

# Team Homework 2 - MATH 420

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## NYC Covid Data

```
In [1]: # activate env and instantiate pkgs
using Pkg
Pkg.activate("../p1")
Pkg.instantiate()
```

Activating project at `~/m420p1/p1`

```
In [2]: using CSV, DataFrames, Dates, LaTeXStrings, Plots, Distances, OrderedCollect
```

```
In [3]: df = CSV.read("../resources/data.csv", DataFrame) # entire df

v_infected = values(df[1, 13:end]) # vector of infected numbers
infected_dates = names(df[1, 13:end]) # vector of infected_dates
@assert length(v_infected) == length(infected_dates)
```

## Exercise 1

1)

```
In [4]: yd = values(df[2, 13:end]);
```

```
In [5]: v(t::Int)::Int = v_infected[t]
y(t::Int)::Int = yd[t];
```

```
In [6]: findfirst([x>=5 for x in v_infected])
```

Out[6]: 46

```
In [7]: y_deaths = values(df[2, 13+46-1:end]); # vector of deaths
```

$T_0 = 46$

```
In [8]: function I(t::Int, τ::Int=7)::Real
    t0 = 46
    return v(t+t0+τ) - v(t+t0-τ)
end
```

Out[8]: I (generic function with 2 methods)

```
In [9]: popu=df.Population[1];  
tmax = 119;
```

```
In [10]: # df = CSV.read("../resources/data.csv", DataFrame) # entire df  
  
# read the values of infections and deaths from the table  
infected = values(df[1, 13:end]) # vector of infected numbers  
deaths = values(df[2, 13:end]) # vector of death numbers  
infected_dates = names(df[1, 13:end]) # vector of infected_dates  
population = df.Population[1];  
@assert length(infected) == length(deaths)  
# parameters  
Vmin = 5  
 $\tau_0 = 7$   
 $\lambda = 1$   
  
t0 = findfirst([x>=Vmin for x in infected])  
tmax = 119  
# define V, Y, and I for the range of times we are interested in  
V(t) = infected[t + t0]  
Y(t) = deaths[t + t0]  
I(t) = infected[t + t0 +  $\tau_0$ ] - infected[t + t0 -  $\tau_0$ ]
```

Out[10]: I (generic function with 3 methods)

```
In [11]: # function to find the optimal gamma for a given R_sim and p  
function optGamma(R_sim, p)  
    if p == 2  
        s1(t) = Y(t)*R_sim[t+1]  
        s2(t) = abs(R_sim[t+1])^2  
        gamma_c = sum(s1, 0:tmax)/sum(s2, 0:tmax)  
        if gamma_c < 0  
            return 0  
        elseif gamma_c > 1  
            return 1  
        else  
            return gamma_c  
        end  
    elseif p == 1  
        # min = Inf  
        # minimizer = -1  
        md = Dict()  
        for k in 0:tmax  
            (R_sim[k+1]==0) && continue  
            r = Y(k)/R_sim[k+1]  
            if r >= 0 && r <= 1  
  
                s3(t) = abs(Y(t) - r*R_sim[t+1])  
                #f = sum(s3, 0:tmax)  
                md[r] = Cityblock()(Y.(0:tmax), r*R_sim) #sum(s3, 0:tmax)  
                # if f < min  
                #     min = f  
                #     minimizer = r  
                # end  
            end  
        end  
    end
```

```

        end

        return argmin(md)
    else
        model = Model()
        set_optimizer(model, GLPK.Optimizer)

        @variable(model, x1 >= 0)
        @variable(model, 1 >= x2 >= 0)

        for t = 0:tmax
            @constraint(model, (-x1) - R_sim[t+1]*x2 <= -Y(t))
            @constraint(model, (-x1) + R_sim[t+1]*x2 <= Y(t))
        end

        @objective(model, Min, x1)
        optimize!(model)

        return value.(x2)
    end
end
end

```

Out[11]: optGamma (generic function with 1 method)

```

In [12]: # Euler scheme
function euler(alpha, beta, N)::NTuple{3, Vector{Float64}}
    h = 0.01
    n = 2

    S_sim = Vector{Float64}(undef, 120)
    I_sim = Vector{Float64}(undef, 120)
    R_sim = Vector{Float64}(undef, 120)

    s = N
    i = I(0)
    r = 0

    t = 0

    S_sim[1] = s
    I_sim[1] = i
    R_sim[1] = r
    while n <= 120
        ds = -beta*s*(i/N)
        di = beta*s*(i/N) - alpha*i
        dr = alpha*i

        s += h*ds
        i += h*di
        r += h*dr
        t += h
        rt = round(t, digits=2);
        if isinteger(rt)
            S_sim[n] = s
            I_sim[n] = i

```

```

        R_sim[n] = r
        n += 1
    end
end

    return S_sim, I_sim, R_sim
end

```

Out[12]: euler (generic function with 1 method)

In [13]: Y1 = first(y\_deaths, 120);

In [14]: **function** J1( $\gamma$ , Is, Rs, p)::**Real**

```

     $\lambda$  = 1
    f1(t) = abs(I(t) - Is[t+1])^p
    f2(t) = abs(y_deaths[t+1] -  $\gamma$ *Rs[t+1])^p
    return sum(f1, 0:119) +  $\lambda$ *sum(f2, 0:119)
end

```

Out[14]: J1 (generic function with 1 method)

$$p = 1$$

In [15]: Jd = OrderedDict()

Out[15]: OrderedDict{Any, Any}()

In [16]: **for**  $\alpha$  **in** ProgressBar(0.05:0.01:0.2)

```

    for r0 in 1.5:0.1:1.9
        for nn in 2:0.5:10
             $\beta$  = r0* $\alpha$ 
            N = popu*nn/100
            Ss, Is, Rs = euler( $\alpha$ ,  $\beta$ , N)
            #ds = OrderedDict()
            #for  $\gamma$  in 0:0.0001:1
            #    ds[ $\gamma$ ] = Cityblock()(Y1,  $\gamma$ *Rs)
            #end
            # $\hat{\gamma}$  = argmin(ds)
             $\hat{\gamma}$  = optGamma(Rs, 1)
            Jd[( $\alpha$ ,  $\beta$ , r0, N,  $\hat{\gamma}$ )] = J1( $\hat{\gamma}$ , Is, Rs, 1)
        end
    end
end

```

0.0%	0/16 [00:00<00:-1, -0
s/it]	
6.2%	1/16 [00:04<Inf:Inf, InfG
s/it]	
12.5%	2/16 [00:07<01:38, 7
s/it]	
18.8%	3/16 [00:11<01:14, 6
s/it]	
25.0%	4/16 [00:17<01:06, 6
s/it]	
31.2%	5/16 [00:22<01:01, 6
s/it]	
37.5%	6/16 [00:28<00:57, 6
s/it]	
43.8%	7/16 [00:35<00:52, 6
s/it]	
50.0%	8/16 [00:41<00:47, 6
s/it]	
56.2%	9/16 [00:48<00:42, 6
s/it]	
62.5%	10/16 [00:54<00:36, 6
s/it]	
68.8%	11/16 [01:01<00:30, 6
s/it]	
75.0%	12/16 [01:07<00:24, 6
s/it]	
81.2%	13/16 [01:14<00:18, 6
s/it]	
87.5%	14/16 [01:20<00:12, 6
s/it]	
93.8%	15/16 [01:26<00:06, 6
s/it]	
100.0%	16/16 [01:33<00:00, 6
s/it]	
100.0%	16/16 [01:33<00:00, 6
s/it]	

```
In [17]:  $\hat{\alpha}, \hat{\beta}, r\hat{\theta}, \hat{N}, \hat{\gamma} = \text{argmin}(\text{Jd})$ 
         J_min = Jd[argmin(Jd)];
```

```
In [18]: display(L"\text{For} p=1")
         @show  $\hat{\alpha}, \hat{\beta}, r\hat{\theta}, \hat{N}, \hat{\gamma}$ ;
         display(L"J_{min}\approx %$(round(J_min, digits=3))")
```

For  $p = 1$

$(\hat{\alpha}, \hat{\beta}, r\hat{\theta}, \hat{N}, \hat{\gamma}) = (0.17, 0.323, 1.9, 48861.18, 0.0789682964249939)$

$J_{min} \approx 210762.063$

2)

```
In [19]: function Jplot( $\alpha, \beta, p$ )
         S, I, R = euler( $\alpha, \beta, \hat{N}$ )
```

```

return J1( $\hat{\gamma}$ , I, R, p)
end

```

Out[19]: Jplot (generic function with 1 method)

```

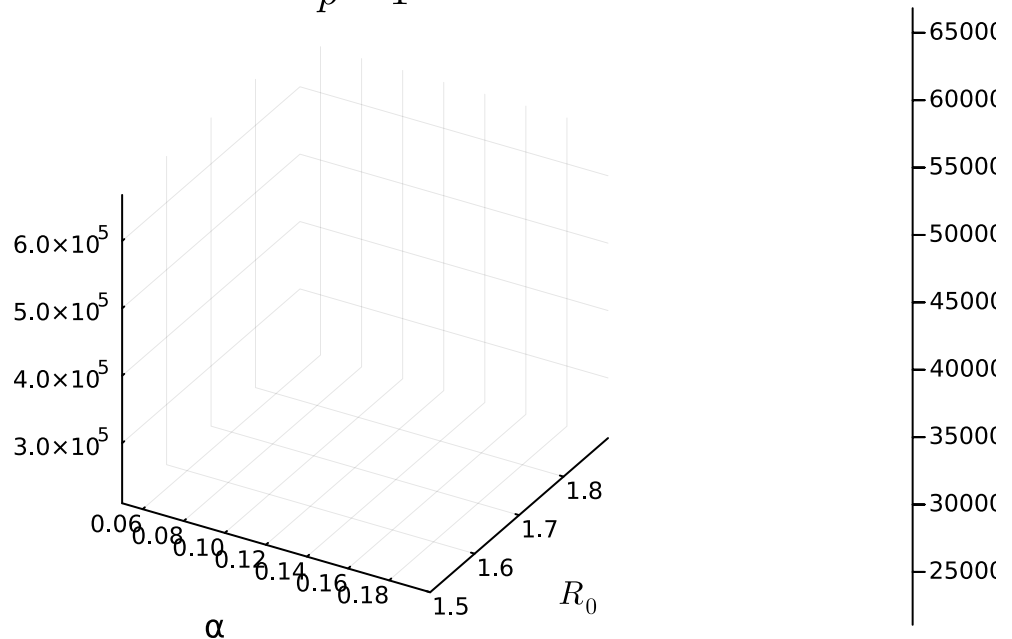
In [20]: surface(0.05:0.01:0.2, 1.5:0.01:1.9, (x,y) -> Jplot(x, x*y, 1))
surface!(title=L"J(\alpha, \beta, \hat{N}, \hat{\gamma})" * "\n" * L"p=1", xlabel=

