Final Project Report

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1 Introduction

1.1 Problem Statement

So far, our simulations have considered the Net Present Value of home elevation under the assumption that homeowners will pay for their home elevation out of pocket. In reality, people have to chose between paying out of pocket (if this is even something they can afford), taking out a loan, or saving up.

Each of these options has tradeoffs in total cost, yearly cost, the cost of damages, and in NPV. In this paper, we will analyze these tradeoffs.

1.2 Selected Feature

Describe the feature you have selected to add to the existing decision-support tool. Discuss how this feature relates to the problem statement and its potential to improve climate risk assessment.

2 Literature Review

Provide a brief overview of the theoretical background related to your chosen feature. Cite at least two relevant journal articles to support your approach (see Quarto docs for help with citations). Explain how these articles contribute to the justification of your selected feature.

3 Methodology

3.1 Implementation

```
import Pkg
Pkg.status() # Note the paths for your local packages
Pkg.gc() # Garbage collect and delete cached package files
```

```
Status `C:\Users\Jonah\OneDrive\Documents\Rice PhD\Climate Damage Mitigation\final-project-js3:
[336ed68f] CSV v0.10.13
[a93c6f00] DataFrames v1.6.1
[1313f7d8] DataFramesMeta v0.15.2
[31c24e10] Distributions v0.25.107
```

[49e40289] HouseElevation v0.1.0 `HouseElevation`

```
[7073ff75] IJulia v1.24.2
     [b964fa9f] LaTeXStrings v1.3.1
     [bcdb8e00] Metaheuristics v3.3.5
     [b6ca2b7d] ParkingGarage v0.1.0 `ParkingGarage`
     [91a5bcdd] Plots v1.40.2
     [295af30f] Revise v3.5.14
     [2913bbd2] StatsBase v0.34.3
     [1986cc42] Unitful v1.19.0
     [9a3f8284] Random
   Info Packages marked with have new versions available and may be upgradable.
         Active manifest files: 10 found
         Active artifact files: 121 found
         Active scratchspaces: 2 found
        Deleted no artifacts, repos, packages or scratchspaces
  using CSV
  using DataFrames
  using DataFramesMeta
  using Distributions
4
  using LaTeXStrings
  using Metaheuristics
  using Plots
  using Random
  using Unitful
9
10
  Plots.default(; margin=5Plots.mm)
11
  using Revise
1
  using HouseElevation
  #include("Finance.jl")
  #using house
```

We put in our house in galveston

Choosing Galveston Pier 21, Texas The guage is at 29° 18.6 N, 94° 47.6 W https://maps.app.goo.gl/GyanSMA2fp9r

Our building is 302 17th St, Galveston, TX 77550, Home area as estimated by google maps: 30ftx50ft home = $1500ft^2$ Home value from zillow: 247,700 (Round up to 250,000)

The home is 4.41 feet or 1.34 meters above sea level in elevation. Looking at it on street view, the house appears to be on concrete blocks about 6 inches tall, giving it an effective height of 4.91 feet. Round this up to 5 so that it works.

Row 98 from the data is two-story, no basement in Galveston, so we'll be using that for our depth-damage curve. The home is on concrete blocks, so we can be confident that it doesn't have a basement.

```
house = let
haz_fl_dept = CSV.read("data/haz_fl_dept.csv", DataFrame) # read in the file
desc = "Two-story, no basement in Galveston"
row = @rsubset(haz_fl_dept, :Column1 == 98)[1, :,] # select the row I want
```

```
area = 1500u"ft^2"
       height_above_gauge = height_above_gauge = 5u"ft"
6
       House(row; area=area, height_above_gauge=height_above_gauge, value_usd=250_000)
   end
   slr scenarios = let
       df = CSV.read("data/slr oddo.csv", DataFrame)
2
        [Oddo17SLR(a, b, c, tstar, cstar) for (a, b, c, tstar, cstar) in eachrow(df)]
3
   end
4
5
   function draw_surge_distribution()
         = rand(Normal(5, 1))
         = rand(Exponential(1.25))
8
         = rand(Normal(0.1, 0.05))
9
       return GeneralizedExtremeValue( , , )
10
   end
11
12
   function draw_discount_rate()
13
       return rand(Normal(0.05, 0.03))
14
   end
15
16
   function draw_sow()
17
       slr = rand(slr scenarios)
18
       surge_params = draw_surge_distribution()
19
       discount = draw_discount_rate()
20
       return SOW(slr, surge_params, discount)
21
   end
```

