

1. You are indifferent between receiving A for sure and a lottery that gives you B with a probability of 0.9 and C with a probability of 0.1. You are also indifferent between receiving A for sure and a lottery that gives you B with a probability of 0.7 and D with a probability of 0.3. Your preferences satisfy the von N-M axioms: If the utility of B = 1 and your utility for C is 0, then what are your utilities for A and D?

$$A = 1$$

$$u_A A \sim u_B B + u_C C \sim 1(.9) + 0(.1)$$

$$u_A \sim \boxed{0.9}$$

$$u_A A \sim u_B B + u_D D$$

$$u_D \sim \frac{u_A A - u_B B}{D} \sim \frac{0.9 - (1)(0.7)}{0.3}$$

$$u_D \sim \boxed{0.6667}$$

2. For the Motor Design problem, use the Monte Carlo method to compare the expected utility of two different motor designs. Use the following uncertain parameters: Use the exponential forms in the slides and assume a risk aversion coefficient of 10, a “high” value of 40 and a “low” value of 2. In this problem, assume you prefer less weight. Which design has higher expected utility?

$$D \sim N(7.5, 0.5)$$

$$L \sim N(9.5, 0.5)$$

$$\rho_{cu} \sim N(8.94e3, 100)$$

$$\rho_{fe} \sim N(7.98e3, 100)$$

Using 10000 samples I found that Design 2 had a higher utility (See attached code).

Design 1

- Expected Utility : $\boxed{0.9203}$
- Expected Weight: $\boxed{16.77kg}$

Design 2

- Expected Utility : 0.9798
- Expected Weight: 8.18kg

3. Repeat problem 2 but use the FFNI (i.e. Gaussian quadrature) method using 3 nodes: Which design has higher expected utility? If you were to compare motor designs 1 and 2 as a risk neutral decision maker, which one would be preferred?

$$D \sim N(6.5, 0.5)$$

$$L \sim N(11.5, 0.5)$$

$$\rho_{cu} \sim N(8.94e3, 100)$$

$$\rho_{fe} \sim N(7.98e3, 100)$$

With a risk aversion coefficient of 10 I found that design 2 had a higher utility (See attached code).

Design 1

- Expected Weight: 15.67kg
- CE: 16.21kg
- Expected Utility : 0.9310

Design 2

- Expected Weight: 9.06kg
- CE: 9.31kg
- Expected Utility : 0.9758

As a risk neutral decision maker I found that design 2 still had a higher utility (See attached code).

Design 1

- Expected Weight: 15.67kg
- CE: 15.74kg
- Expected Utility : 0.6403

Design 2

- Expected Weight: 9.06kg
- CE: 9.00kg
- Expected Utility : 0.8142

4. **Redo problem 3 but use the Robust Design formulation in which $CE = \mu + \frac{\sigma^2}{2\rho}$ and the FTNI method (use 5 nodes): Compare the CE equivalents from both problems 3 and 4—how do they differ?**

With a risk aversion coefficient of 10 I found that design 2 had a higher utility (See attached code).

Design 1

- Expected Weight: 15.67kg
- CE: 15.90kg
- Expected Utility : 0.9310

Design 2

- Expected Weight: 9.06kg
- CE: 9.22kg
- Expected Utility : 0.9758

As a risk neutral decision maker I found that design 2 still had a higher utility (See attached code).

Design 1

- Expected Weight: 15.67kg
- CE: 15.66kg
- Expected Utility : 0.6403

Design 2

- Expected Weight: 9.06kg
- CE: 9.06kg
- Expected Utility : 0.8142

For design 1 I found that using the Robust Design formulation resulted in a slightly lower CE than when using the exponential form for both risk adverse ($15.09\text{kg} < 16.21\text{kg}$) and risk neutral view points ($15.66\text{kg} < 15.74\text{kg}$). For design 2 I found that using the Robust Design formulation resulted in a slightly lower CE for risk adverse ($9.22\text{ kg} < 9.31\text{kg}$) but a slightly higher CE for risk neutral ($9.06\text{ kg} > 9.00\text{ kg}$). (See attached code)

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% Design Under Uncertainty : HW4 Utility Theory
% Heather Miller
% Started 5/25/20
% Due 5/27/20