```

Out[20]:

$$J(\alpha, \beta, \hat{N}, \hat{\gamma})$$

$$p = 1$$



3)

```

In [21]: Ss, Is, Rs = euler( $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{N}$ );

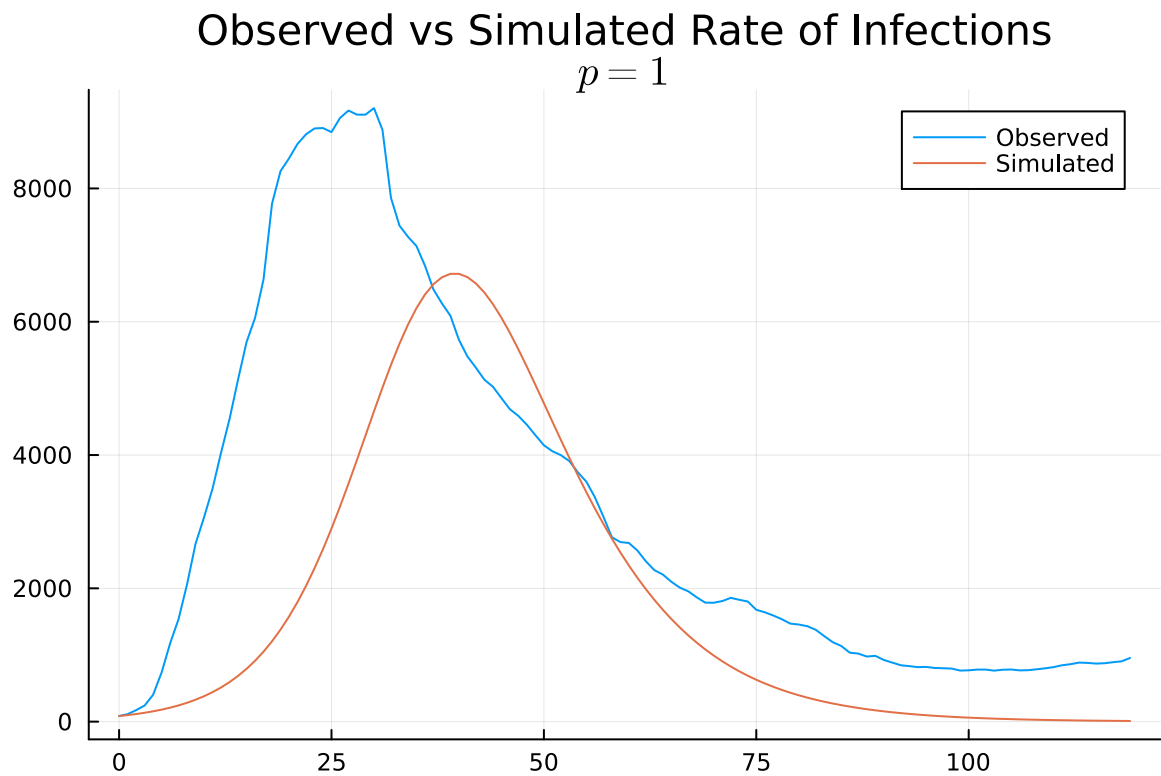
```

```

In [22]: plot(0:119, I, labels="Observed")
plot!(0:119, Is, label="Simulated")
plot!(title = "Observed vs Simulated Rate of Infections" * "\n" * L"p=1")

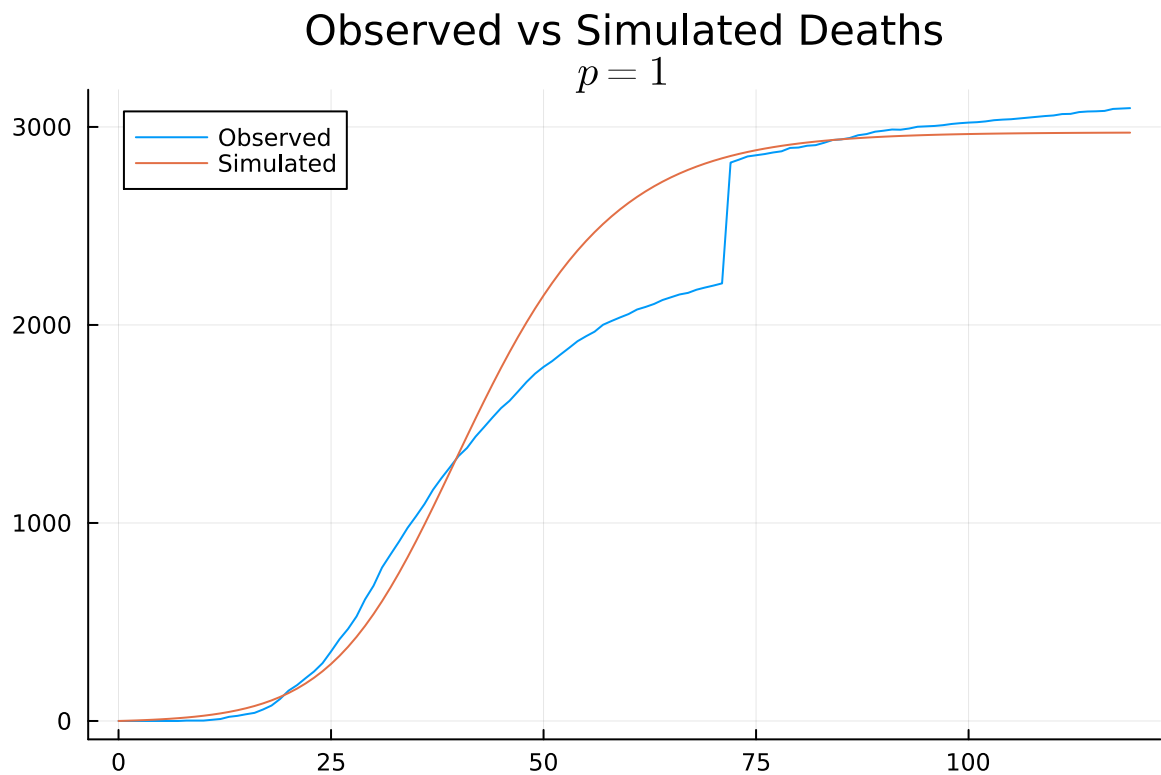
```

Out [22]:



```
In [23]: plot(0:119, Y1, labels="Observed")
plot!(0:119,  $\hat{y}$ *Rs, label="Simulated")
plot!(title = "Observed vs Simulated Deaths" * "\n" * L"p=1")
```

Out [23]:



$p = 2$

```
In [24]: Jd = OrderedDict()
for  $\alpha$  in 0.05:0.01:0.2
    for  $r_0$  in 1.5:0.1:1.9
        for nn in 2:0.5:10
             $\beta = r_0 * \alpha$ 
            N = popu*nn/100
            Ss, Is, Rs = euler( $\alpha$ ,  $\beta$ , N)
            #ds = OrderedDict()
            #for  $\gamma$  in 0:0.0001:1
            #    ds[ $\gamma$ ] = Euclidean()(Y1,  $\gamma * Rs$ )
            #end
            # $\hat{\gamma} = \text{argmin}(ds)$ 
             $\hat{\gamma} = \text{optGamma}(Rs, 2)$ 
            Jd[( $\alpha$ ,  $\beta$ ,  $r_0$ , N,  $\hat{\gamma}$ )] = J1( $\hat{\gamma}$ , Is, Rs, 2)
        end
    end
end
```

```
In [25]:  $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{r}_0$ ,  $\hat{N}$ ,  $\hat{\gamma} = \text{argmin}(Jd)$ 
J_min = Jd[argmin(Jd)];
```

```
In [26]: display(L"\text{For} p=2")
@show  $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{r}_0$ ,  $\hat{N}$ ,  $\hat{\gamma}$ ;
display(L"J_{min} \approx %$(round(J_min, digits=3))")
```

For  $p = 2$

$(\hat{\alpha}, \hat{\beta}, \hat{r}_0, \hat{N}, \hat{\gamma}) = (0.2, 0.38, 1.9, 65148.24, 0.053995703004223516)$