Generate our possible states of the world

```
Random.seed!(421521)
N_SOW = 10
N_SOW_opt = 50 # to start
sows = [draw_sow() for _ in 1:N_SOW]
sows_opt = first(sows, N_SOW_opt)
```

```
10-element Vector{SOW{Float64}}:
SOW{Float64}(Oddo17SLR{Float64}(58.66628415, 2.709557857, 0.003600726, 2061.196955, 24.785578, SOW{Float64}(Oddo17SLR{Float64}(18.98238754, 2.076689405, 0.002459112, 2017.216294, 15.584015; SOW{Float64}(Oddo17SLR{Float64}(13.54334493, 1.576769557, -0.001892038, 2019.630551, 13.56306; SOW{Float64}(Oddo17SLR{Float64}(38.14589083, 2.623948719, 0.005085519, 2016.508542, 8.1346290; SOW{Float64}(Oddo17SLR{Float64}(40.9256403, 2.858600773, 0.008671168, 2024.544892, 18.7522310; SOW{Float64}(Oddo17SLR{Float64}(36.12774556, 2.258399655, 0.002399946, 2066.78925, 19.7923147; SOW{Float64}(Oddo17SLR{Float64}(54.44686357, 2.440676913, 0.003209864, 2039.323894, 22.8970515; SOW{Float64}(Oddo17SLR{Float64}(42.54886307, 2.374354031, 0.003786949, 2018.717192, 9.58709085; SOW{Float64}(Oddo17SLR{Float64}(5.951503134, 1.501882544, -0.002460307, 2022.058777, 10.789535; SOW{Float64}(Oddo17SLR{Float64}(17.66182918, 1.54415968, -0.002055399, 2058.500619, 31.6412585)
```

Write our function we'll use to evaluate financial plans

```
#"'{julia}

p = ModelParams(; house=house, years=2024:2083, finance=finance_basic) #make our model
elevations_try = 0:0.5:14 #establish actions to try actions_try = Action.(elevations_try)

#Get all the contruction costs we can afford construction_costs = [elevation_cost(house, elevation)
for elevation in elevations_try if elevation_cost(house, elevation) <= 170_000]
actions_try = actions_try[1:length(construction_costs)] #get the list of actions we can afford
#"'
```

3.2 Financial NPVS function

```
#assume physical house and SLR scenarios stay constant, only financial conditions change
   function financial_npvs(finance)
       p = ModelParams(; house=house, years=2024:2083, finance=finance) #make our model
3
       elevations_try = 0:0.5:14 #establish actions to try
5
       actions_try = Action.(elevations_try)
6
       #Get all the contruction costs we can afford
8
       construction_costs = [elevation_cost(house, elevation) for elevation in elevations_try if
9
10
       actions_try = actions_try[1:length(construction_costs)] #get the list of actions we can a
11
       elevations_try = elevations_try[1:length(construction_costs)] #limit elevations as well s
12
13
       #Run simulations
14
       npvs_opt = [mean([run_sim(a, sow, p) for sow in sows_opt]) for a in actions_try]
15
16
       #get height that minimizes NPV
17
       min_npv, min_idx = findmax(npvs_opt)
       minimizer = elevations_try[min_idx]
19
20
       #return elevations we tried (x), npvs(y) and minimizing elevation
21
       return elevations_try, npvs_opt, minimizer
22
23
   end
24
```

financial_npvs (generic function with 1 method)

Basic financing, just paying up fronts

println(methods(Finance)) println(fieldtypes(Finance)) ## Finance_Basic

```
finance_basic = let
loan = 0
loan_years = 0
loan_rate = 0.0
paid_off_percent = 1.0
amnt_paid_off = paid_off_percent * house.value_usd # Calculate amnt_paid_off
```

```
# Create Finance object using keyword arguments
Finance(;
loan = loan,
loan_years = loan_years,
loan_rate = loan_rate,
paid_off_percent = paid_off_percent,
amnt_paid_off = amnt_paid_off
)
end
```

