

BEE 4750/5750 Homework 2

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Problem 1

Problem 1.1

```
julia> function calc_conc(x, U, Cs, Co, Bo, No, ka, kc, kn)
    # U is the stream velocity, Cs is the saturated oxygen saturation, Co
    is the initial dissolved DO conc,
    # Bo is the initial CBOD conc, No is the initial NBOD conc, ka is the
    reaeration rate, kc is the CBOD
    # CBOD decay rate, kn is the NBOD decay rate

    # terms that will end up in final equation
    a1 = exp((-ka*x)/U)
    a2 = (kc/(ka-kc))*(exp((-kc*x)/U)-exp((-ka*x/U)))
    a3 = (kn/(ka-kn))*(exp((-kn*x)/U)-exp((-ka*x/U)))

    # eq for concentration
    conc = Cs*(1-a1)+Co*a1-Bo*a2-No*a3

    # eq for CBOD
    CBOD = Bo*exp((-kc*x)/U)

    # eq for NBOD
    NBOD = No*exp((-kn*x)/U)

    return conc, CBOD, NBOD
end
calc_conc (generic function with 1 method)

julia> # initialize for plotting
C = zeros(51);

julia> B = zeros(51);

julia> N = zeros(51);

julia> # find vals at x=0
init_DO = ((7.5*1000*100000)*(5*1000*10000))/(1000*100000+1000*10000)
7.2727272727272725

julia> init_CBOD = ((5*1000*100000)*(50*1000*10000))/(1000*100000+1000*10000)
```

9.090909090909092

```
julia> init_NBOD = ((5*1000*100000)+(35*1000*10000))/(1000*100000+1000*10000)
7.7272727272727275
```

```
julia> C[1] = init_DO;
```

```
julia> B[1] = init_CBOD;
```

```
julia> N[1] = init_NBOD;
```

```
julia> # from x=1 to x=15
      for i in 2:16
          C[i], B[i], N[i] = calc_conc(i-1, 6, 10, init_DO, init_CBOD,
init_NBOD, .55, .35, .25)
      end
```

```
julia> # from x=15 to x=50
      init_DO2 = ((C[16]*1000*110000)+(5*1000*15000))/(1000*110000+1000*15000)
4.896480650455226
```

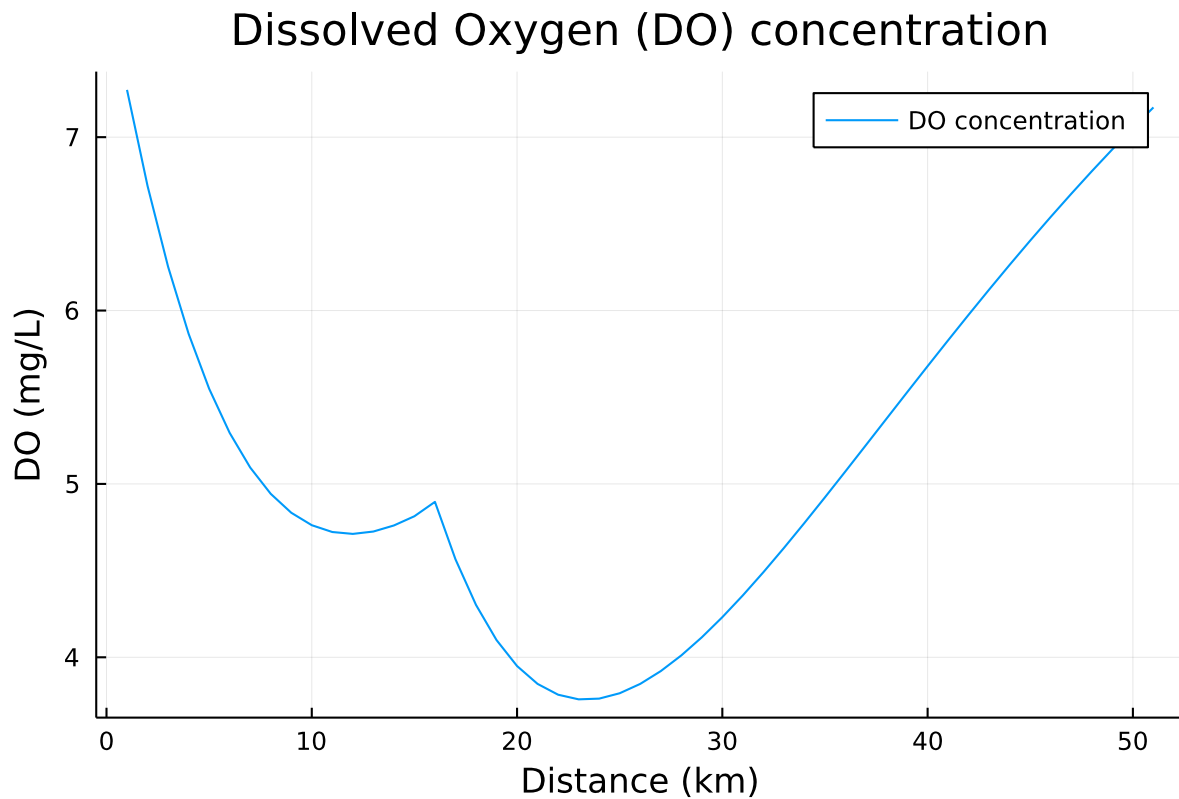
```
julia> init_CBOD2 =
((B[16]*1000*110000)+(45*1000*15000))/(1000*110000+1000*15000)
8.734896157428066
```

```
julia> init_NBOD2 =
((N[16]*1000*110000)+(35*1000*15000))/(1000*110000+1000*15000)
7.839777713929134
```

```
julia> for i in 16:51
          C[i], B[i], N[i] = calc_conc(i-16, 6, 10, init_DO2, init_CBOD2,
init_NBOD2, .55, .35, .25)
      end
```

```
julia> # plot DO
      using Plots
```

```
julia> plot(C, title= "Dissolved Oxygen (DO) concentration", label= "DO
concentration", xlabel="Distance (km)", ylabel= "DO (mg/L)")
```