$J_{min} \approx 5.29978911123e8$

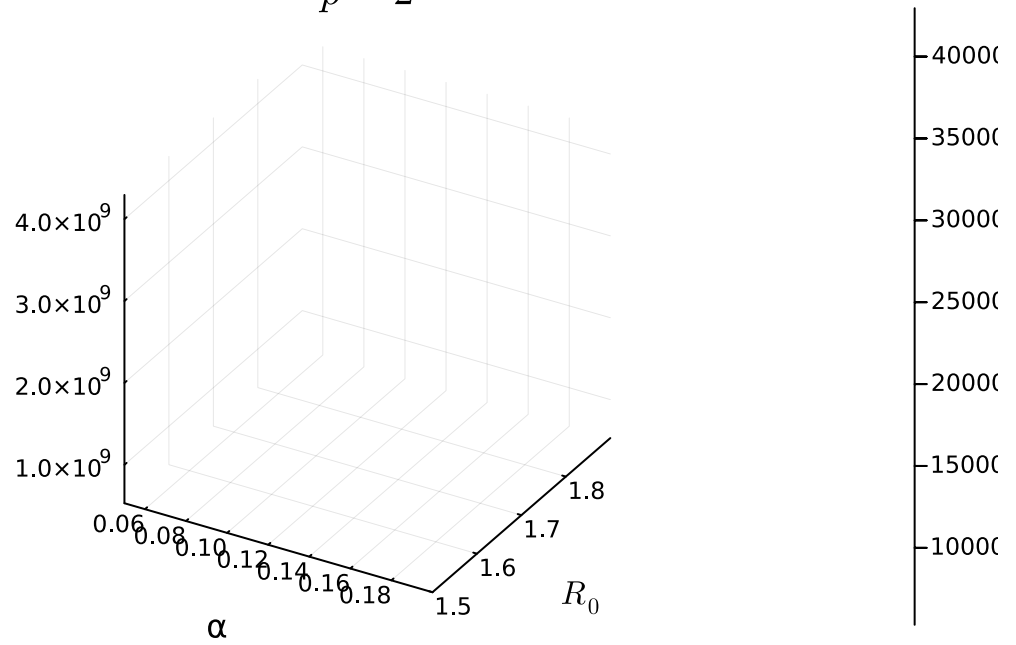
2)

```
In [27]: surface(0.05:0.01:0.2, 1.5:0.01:1.9, (x,y) -> Jplot(x, x*y, 2))
surface!(title=L"J(\alpha, \beta, \hat{N}, \hat{\gamma})"*"\n"*L"p=2", xlabel
```



Out [27]:

$$J(\alpha, \beta, \hat{N}, \hat{\gamma})$$
$$p = 2$$

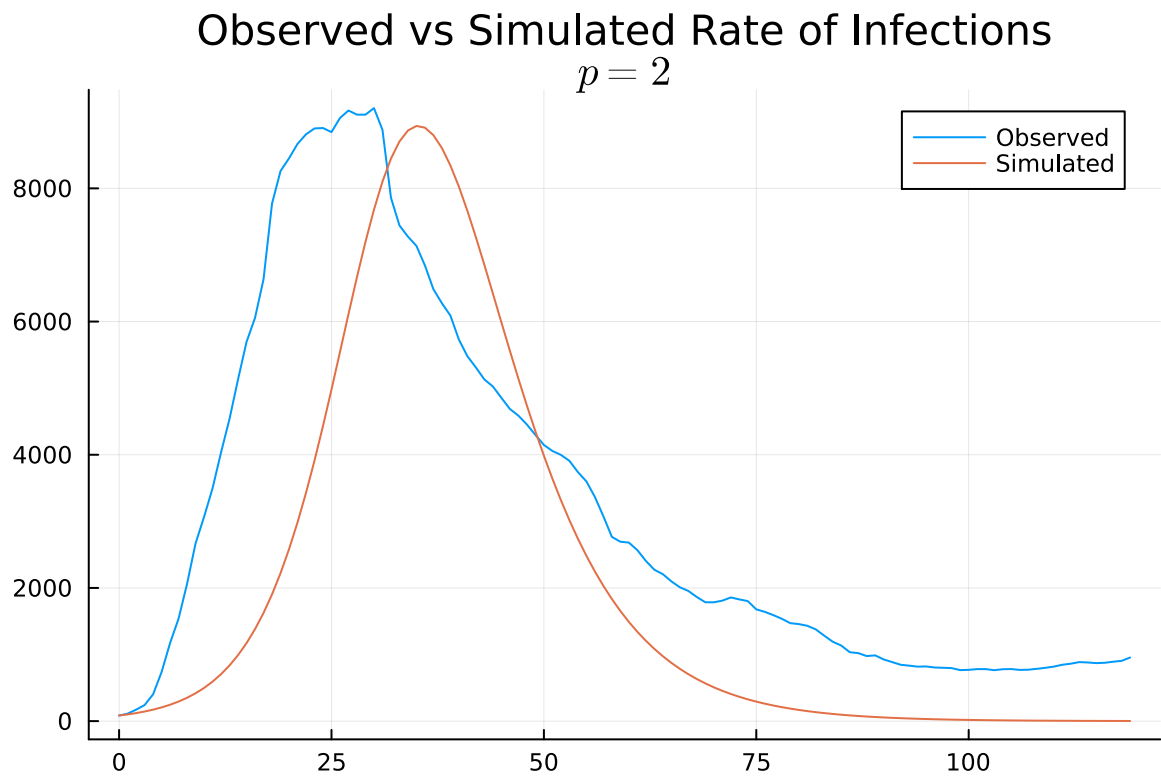


3)

```
In [28]: Ss, Is, Rs = euler( $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{N}$ );
```

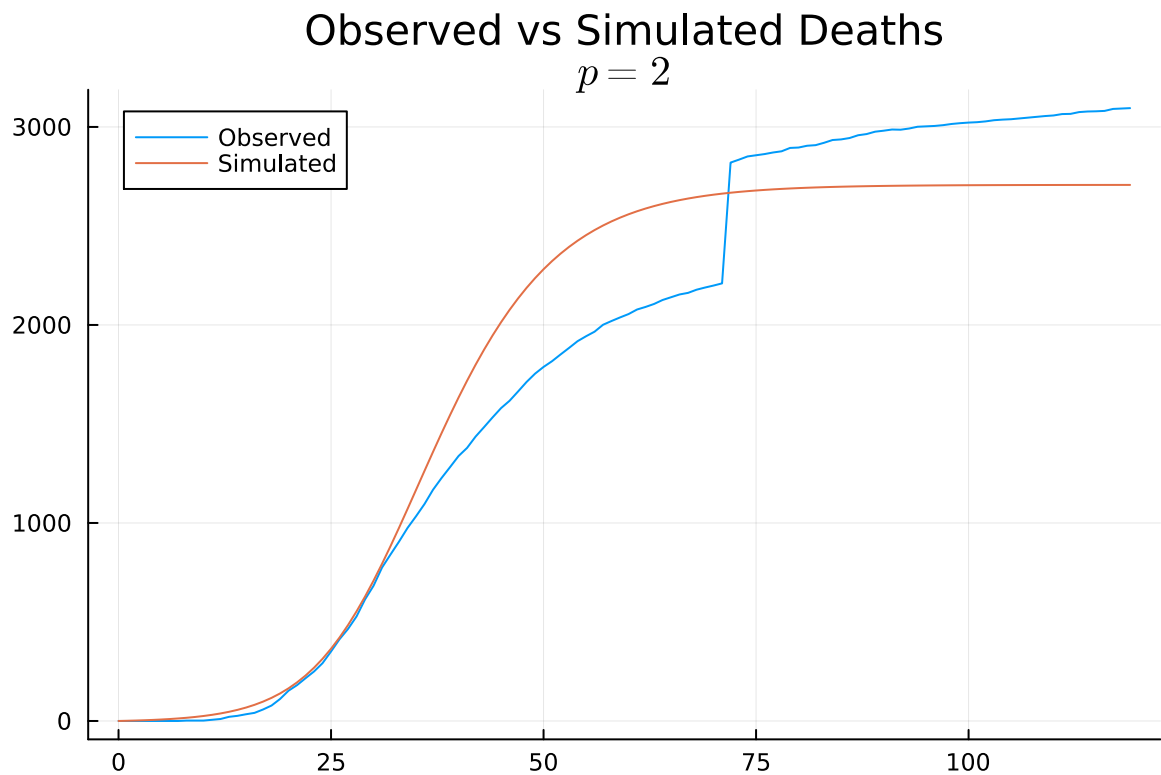
```
In [29]: plot(0:119, I, labels="Observed")
plot!(0:119, Is, label="Simulated")
plot!(title = "Observed vs Simulated Rate of Infections" * "\n" * L"p=2")
```

Out [29]:



```
In [30]: plot(0:119, Y1, labels="Observed")
plot!(0:119,  $\hat{y}$ *Rs, label="Simulated")
plot!(title = "Observed vs Simulated Deaths" * "\n" * L"p=2")
```

Out [30]:



$p = \infty$

```
In [31]: function J2( $\gamma$ , Is, Rs, p)::Real
     $\lambda$  = 1
    f1(t) = abs(I(t) - Is[t+1])
    f2(t) = abs(y_deaths[t+1] -  $\gamma$ *Rs[t+1])
    return maximum(f1, 0:119) +  $\lambda$ *maximum(f2, 0:119)
end
```

Out[31]: J2 (generic function with 1 method)

```
In [32]: Jd = OrderedDict()
for  $\alpha$  in 0.05:0.01:0.2
    for r0 in 1.5:0.1:1.9
        for nn in 2:0.5:10
             $\beta$  = r0* $\alpha$ 
            N = popu*nn/100
            Ss, Is, Rs = euler( $\alpha$ ,  $\beta$ , N)
            ds = OrderedDict()
            #for  $\gamma$  in 0:0.01:1
            #    ds[ $\gamma$ ] = maximum(abs.(Y1 -  $\gamma$ *Rs))
            #end
            # $\hat{\gamma}$  = argmin(ds)
             $\hat{\gamma}$  = optGamma(Rs, 3)
            Jd[( $\alpha$ ,  $\beta$ , r0, N,  $\hat{\gamma}$ )] = J1( $\hat{\gamma}$ , Is, Rs, 2)
        end
    end
end
```

```
In [33]:  $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{r}_0$ ,  $\hat{N}$ ,  $\hat{\gamma}$  = argmin(Jd)
J_min = Jd[argmin(Jd)];
```

```
In [34]: display(L"\text{For} p=\infty")
@show  $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{r}_0$ ,  $\hat{N}$ ,  $\hat{\gamma}$ ;
display(L"J_{min}\approx %$(round(J_min, digits=3))")
```