Finance(0, 0, 0.0, 1.0, 250000.0)

Now lets calculate our basic finance options

```
elevations_tried, npvs, min = financial_npvs(finance_basic)
```

(0.0:0.5:14.0, [-1.0914604085744503e6, -1.0905425222436262e6, -953694.5075908549, -824837.0348)

3.3 plot_many

Define our function to plot many financial scenarios

```
function plot_many(finance_list, initial_plot)
3
       for finance in finance_list
4
5
            #println(finance)
6
            elevations_tried, npvs, min = financial_npvs(finance)
            if finance.loan == 1 #if taking out a loan
                label = "$(round(finance.loan_rate*100))% loan over $(finance.loan_years) years"
            elseif finance.loan > 1
10
                label = "Saving over $(finance.loan_years) years"
11
            else
12
                label = "Paying out of pocket"
13
            end
14
            plot!(
15
            initial_plot,
16
            elevations_tried,
17
            npvs ./ 1000;
18
            xlabel="Elevation [ft]",
19
            ylabel="NPV [1000 USD]",
20
            label=label,
21
            marker=:circle,
            #color=line_color
23
24
25
             #make the vertical line color the same as the horizontal line
26
            line_color = initial_plot.series_list[end][:linecolor]
27
```

```
#line_color = plot_object.series_list[end].plotseries[:linecolor]
vline!([min]; label="$(min)ft elevation", linestyle=:dash, color=line_color)
end
end
```

plot_many (generic function with 1 method)

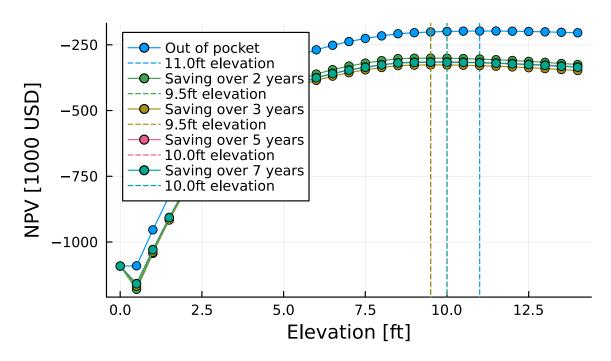
3.4 reasonable saving time

Generate many financial scenarios where we save up for different periods of time

Any[Finance(2, 2, 0.0, 1.0, 250000.0), Finance(2, 3, 0.0, 1.0, 250000.0), Finance(2, 5, 0.0, 1)
Plot our saving options

```
p = plot(
       elevations_tried,
2
       npvs ./ 1000;
3
       xlabel="Elevation [ft]",
4
       ylabel="NPV [1000 USD]",
5
       title="Out of Pocket Vs Various saving periods",
6
       label="Out of pocket",
       marker=:circle#, color=line_color
   )
9
10
11
   line_color = p.series_list[end][:linecolor] #make the vertical line color the same as the hor
12
   vline!([min]; label="$(min)ft elevation", linestyle=:dash, color=line_color)
13
   plot_many(finance_saving, p)
   display(p)
```

Out of Pocket Vs Various saving periods



Results of this graph show two interesting things:

- 1. NPV of saving is always a lot lower than paying up front. This makes sense, since we're increasing construction cost with inflation, so the only change to the NPV will be more damages while we wait to elevate the house.
- 2. NPV results recommend lower elevations if we chose to wait. This seems strange at first, since most people might save money so they can elevate higher, and you'd think that if we do wait to elevate, it'd make more sense to do a higher elevation to compensate for the damages we experienced in the first few years.

But since we are doing this simulation over a fixed period of time, the longer we wait to elevate, the less value is gained by elevating, since we'll have spent less of our time avoiding damages. So assuming a non-infinite time duration for this experiment, the longer we wait to elevate, the lower the recommended elevation height gets.

We can demonstrate this more clearly by displaying some extremely long saving periods.