% NOTE: I have modified the provided motorDesignHW4.m file to take in
% only
% variables and design number (1 or 2)

clear all
clc

%adding CTools folder to paths
o_path = path;
path(o_path, 'C:\Users\MTH\Documents\Gradwork
\Spring2020\Design_Under_Uncertainty\HW\HW4\CTools');

problem = 4;

% used for problems 2-4 remaining problems
samples = 10000;
rho = inf; %10 for risk adverse inf for risk neutral
high = 40;
low = 2;

% Problem 1 - Determining the utility of A and D
% uA~uB1+uC and uA~uB2+uD
if problem == 1
    prob_b_1 = 0.9;
    prob_b_2 = 0.7;
    prob_c = 0.1;
    prob_d = 0.3;

    utility_b = 1;
    utility_c = 0;

    utility_a = prob_b_1*utility_b + prob_c*utility_c;
    utility_d = (utility_a-(prob_b_2*utility_b))/prob_d;

    fprintf('Problem 1\n')
    fprintf('Utility of A: %.4f\n', utility_a)
    fprintf('Utility of D: %.4f\n', utility_d)
end

% problem 2 - Monte Carlo method to compare the expected utility of
% two different motor models
if problem == 2
    d_1 = NaN(1, samples);
    d_2 = NaN(1, samples);
    ut_d1 = NaN(1, samples);
    ut_d2 = NaN(1, samples);
```

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for i = 1:samples
    %design variables
    D = normrnd(7.5, 0.5);
    L = normrnd(9.5, 0.5);
    dcu = normrnd(8.94e3, 100);
    dfe = normrnd(7.98e3, 100);

    %weights given randomly generated values
    d_1(i) = motorDesignHW4(D, L, dcu, dfe, 1);
    d_2(i) = motorDesignHW4(D, L, dcu, dfe, 2);

    %utility
    ut_d1(i) = utility(d_1(i), rho, high, low);
    ut_d2(i) = utility(d_2(i), rho, high, low);
end

fprintf('Problem 2\n')
fprintf('Design 1: %.4f\n', sum(d_1)/samples)
fprintf('Utility 1: %.4f\n', sum(ut_d1)/samples)
fprintf('Design 2: %.4f\n', sum(d_2)/samples)
fprintf('Utility 2: %.4f\n', sum(ut_d2)/samples)
end

% problem 3 - Using FFNI method with 3 nodes to compare the expected
% utility of 2 different motor models
if problem ==3
    indx = fullfact([3 3 3 3]);
    %design variables
    [D, d_weight] = qnwnorm(3, 6.5, 0.5^2);
    [L, l_weight] = qnwnorm(3, 11.5, 0.5^2);
    [dcu, dcu_weight] = qnwnorm(3, 8.94e3, 100^2);
    [dfe, dfe_weight] = qnwnorm(3, 7.98e3, 100^2);

    design_1 = NaN(1, length(indx));
    design_2 = NaN(1, length(indx));
    W = NaN(1, length(indx));
    ut_d1 = NaN(1, length(indx));
    ut_d2 = NaN(1, length(indx));

    d1_CE = NaN(1, length(indx));
    d2_CE = NaN(1, length(indx));

    for i = 1: length(indx)
        % the index for each variable to use on this iteration
        D_idx = indx(i, 1);
        L_idx = indx(i, 2);
        dcu_idx = indx(i, 3);
        dfe_idx = indx(i, 4);
        %designs
        design_1(i) = motorDesignHW4(D(D_idx),
L(L_idx),dcu(dcu_idx),...
        dfe(dfe_idx), 1);
        design_2(i) = motorDesignHW4(D(D_idx),
L(L_idx),dcu(dcu_idx),...

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        dfe(dfe_idx, 2);
    % weights
    W(i) =
d_weight(D_idx)*l_weight(L_idx)*dcu_weight(dcu_idx)*....
        dfe_weight(dfe_idx);
    %utility
    ut_d1(i) = utility(design_1(i), rho, high, low)*W(i);
    ut_d2(i) = utility(design_2(i), rho, high, low)*W(i);

    %CE design 1
    d1_CE(i) = exp(design_1(i)/rho);

    %CE design 2
    d2_CE(i) = exp(design_2(i)/rho);

end
% Design 1 moments