For  $p = \infty$   
 $(\hat{\alpha}, \hat{\beta}, \hat{r}_0, \hat{N}, \hat{\gamma}) = (0.2, 0.38, 1.9, 65148.24, 0.05269036383890261)$   
 $J_{min} \approx 5.30297452273e8$

```
In [35]: function Jplot2( $\alpha$ ,  $\beta$ , p)
    S, I, R = euler( $\alpha$ ,  $\beta$ ,  $\hat{N}$ )
    return J2( $\hat{\gamma}$ , I, R, p)
end
```

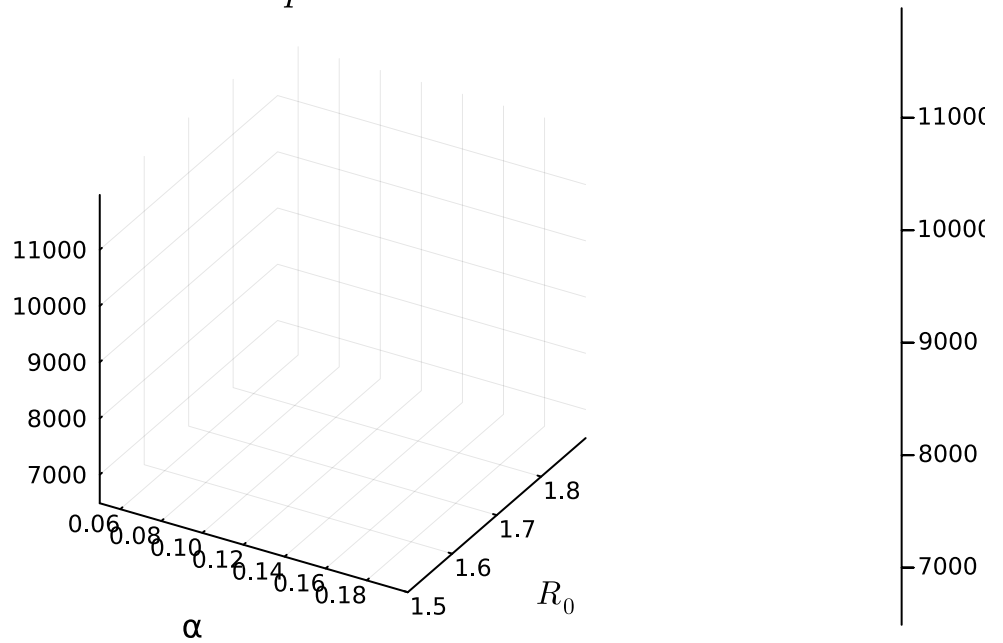
Out[35]: Jplot2 (generic function with 1 method)

2)

```
In [36]: surface(0.05:0.01:0.2, 1.5:0.01:1.9, (x,y) -> Jplot2(x, x*y, 2))
surface!(title=L"J(\alpha, \beta, \hat{N}, \hat{\gamma})"*"\n"*L"p=\infty", xlabel
```

Out [36]:

$$J(\alpha, \beta, \hat{N}, \hat{\gamma})$$
$$p = \infty$$

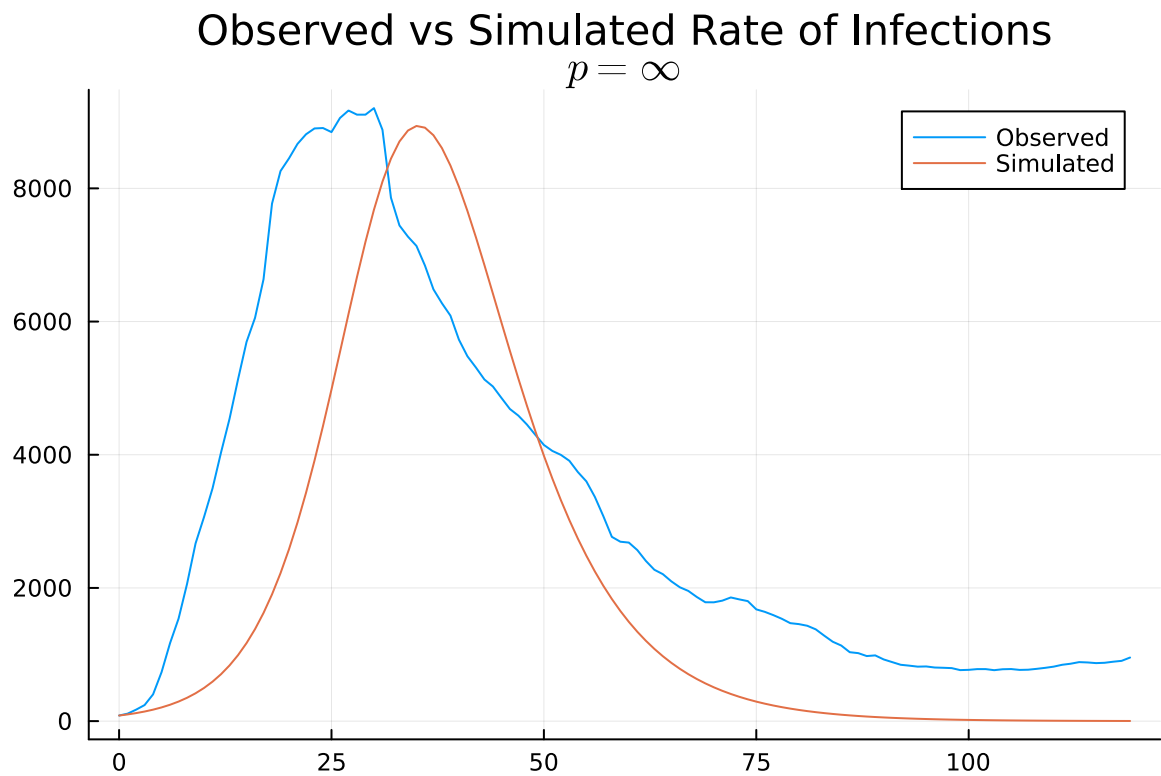


3)

```
In [37]: Ss, Is, Rs = euler( $\hat{\alpha}$ ,  $\hat{\beta}$ ,  $\hat{N}$ );
```

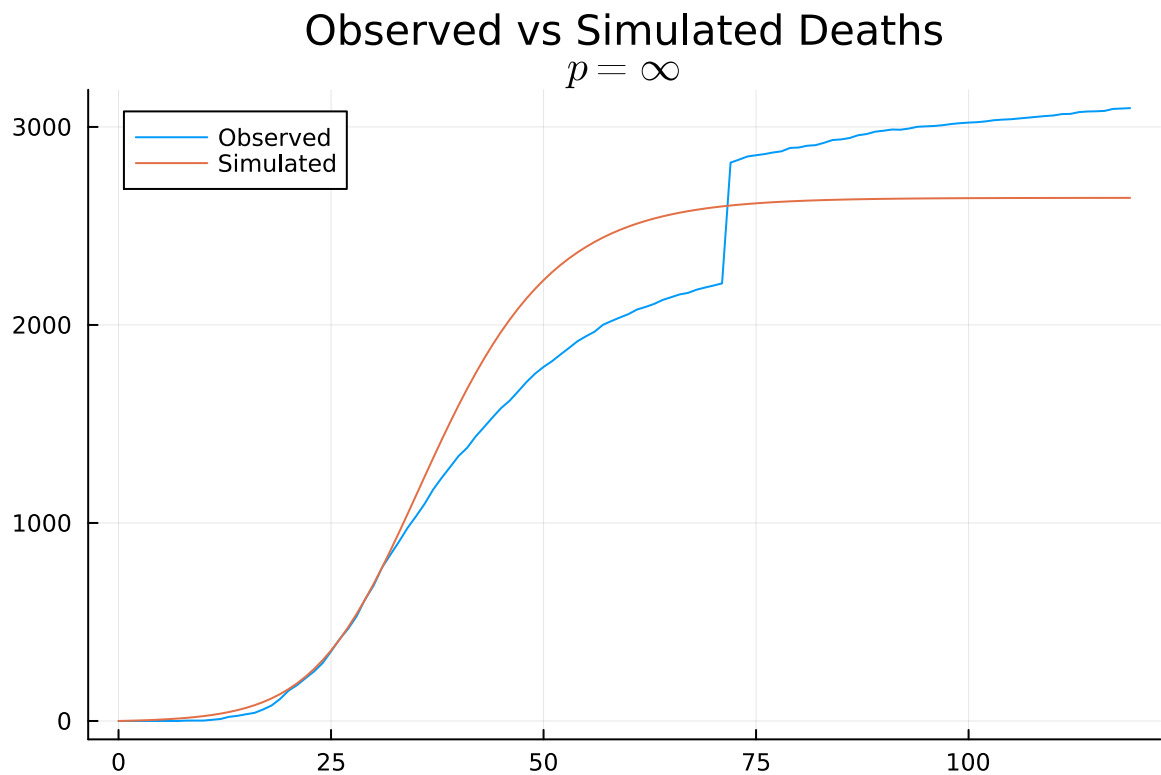
```
In [38]: plot(0:119, I, labels="Observed")
plot!(0:119, Is, label="Simulated")
plot!(title = "Observed vs Simulated Rate of Infections" * "\n" * L"p= $\infty$ ")
```

Out [38]:



```
In [39]: plot(0:119, Y1, labels="Observed")
plot!(0:119,  $\hat{y}$ *Rs, label="Simulated")
plot!(title = "Observed vs Simulated Deaths" * "\n" * L"p= $\infty$ ")
```

Out [39]:



4)

We would increase the value of  $\alpha$  to maybe  $\alpha \in [0.2, 0.6]$  to start. Also, an increase in  $\hat{N}$  seems to return simulated values closer to the observed ones, so we would start by establishing  $\frac{N}{Population} \in [10\%, 20\%]$ . For  $R_0$  we would increase its value since for all values of  $p$  we got  $R_0 = \sup[1.2, 1.9]$ , a good place to start would be  $R_0 \in [1.9, 3]$