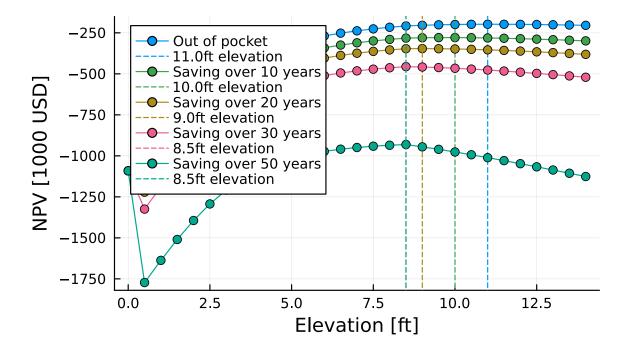
3.5 Long saving time

```
Long_Savings = []

for i in [10, 20, 30, 50]
    #print(i)
    paid_off_percent = 1.0
    fin = Finance(; loan=2, loan_years=i, loan_rate=0.0, paid_off_percent=paid_off_percent,
        amnt paid off=(house.value usd*paid off percent)
```

```
push!(Long_Savings, fin)
9
   end
10
   println(Long_Savings)
   Any[Finance(2, 10, 0.0, 1.0, 250000.0), Finance(2, 20, 0.0, 1.0, 250000.0), Finance(2, 30, 0.0
   p = plot(
       elevations_tried,
       npvs ./ 1000;
3
       xlabel="Elevation [ft]",
       ylabel="NPV [1000 USD]",
       title="Out of Pocket Vs Various saving periods",
       label="Out of pocket",
       marker=:circle#, color=line color
9
10
11
   line_color = p.series_list[end][:linecolor] #make the vertical line color the same as the hor
12
   vline!([min]; label="$(min)ft elevation", linestyle=:dash, color=line_color)
13
14
   plot_many(Long_Savings, p)
15
   display(p)
```

Out of Pocket Vs Various saving periods



3.6 Different loan durations

Generate many financial scenarios with different years to pay off the loan at a 7% interest rate

```
finance_list_years = []

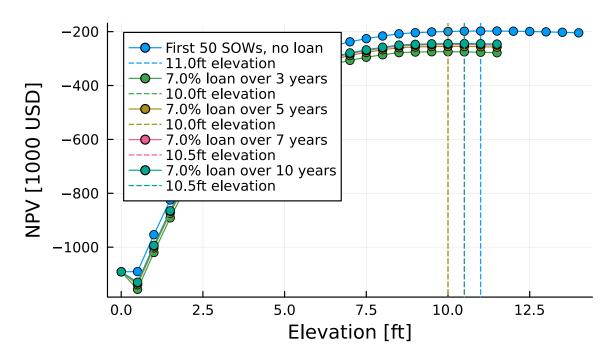
for i in [3, 5, 7, 10]
    #print(i)
    paid_off_percent = 0.7
    fin = Finance(; loan=1, loan_years=i, loan_rate=0.07, paid_off_percent=paid_off_percent,
        amnt_paid_off=(house.value_usd*paid_off_percent)
        )
    push!(finance_list_years, fin)
end

println(finance_list_years)
```

Any [Finance(1, 3, 0.07, 0.7, 175000.0), Finance(1, 5, 0.07, 0.7, 175000.0), Finance(1, 7, 0.07). Plot our scenarios of different years to pay off the loan

```
p = plot(
       elevations_tried,
       npvs ./ 1000;
3
       xlabel="Elevation [ft]",
4
       ylabel="NPV [1000 USD]",
5
       title="Out of Pocket Vs Various payment periods",
6
       label="First $(N_SOW_opt) SOWs, no loan",
       marker=:circle#, color=line_color
   )
10
11
   line_color = p.series_list[end][:linecolor] #make the vertical line color the same as the hor
12
   vline!([min]; label="$(min)ft elevation", linestyle=:dash, color=line_color)
13
14
   plot_many(finance_list_years, p)
   display(p)
```

Out of Pocket Vs Various payment periods



3.7 Different interest rates

Generate many financial scenarios with different interest rates

```
finance_list_rates = []
   for i in range(start=3, stop=9, length=4)
3
       #print(i)
4
       paid_off_percent = 0.7
5
       fin = Finance(; loan=1, loan_years=5, loan_rate=i/100, paid_off_percent=paid_off_percent,
6
           amnt_paid_off=(house.value_usd*paid_off_percent)
       push!(finance_list_rates, fin)
   end
10
11
   println(finance_list_rates)
12
```

