- $L \sim \text{Uniform}(8.5, 10.5)$ %m rotor axial length
- $dcu \sim \text{Uniform}(8840, 9040)$ %copper density density at 25C (kg/m³)
- $dfe \sim \text{Uniform}(7880, 8080)$ %iron density density at 25C (kg/m³)

```

[27]: d_u = np.random.uniform(0.065, 0.085, samples) # rotor diameter (cm)
l_u = np.random.uniform(0.085, 0.105, samples) # rotor axial length (cm)
rho_cu_u = np.random.uniform(8840, 9040, samples) # copper density density at
→25C (kg/m^3)
rho_fe_u = np.random.uniform(7880, 8080, samples) # iron density density at 25C
→(kg/m^3)
weight_p2 = [calculate_weight(d_u[i], l_u[i], rho_cu_u[i], rho_fe_u[i]) for i in
→range(samples)]
success_probabilities.append(prob_success_mc(weight_p2, limit))

```

3 Using MVFOSM method

```
[28]: # no correlation
correlation_matrix_3 = [[1, 0, 0, 0], [0, 1, 0, 0], [0, 0, 1, 0], [0, 0, 0, 1]]

variables = {"diameter": (0.075, 0.005),
            "length": (0.095, 0.005),
            "rho_cu": (8.94e3, 100),
            "rho_fe": (7.98e3, 100)}

# determine mean of function using variable means
function_mu = calculate_weight(variables["diameter"][0],
                               variables["length"][0], variables["rho_cu"][0],
                               variables["rho_fe"][0])

variable_sigmas = [variables["diameter"][1], variables["length"][1],
                  variables["rho_cu"][1], variables["rho_fe"][1]]

diameter_output = first_derivative(variables, "diameter")
length_output = first_derivative(variables, "length")
cu_output = first_derivative(variables, "rho_cu")
fe_output = first_derivative(variables, "rho_fe")

variable_derivatives = [diameter_output, length_output, cu_output, fe_output]

sigma = calculate_sigma(variable_sigmas, variable_derivatives,
                        correlation_matrix_3)

success_probabilities.append(norm.cdf(22, function_mu, sigma))
```

4 Variables with Correlation.

How does the correlation change the solution?

```
[29]: correlation_matrix_4 = [[1, .2, .3, .7], [.2, 1, .5, .6], [.3, .5, 1, .2], [.7, .6, .2, 1]]

#same variables as previous part
sigma = calculate_sigma(variable_sigmas, variable_derivatives,
                        correlation_matrix_4)

success_probabilities.append(norm.cdf(22, function_mu, sigma))
```

Answers

```
[30]: for i, answer in enumerate(success_probabilities):  
        print("The probability of a motor being less than 22kg with the parameters_"  
        →in #", i+1, " is", answer)  
  
print('The correlation matrix in #4 created a',  
        →(success_probabilities[2]-success_probabilities[3]),  
        " decrease in probability of meeting design parameters.")
```

The probability of a motor being less than 22kg with the parameters in # 1 is
0.98862

The probability of a motor being less than 22kg with the parameters in # 2 is
0.99015

The probability of a motor being less than 22kg with the parameters in # 3 is
0.6064653429109432

The probability of a motor being less than 22kg with the parameters in # 4 is
0.6052891090555454

The correlation matrix in #4 created a 0.0011762338553977791 decrease in
probability of meeting design parameters.

```
[ ]:
```