## Excercise 2

```
In [40]: # euler method to compute S_sim, E_sim, I_sim, and R_sim
function euler(alpha, beta, delta, N)
    h = 0.01

    S_sim = zeros(tmax+1)
    E_sim = zeros(tmax+1)
    I_sim = zeros(tmax+1)
    R_sim = zeros(tmax+1)

    s = N
    e = I(0)
    i = I(0)
    r = 0

    t = 0
    while t < tmax + 0.0001
        if abs(round(Int, t) - t) < 0.0001
            S_sim[round(Int, t)+1] = s
            E_sim[round(Int, t)+1] = e
            I_sim[round(Int, t)+1] = i
            R_sim[round(Int, t)+1] = r
        end

        ds = -beta*s*(i/N)
        de = beta*s*(i/N) - delta*e
        di = delta*e - alpha*i
        dr = alpha*i

        s += h*ds
        e += h*de
        i += h*di
        r += h*dr
        t += h
    end

    return S_sim, E_sim, I_sim, R_sim
end
```

Out[40]: euler (generic function with 2 methods)

```
In [41]: # objective function J
function J(alpha, beta, delta, N, gamma, p, I_sim, R_sim)
    is_inf = false
    if p == 3
        is_inf = true
```

```

        p = 1
    end

    s1(t) = abs(I(t) - I_sim[t+1])^p
    s2(t) = abs(Y(t) - gamma*R_sim[t+1])^p

    if is_inf
        return maximum(s1.(0:tmax) + λ*s2.(0:tmax))
    else
        return sum(s1, 0:tmax) + λ*sum(s2, 0:tmax)
    end
end
end

```

Out[41]: J (generic function with 1 method)

```

In [42]: min = [Inf, Inf, Inf]
minimizer = [(-1.0,-1.0,-1.0,-1.0,-1.0), (-1.0,-1.0,-1.0,-1.0,-1.0), (-1.0,-

mind = [Dict(), Dict(), Dict()]

# run the euler method for each combination of parameters and find the minim
for alpha in ProgressBar(0.05:0.05:0.4)
    for beta in (1.5:0.1:1.9)*alpha
        for delta in 0.05:0.05:0.4
            for N in (2:10)*(population/100)
                I_sim, R_sim = euler(alpha, beta, delta, N)[3:4]

                for p in 1:3
                    gamma = optGamma(R_sim, p)
                    mind[p][(alpha, beta, delta, N, gamma)] = J(alpha, beta,

                    # if objective_function < min[p]
                    #     min[p] = objective_function
                    #     minimizer[p] = (alpha, beta, delta, N, gamma)
                    # end
                end
            end
        end
    end
end
end
end

```

```

0.0%|          | 0/8 [00:00<00:00, -0
s/it]
12.5%|██████    | 1/8 [00:12<Inf:Inf, InfG
s/it]
25.0%|██████████ | 2/8 [00:34<03:26, 34
s/it]
37.5%|███████████ | 3/8 [01:01<02:32, 30
s/it]
50.0%|██████████████ | 4/8 [01:28<01:58, 29
s/it]
62.5%|██████████████████ | 5/8 [01:56<01:27, 29
s/it]
75.0%|██████████████████████ | 6/8 [02:24<00:58, 29
s/it]
87.5%|██████████████████████████ | 7/8 [02:53<00:29, 29
s/it]
100.0%|██████████████████████████████ | 8/8 [03:21<00:00, 29
s/it]
100.0%|██████████████████████████████████ | 8/8 [03:21<00:00, 29
s/it]

```

```

In [43]: for p in 1:3
          minimizer[p] = argmin(mind[p])
          min[p] = mind[p][minimizer[p]]
        end

```

(1)

```

In [44]: display_p = ["1", "2", "∞"]

# print out the min and minimizer for each value of p
for p in 1:3
    println("p = " * display_p[p] * ":")
    println("minimum of J(alpha, beta, delta, N, gamma) = $(min[p])")
    println("alpha = $(minimizer[p][1]), beta = $(minimizer[p][2]), delta =")
    println("")
end

p = 1:
minimum of J(alpha, beta, delta, N, gamma) = 240107.84793462453
alpha = 0.35, beta = 0.6649999999999999, delta = 0.4, N = 81435.3, gamma =
0.04696363166542318

p = 2:
minimum of J(alpha, beta, delta, N, gamma) = 8.278247653300321e8
alpha = 0.4, beta = 0.76, delta = 0.4, N = 97722.36, gamma = 0.036569733756
936326

p = ∞:
minimum of J(alpha, beta, delta, N, gamma) = 7135.430012387385
alpha = 0.4, beta = 0.76, delta = 0.4, N = 162870.6, gamma = 0.021501287355
507076

```



(2)

```
In [45]: # make the 3 graphs of cross sections of J for each p
for p in 1:3
    alpha, beta, delta, N, gamma = minimizer[p]

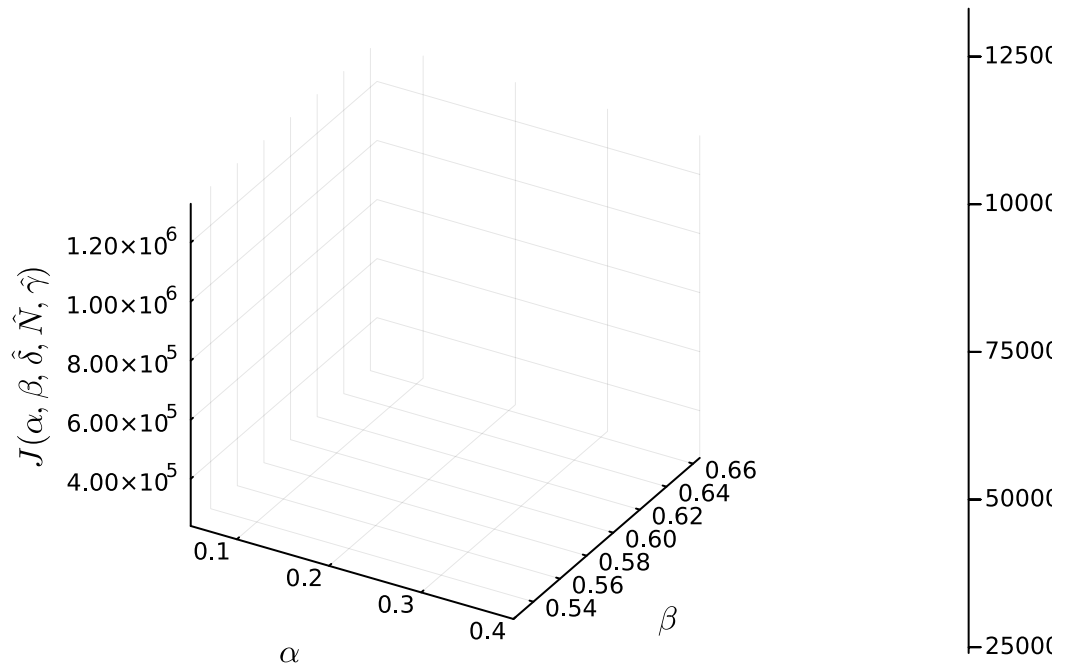
    function f_plot1(alpha, beta)
        S_sim, E_sim, I_sim, R_sim = euler(alpha, beta, delta, N)
        return J(alpha, beta, delta, N, gamma, p, I_sim, R_sim)
    end
    display(surface(0.05:0.05:0.4, (1.5:0.1:1.9)*alpha, f_plot1, label = ["a
        xlabel=L"J(\alpha, \beta, \hat{\delta}, \hat{N}, \hat{\gamma})", xlabel

    function f_plot2(alpha, delta)
        S_sim, E_sim, I_sim, R_sim = euler(alpha, beta, delta, N)
        return J(alpha, beta, delta, N, gamma, p, I_sim, R_sim)
    end
    display(surface(0.05:0.05:0.4, 0.05:0.05:0.4, f_plot2, label = ["alpha"
        xlabel=L"J(\alpha, \hat{\beta}, \delta, \hat{N}, \hat{\gamma})", xlabel

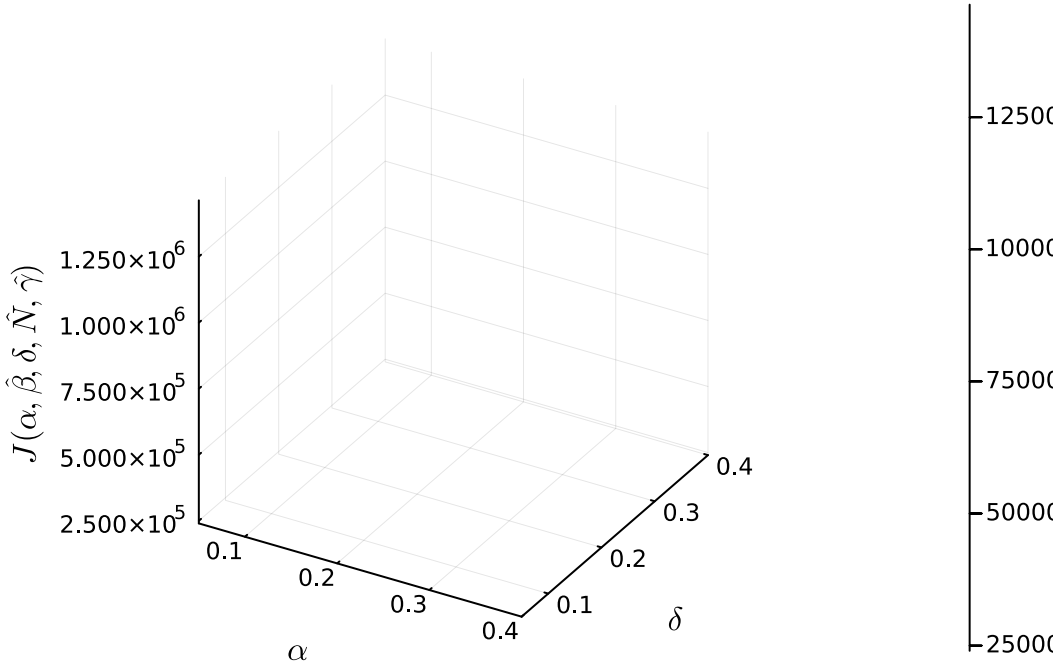
    function f_plot3(beta, delta)
        S_sim, E_sim, I_sim, R_sim = euler(alpha, beta, delta, N)
        return J(alpha, beta, delta, N, gamma, p, I_sim, R_sim)
    end
    display(surface((1.5:0.1:1.9)*alpha, 0.05:0.05:0.4, f_plot3, label = ["a
        xlabel=L"J(\hat{\alpha}, \beta, \delta, \hat{N}, \hat{\gamma})", xlabel

end
```