Any[Finance(1, 5, 0.03, 0.7, 175000.0), Finance(1, 5, 0.05, 0.7, 175000.0), Finance(1, 5, 0.07

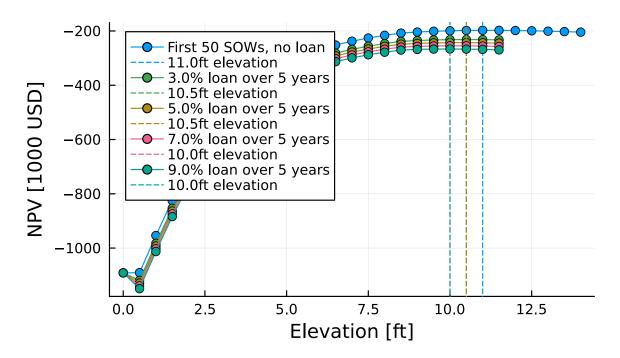
#colors = [colorant"red", colorant"green", colorant"blue", colorant"orange", colorant"purple"]

#line_color = colors[1]

p = plot(
 elevations_tried,
 npvs ./ 1000;

```
xlabel="Elevation [ft]",
8
       ylabel="NPV [1000 USD]",
9
       title="Out of Pocket Vs Various 5 year loans, 70% paid off",
10
       label="First $(N_SOW_opt) SOWs, no loan",
       marker=:circle#, color=line color
12
13
   #println(p.series_list[end][:linecolor] )
14
15
   #colors = Plots.palette()
16
17
   line_color = p.series_list[end][:linecolor] #make the vertical line color the same as the hor
   vline!([min]; label="$(min)ft elevation", linestyle=:dash, color=line_color)
19
   plot_many(finance_list_rates, p)
   display(p)
```

Out of Pocket Vs Various 5 year loans, 70% paid



3.8 Compare NPVS and annual costs

Now we want to compare NPVS and annual costs for all of these scenarios

Make a function to get NPV, annual cost, and optimal elevatin

```
function annual_loan_cost(p, r, n)

#P is principle amount, r is rate, n is number of years

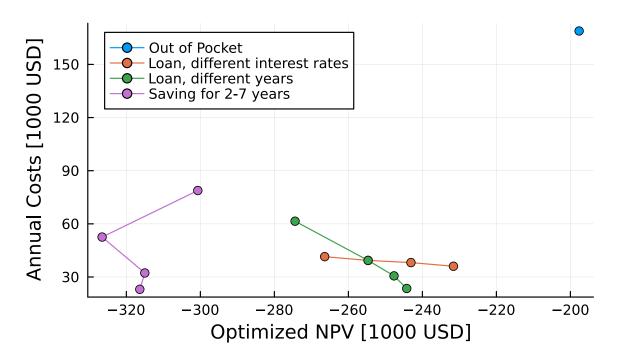
a = p * ( r * ((1+r)^n) / ((1+r)^n - 1))

return a #a is annual payments
```

```
end
6
   function annual_v_npv(finance)
9
       elevations_tried, npvs, best_elevation = financial_npvs(finance)
10
       #An engineer is not someone who makes things maximally efficient, but someone who makes th
11
       best_npv, best_idx = findmax(npvs)
12
13
       construction cost = elevation cost(house, best elevation)
14
       #println(elevation_cost(house, best_elevation))
15
16
       if finance.loan == 0 #out of pocket
17
            annual_cost = construction_cost
18
       elseif finance.loan == 1 #taking out a loan
19
            annual_cost = annual_loan_cost(construction_cost, finance.loan_rate, finance.loan_year
21
       else #saving up
            annual_cost = construction_cost/finance.loan_years
22
       end
23
24
       return best_npv, annual_cost, best_elevation
25
26
   end
27
   annual_v_npv (generic function with 1 method)
   function get_costs(finances)
       figures = annual_v_npv.(finances)
2
       best_npvs = [fig[1] for fig in figures]
3
       annual_costs = [fig[2] for fig in figures]
4
       best_elevations = [fig[3] for fig in figures]
5
6
       return best_npvs, annual_costs, best_elevations
   end
8
9
10
   function plot_money_values(finance, initial_plot, label)
11
12
       best_npvs, annual_costs, best_elevations = get_costs(finance)
13
       plot!(
14
            initial_plot,
15
            best_npvs ./ 1000,
16
            annual_costs ./ 1000;
17
            xlabel="Optimized NPV [1000 USD]",
           ylabel="Annual Costs [1000 USD]",
19
            label=label,
20
           marker=:circle,
21
            #color=line_color
22
```

```
23
24
   end
   plot_money_values (generic function with 1 method)
   get costs(finance list rates)
   optimal_npv, annual_cost, best_elevation = annual_v_npv(finance_basic)
   Annual_v_npv_plot = plot(
3
        [optimal_npv] ./ 1000,
4
        [annual_cost] ./ 1000;
5
       xlabel="Optimized NPV [1000 USD]",
6
       ylabel="Annual Costs [1000 USD]",
       title="NPV vs Yearly Costs",
       label="Out of Pocket",
       marker=:circle#, color=line color
10
11
   plot_money_values(finance_list_rates, Annual_v_npv_plot, "Loan, different interest rates")
12
   plot_money_values(finance_list_years, Annual_v_npv_plot, "Loan, different years")
13
   plot_money_values(finance_saving, Annual_v_npv_plot, "Saving for 2-7 years")
   display(Annual_v_npv_plot)
```