Problem 1.2

```
julia> # initialize distance from waste stream 2
distfrom1=16
16

julia> # find distance where DO recovers to 6 mg/L
while C[distfrom1] < 6
    global distfrom1 += 1
end

julia> println(distfrom1)
43

julia> println(C[distfrom1])
6.121508813045605
```

The distance at which the stream recovers to a DO of 6 mg/L if both waste streams are untreated is around 43 m after the first waste stream, or around 27 m after the second waste stream.

Problem 1.3

```
julia> function find_treatment_min(U, Cs, Co, Bo, No, C1, B1, N1, C2, B2, N2,
ka, kc, kn, E1, E2)

    # find new treated vals for waste stream 1
    CBOD_treated1 = (1-(E1/100))*B1
```

```

NBOD_treated1 = (1-(E1/100))*N1

# find stream vals at x=0
initial_DO1 =
((Co*1000*100000)+(C1*1000*10000))/(1000*100000+1000*10000)
initial_CBOD1 =
((Bo*1000*100000)+(CBOD_treated1*1000*10000))/(1000*100000+1000*10000)
initial_NBOD1 =
((No*1000*100000)+(NBOD_treated1*1000*10000))/(1000*100000+1000*10000)

# find stream vals at x = 15
DO_15, CBOD_15, NBOD_15 = calc_conc(15, U, Cs, initial_DO1,
initial_CBOD1, initial_NBOD1, .55, .35, .25)

# find new treated vals for waste stream 2
CBOD_treated2 = (1-(E2/100))*B2
NBOD_treated2 = (1-(E2/100))*N2

# find stream vals after waste stream 2
initial_DO2 =
((DO_15*1000*110000)+(C2*1000*15000))/(1000*110000+1000*15000)
initial_CBOD2 =
((CBOD_15*1000*110000)+(CBOD_treated2*1000*15000))/(1000*110000+1000*15000)
initial_NBOD2 =
((NBOD_15*1000*110000)+(NBOD_treated2*1000*15000))/(1000*110000+1000*15000)

# find minimum for the treatment
Cmin = 1000
for i in 1:36
    newC, newB, newN = calc_conc(i-1, U, Cs, initial_DO2,
initial_CBOD2, initial_NBOD2, ka, kc, kn)
    if newC < Cmin
        Cmin = newC
    end
end
return Cmin

end

find_treatment_min (generic function with 1 method)

julia> # create vector of all possible treatments
treatment_plans = collect(0:100)
101-element Vector{Int64}:
 0
 1
 2
 3
 4
 5
 6
 7
 8
 9
⋮
92
93
94

```

```

95
96
97
98
99
100

julia> # initialize vector to store minimums
      mins = zeros(101);

julia> # iterate treatments
      for i in 1:101
          mins[i] = find_treatment_min(6, 10, 7.5, 5, 5, 5, 50, 35, 5, 45, 35,
.55, .35, .25, 0, treatment_plans[i])
      end

julia> # minimum %removal that has DO always less than 4 mg/L
      tr = 1
1

julia> while mins[tr] < 4
      global tr += 1
      end

julia> println(tr-1)
12

julia> println(mins[tr])
4.007015974604083

```

The minimum treatment plan that guarantees all DO values are above 4 mg/L is waste stream 2 having a treatment of 12%.