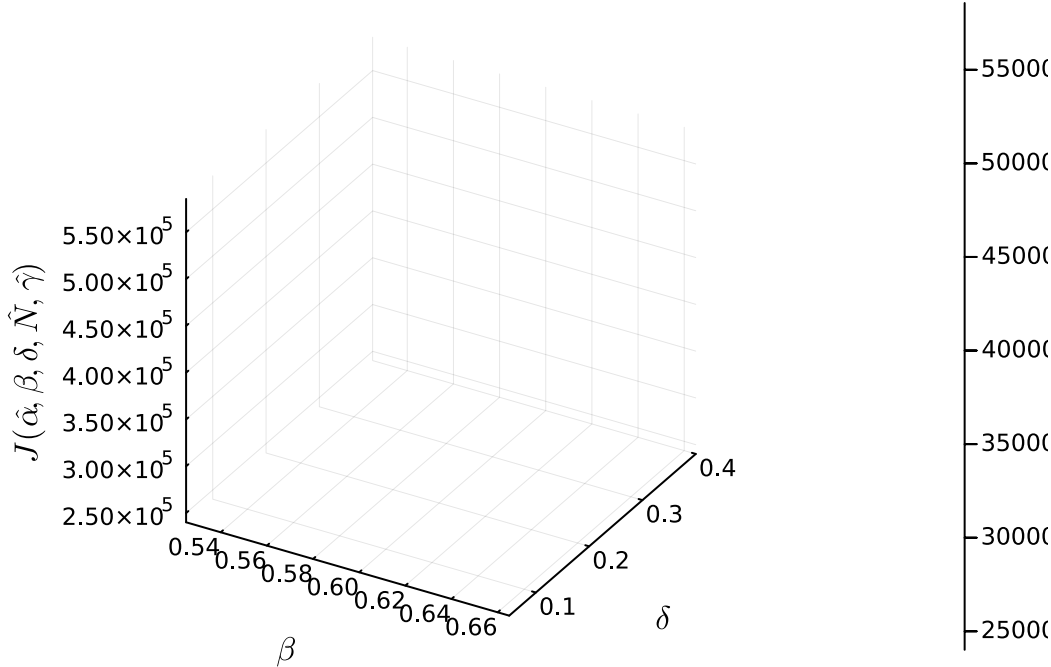
$$J(\alpha, \beta, \hat{\delta}, \hat{N}, \hat{\gamma}) \text{ for } p = 1$$



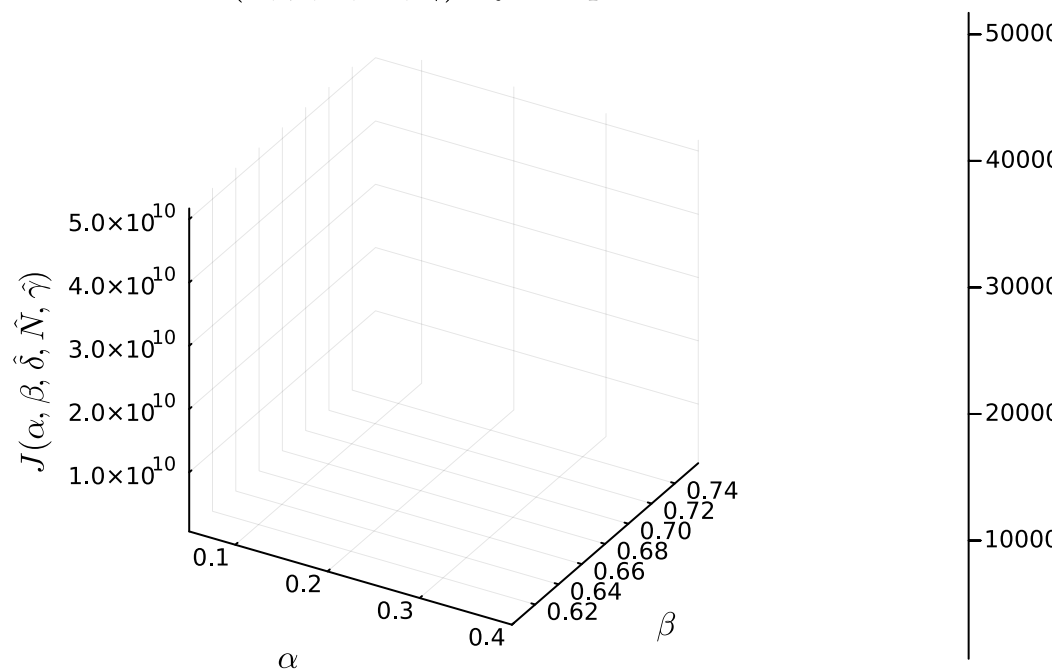
$$J(\alpha, \hat{\beta}, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = 1$$



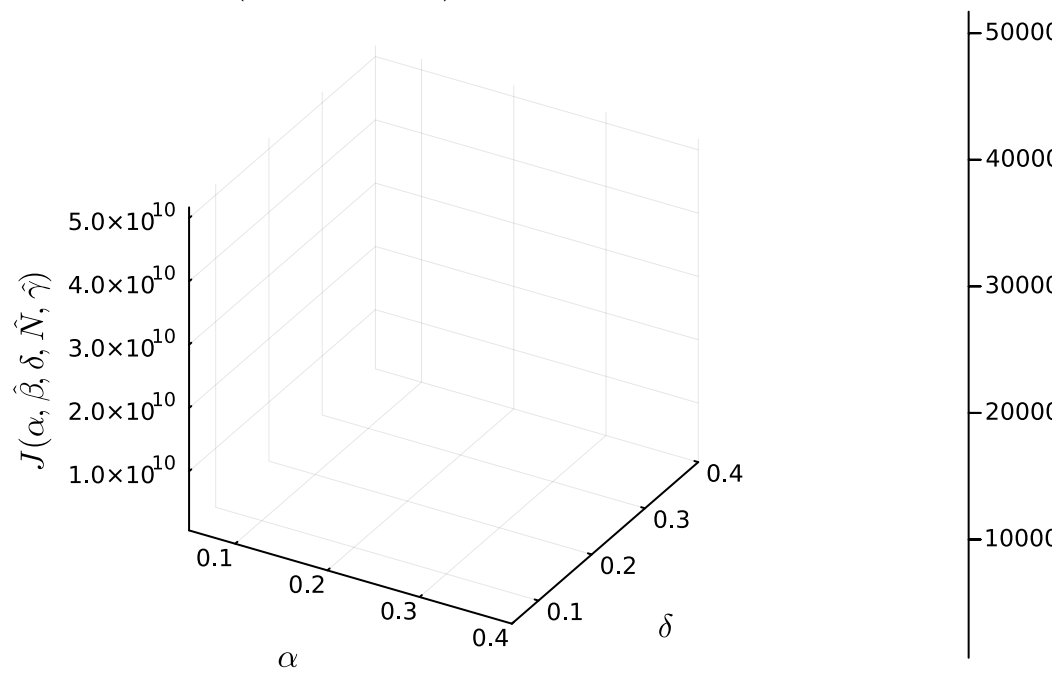
$$J(\hat{\alpha}, \beta, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = 1$$



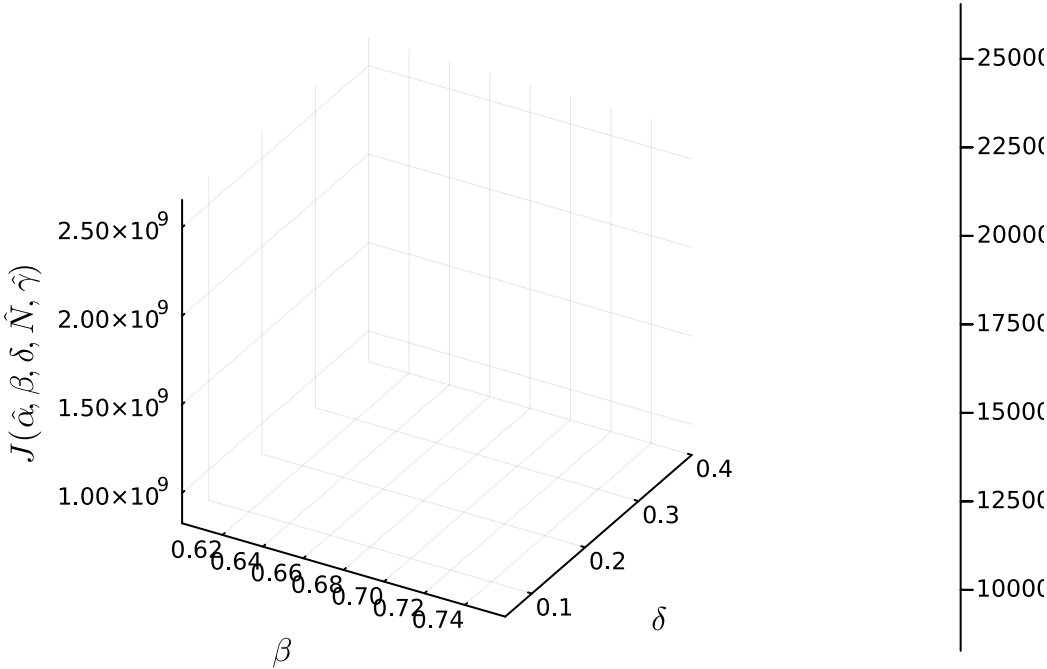
$$J(\alpha, \beta, \hat{\delta}, \hat{N}, \hat{\gamma}) \quad \text{for } p = 2$$