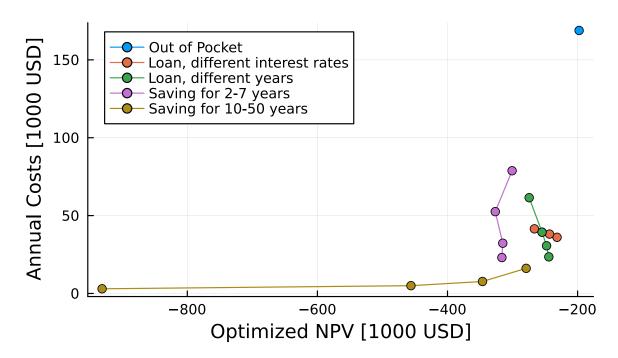
NPV vs Yearly Costs



The chart changes radically when we display long savings times

plot_money_values(Long_Savings, Annual_v_npv_plot, "Saving for 10-50 years")
display(Annual_v_npv_plot)

NPV vs Yearly Costs



3.9 compare NPVS and construction heights for all these different scenarios

Now we want to compare NPVS and construction heights for all these different scenarios

3.10 Next: add indicators to annual_v_npv plot that shows what thing is what

You should make your modifications in either the HouseElevation or ParkingGarage module. Detail the steps taken to implement the selected feature and integrate it into the decision-support tool. Include code snippets and explanations where necessary to clarify the implementation process.

3.11 Validation

In this case, we don't really *need* optimization – we can use brute force. We can compare by plotting the objective function for a range of elevations (from 0 to 14 ft) using all SOWs.

As we have seen in labs, mistakes are inevitable and can lead to misleading results. To minimize the risk of errors making their way into final results, it is essential to validate the implemented feature. Describe the validation techniques used to ensure the accuracy and reliability of your implemented feature. Discuss any challenges faced during the validation process and how they were addressed.

4 Results

Present the results obtained from the enhanced decision-support tool. Use tables, figures, and visualizations to clearly communicate the outcomes. Provide sufficient detail to demonstrate how the implemented feature addresses the problem statement. Use the #| output: false and/or #| echo: false tags to hide code output and code cells in the final report except where showing the output (e.g.g, a plot) or the code (e.g., how you are sampling SOWs) adds value to the discussion. You may have multiple subsections of results, which you can create using ##.

5 Conclusions

5.1 Discussion

Analyze the implications of your results for climate risk management. Consider the context of the class themes and discuss how your findings contribute to the understanding of climate risk assessment. Identify any limitations of your approach and suggest potential improvements for future work.

5.2 Conclusions

Summarize the key findings of your project and reiterate the significance of your implemented feature in addressing the problem statement. Discuss the broader implications of your work for climate risk management and the potential for further research in this area.

6 References