[d1_mean, d1_sigma, d1_skew, d1_kurt] = multi_moments(design_1,
W);
% Design 2 moments
[d2_mean, d2_sigma, d2_skew, d2_kurt] = multi_moments(design_2,
W);

%CE design 1
if rho == inf
    d1_CE = CE_exponential(sum(design_1)/length(indx), rho);
else
    d1_CE = CE_exponential(sum(d1_CE)/length(indx), rho);
end

if rho == inf
    d2_CE = CE_exponential(sum(design_2)/length(indx), rho);
else
    d2_CE = CE_exponential(sum(d2_CE)/length(indx), rho);
end

fprintf('Problem 3 : rho = %i\n', rho)
fprintf('Design 1: %.4f\n', d1_mean)
fprintf('CE 1: %.4f\n', d1_CE)
fprintf('Utility 1: %.4f\n', sum(ut_d1))

fprintf('Design 2: %.4f\n', d2_mean)
fprintf('CE 2: %.4f\n', d2_CE)
fprintf('Utility 2: %.4f\n', sum(ut_d2))

end

% problem 4 - using Robust Design formulation and the FTNI with 5
nodes
% to compare the certainty equivalent of two motor models

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if problem == 4
    indx = fullfact([5 5 5 5]);

    %design variables
    [D, d_weight] = qnwnorm(5, 6.5, 0.5^2);
    [L, l_weight] = qnwnorm(5, 11.5, 0.5^2);
    [dcu, dcu_weight] = qnwnorm(5, 8.94e3, 100^2);
    [dfe, dfe_weight] = qnwnorm(5, 7.98e3, 100^2);

    design_1 = NaN(1, length(indx));
    design_2 = NaN(1, length(indx));
    W = NaN(1, length(indx));
    ut_d1 = NaN(1, length(indx));
    ut_d2 = NaN(1, length(indx));

    for i = 1: length(indx)
        % the index for each variable to use on this iteration
        D_idx = indx(i, 1);
        L_idx = indx(i, 2);
        dcu_idx = indx(i, 3);
        dfe_idx = indx(i, 4);
        %designs
        design_1(i) = motorDesignHW4(D(D_idx),
L(L_idx),dcu(dcu_idx),...
        dfe(dfe_idx), 1);
        design_2(i) = motorDesignHW4(D(D_idx),
L(L_idx),dcu(dcu_idx),...
        dfe(dfe_idx), 2);
        % weights
        W(i) =
d_weight(D_idx)*l_weight(L_idx)*dcu_weight(dcu_idx)*....
        dfe_weight(dfe_idx);
        %utility
        ut_d1(i) = utility(design_1(i), rho, high, low)*W(i);
        ut_d2(i) = utility(design_2(i), rho, high, low)*W(i);

    end

    % Design 1 moments
    [d1_mean, d1_sigma, d1_skew, d1_kurt] = multi_moments(design_1,
W);
    % Design 2 moments
    [d2_mean, d2_sigma, d2_skew, d2_kurt] = multi_moments(design_2,
W);

    %CE design 1
    d1_CE = d1_mean + ((d1_sigma^2)/(2*rho));
    d2_CE = d2_mean + ((d2_sigma^2)/(2*rho));

    fprintf('Problem 4: rho = %i\n', rho)
    fprintf('Design 1: %.4f\n', d1_mean)
    fprintf('CE 1: %.4f\n', d1_CE)

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```
fprintf('Utility 1: %.4f\n', sum(ut_d1))

fprintf('Design 2: %.4f\n', d2_mean)
fprintf('CE 2: %.4f\n', d2_CE)
fprintf('Utility 2: %.4f\n', sum(ut_d2))
end

function output = utility(x, rho, high, low)
    if isinf(rho)
        output = (high-x)/(high-low);
    else
        output = (exp(-(high-x)/rho)-1)/(exp(-(high-low)/rho)-1);
    end
end

function [mean, sigma, skew, kurt] = multi_moments(output, Weight)
    mean = dot(Weight, output);
    sigma = sqrt(dot(Weight, (output-mean).^2));
    skew = dot(Weight, (output-mean).^3)/sigma.^3;
    kurt = dot(Weight, (output-mean).^4)/sigma.^4;
end

function certainty_equivalence = CE_exponential(x, rho)
    if isinf(rho)
        certainty_equivalence = x;
    else
        certainty_equivalence = rho*log(x);
    end
end
```