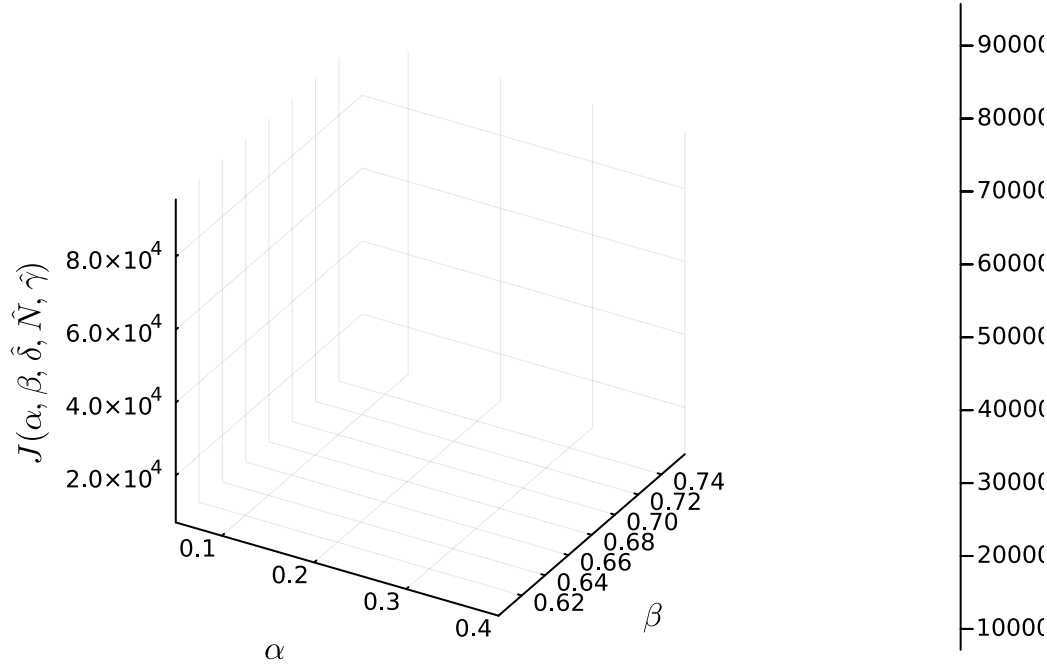
$$J(\alpha, \hat{\beta}, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = 2$$



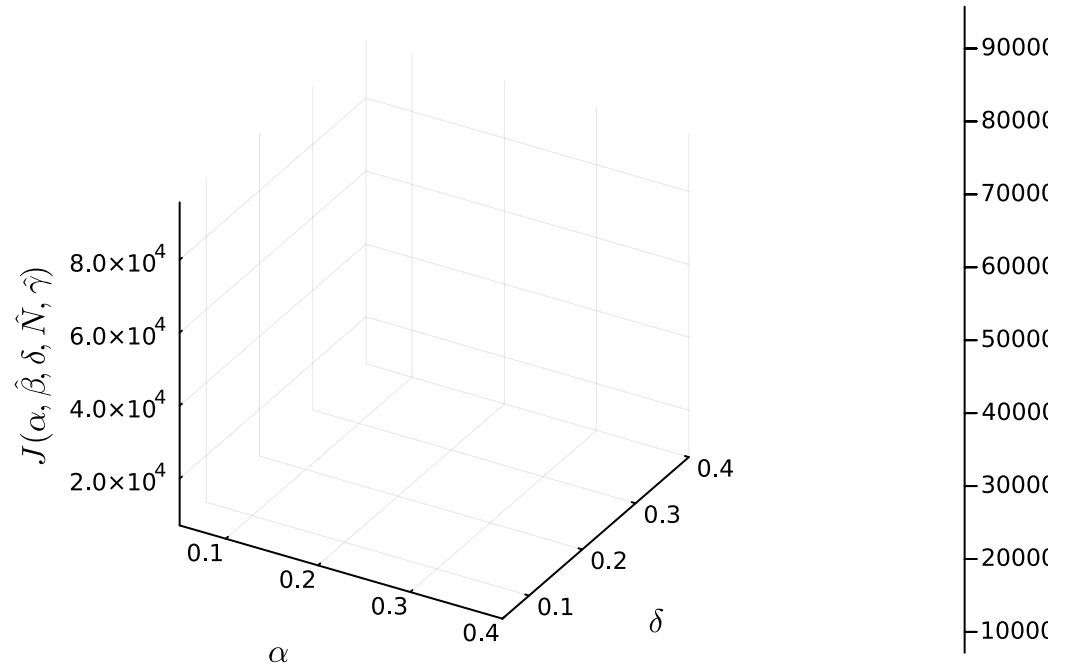
$$J(\hat{\alpha}, \beta, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = 2$$



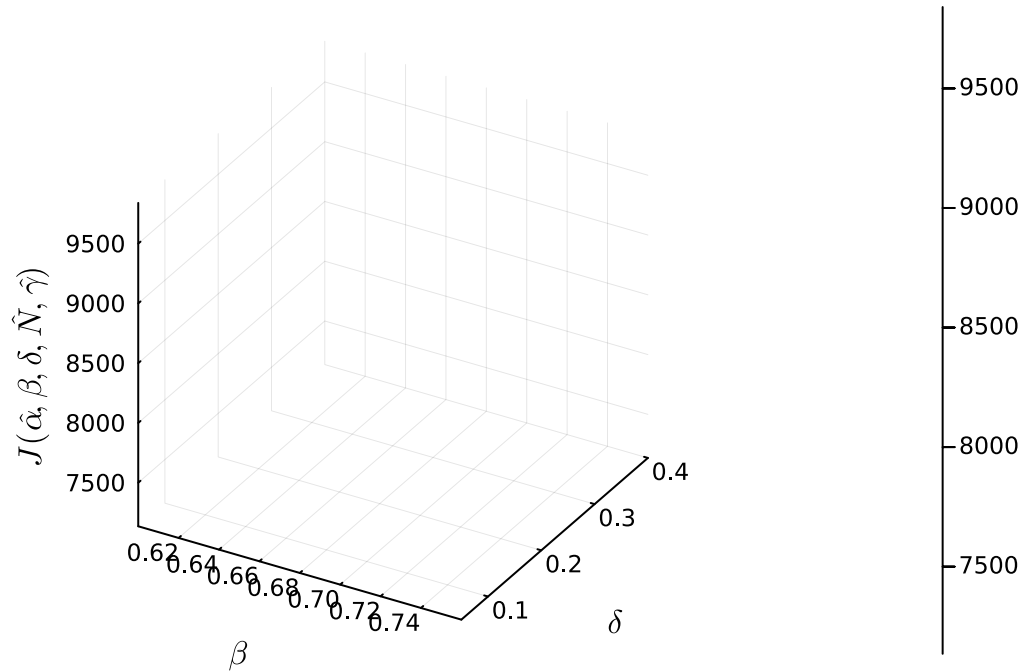
$$J(\alpha, \beta, \hat{\delta}, \hat{N}, \hat{\gamma}) \quad \text{for } p = \infty$$



$$J(\alpha, \hat{\beta}, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = \infty$$



$$J(\hat{\alpha}, \beta, \delta, \hat{N}, \hat{\gamma}) \quad \text{for } p = \infty$$



(3)