Problem 1.4

```

julia> # create vector of all possible treatments
      treatment_plans = collect(0:100)
101-element Vector{Int64}:
 0
 1
 2
 3
 4
 5
 6
 7
 8
 9
 ⋮
92
93
94
95
96
97

```

```

98
99
100

julia> # initialize vector to store minimums
      mins2 = zeros(101);

julia> # iterate treatments, each stream is treated equally
      for i in 1:101
          mins2[i] = find_treatment_min(6, 10, 7.5, 5, 5, 5, 50, 35, 5, 45, 35,
          .55, .35, .25, treatment_plans[i], treatment_plans[i])
      end

julia> # minimum %removal that has DO always less than 4 mg/L
      tr2 = 1

1

julia> while mins2[tr2] < 4
      global tr2 +=1
      end

julia> println(tr2-1)
7

julia> println(mins2[tr2])
4.017179459831182

```

If the streams are treated equally, the treatment value that keeps the DO above 4 mg/L is 7% for each stream.

Problem 1.5

In order to make a decision about the two treatment plans above, cost should be considered, as well as the plants' willingness to participate in a treatment plan. The first plan, only requiring waste stream 2 to be treated, is not fair to the second plant as the first plant does not need to treat their waste. However, treating both waste streams would likely be more expensive, or it would be tough getting both plants to agree to the treatment plan. Realistically, it would be best to choose the cheaper option despite the unfairness of the situation. In a true setting this choice would not be that simple, and other factors would be necessary to consider. However with the information we are given and my general knowledge of the engineer/client relationship, the cheapest option, where only waste stream 2 is treated, would be my choice.

Problem 1.6

```

julia> using Distributions

julia> # initialize vector to store probabilities
      probs = zeros(100);

julia> # iterate to find probabilities
      for j in 1:100

```

```

# initialize
global mins3 = zeros(100);
global fails_regulation = zeros(0);
global CBODs = zeros(100);
global NBODs = zeros(100);

# create vectors for CBOD, NBOD that have uniform dist
for i in 1:100
    CBODs[i] = rand(Uniform(4, 7));
    NBODs[i] = rand(Uniform(3, 8));
end

# find the treatment min for each scenario
for i in 1:100
    mins3[i] = find_treatment_min(6, 10, 7.5, CBODs[i], NBODs[i], 5,
50, 35, 5, 45, 35, .55, .35, .25, 0, 12);
end

# find how many fail the regulation
for i in 1:100
    if mins3[i] < 4
        append!(fails_regulation, mins3[i]);
    end
end

# calc probability
probs[j] = (length(fails_regulation))/(length(mins3))

end

julia> # take average probability
avg_prob = (sum(probs))/(length(probs))
0.7031000000000001

julia> println(avg_prob)
0.7031000000000001

```

The average probability that the treatment does not meet the requirement of DO above 4 mg/L is around 70%.

Problem 1.7

```

julia> # initialize vector to store probabilities
probs = zeros(100);

julia> # iterate to find probabilities
for j in 1:100

    # initialize
    global mins4 = zeros(100);
    global fails_regulation = zeros(0);
    global CBODs = zeros(100);
    global NBODs = zeros(100);

    # create sample vector from function

```

```

sample_CBOD_NBOD = sample_correlated_uniform(100, [4,7], [3,8]);

# create vectors for CBOD, NBOD that have uniform dist
for i in 1:100
    CBODs[i] = sample_CBOD_NBOD[i,1];
    NBODs[i] = sample_CBOD_NBOD[i,2];
end

# find the treatment min for each scenario
for i in 1:100
    mins4[i] = find_treatment_min(6, 10, 7.5, CBODs[i], NBODs[i], 5,
50, 35, 5, 45, 35, .55, .35, .25, 0, 12);
end

# find how many fail the regulation
for i in 1:100
    if mins4[i] < 4
        append!(fails_regulation, mins4[i]);
    end
end

# calc probability
probs[j] = (length(fails_regulation))/(length(mins3))

end

julia> # take average probability
avg_prob = (sum(probs))/(length(probs))
0.6478

julia> println(avg_prob)
0.6478

```

The average probability that the treatment does not meet the requirement of DO above 4 mg/L is around 64%.

Problem 1.8

The uncertainty here must be taken into account. The uncorrelated values while following the previously discussed treatment plan in 1.5 resulted in an uncertainty of 70%. This was slightly lower at 64% when CBOD and NBOD were related. Both of these values have an unacceptable percent of scenarios that fail to reach the regulation. Cost should be considered as a guide for increasing the treatment efficiencies to a certain, feasible point. Additionally, even if reasonable treatment plan is found that minimizes the probability of failure as well as minimizes cost, other environmental factors could change these predictions and could be added to refine the model.

'''

References

Julia: append to an empty vector. Stack Overflow. <https://stackoverflow.com/questions/28524105/append-to-an-empty-vector>

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Repeated Evaluation: Loops. Control Flow. Julia Documentation. <https://docs.julialang.org/en/v1flow/#man-loops>

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