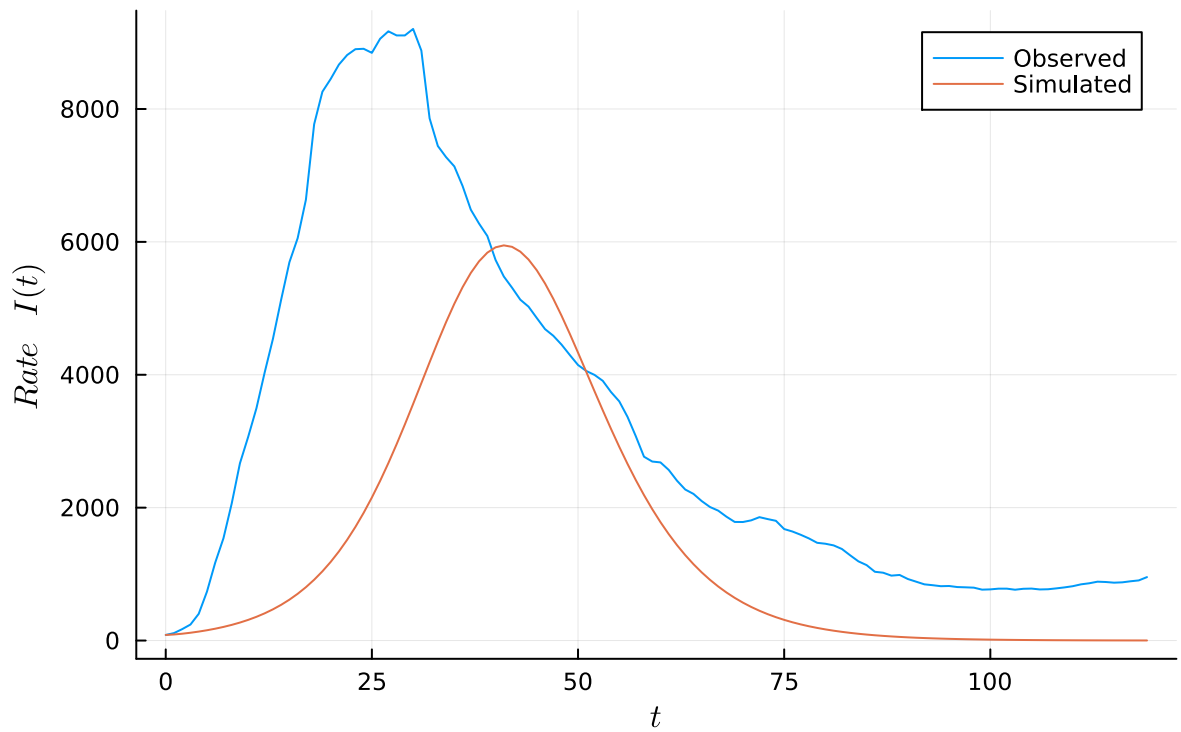
```
In [46]: # plot I vs I_sim and Y vs gamma*R_sim
for p = 1:3
    alpha, beta, delta, N, gamma = minimizer[p]
    S_sim, E_sim, I_sim, R_sim = euler(alpha, beta, delta, N)
```

```

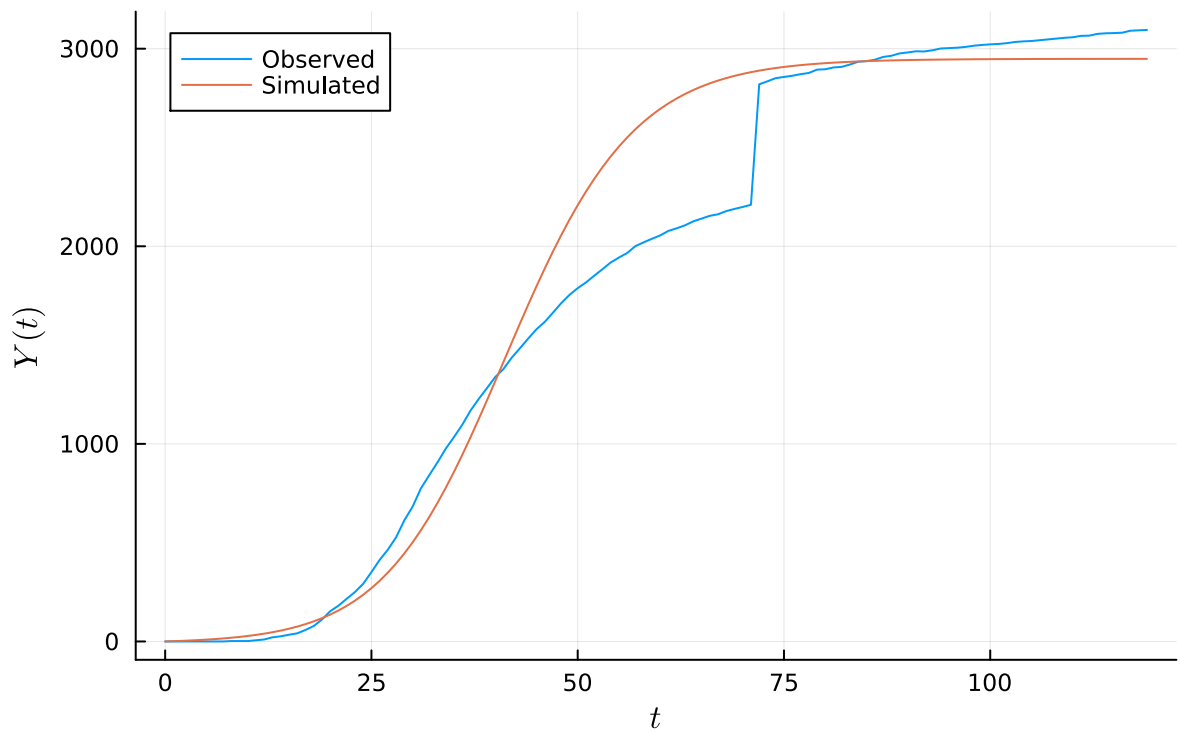
display(plot(0:tmax, [I.(0:tmax), I_sim], label = ["Observed" "Simulated"]
display(plot(0:tmax, [Y.(0:tmax), gamma*R_sim], label = ["Observed" "Simulated"]
end

```

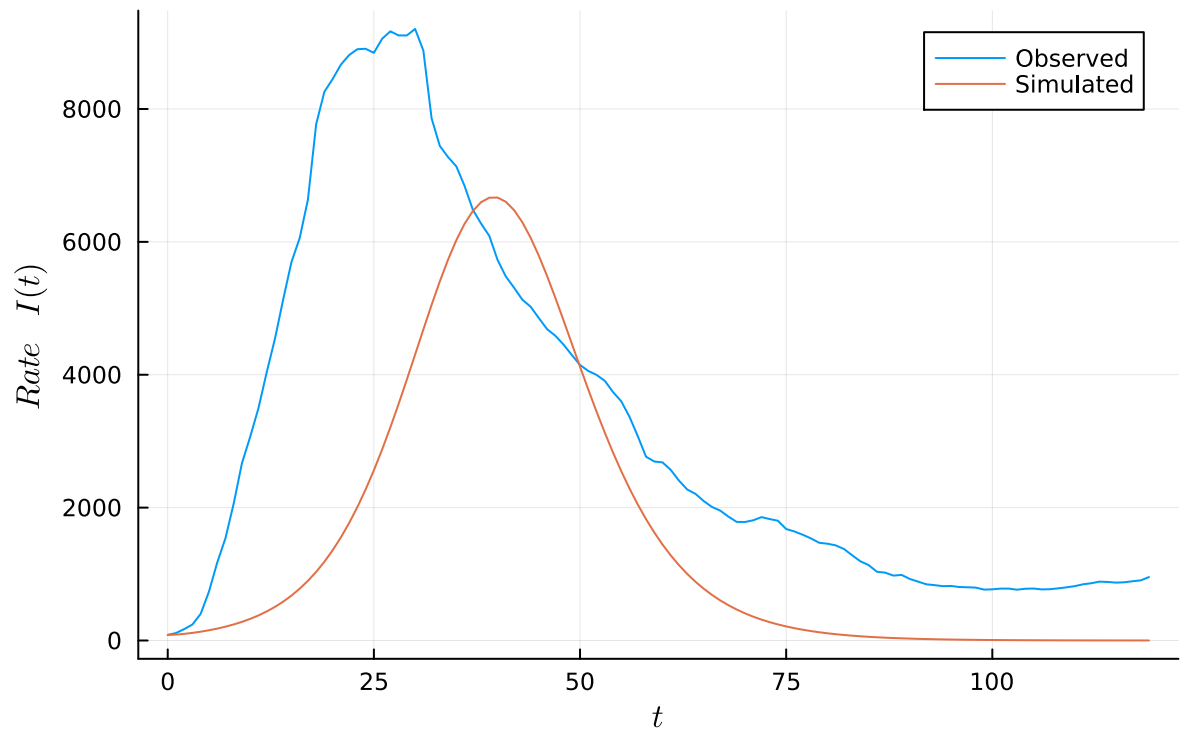
### Rates of Active Infections for $p = 1$



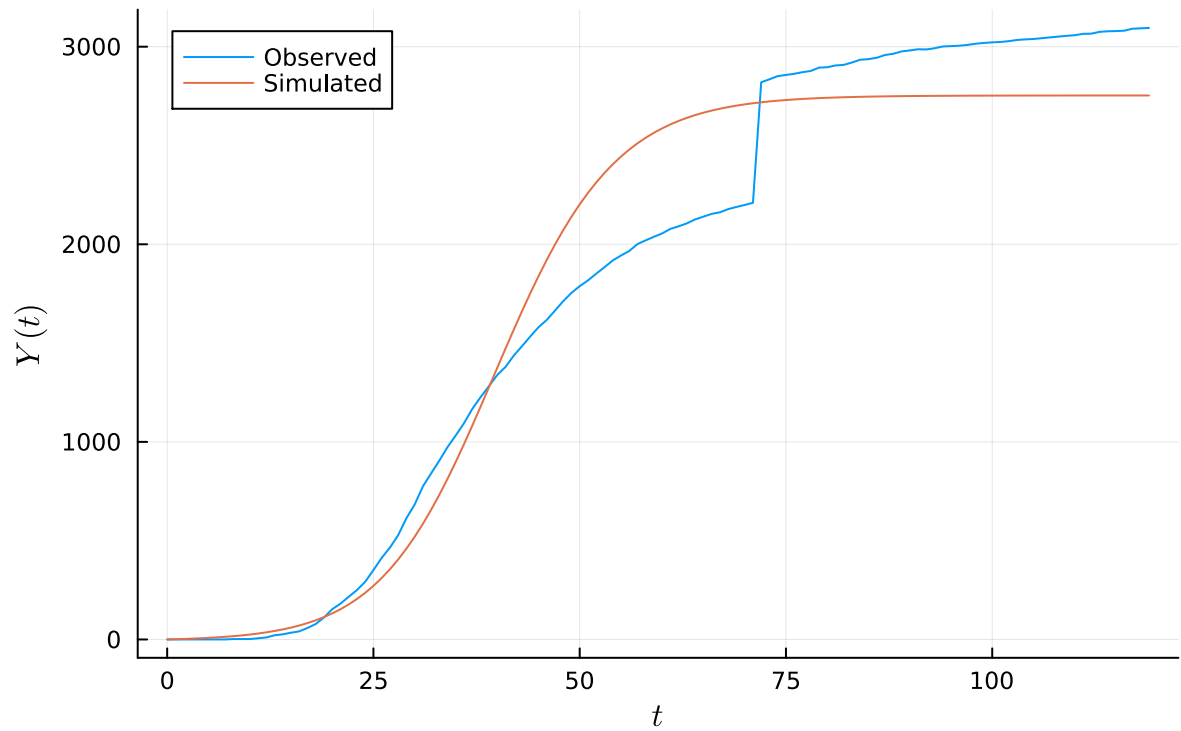
### Cumulative Deaths for $p = 1$



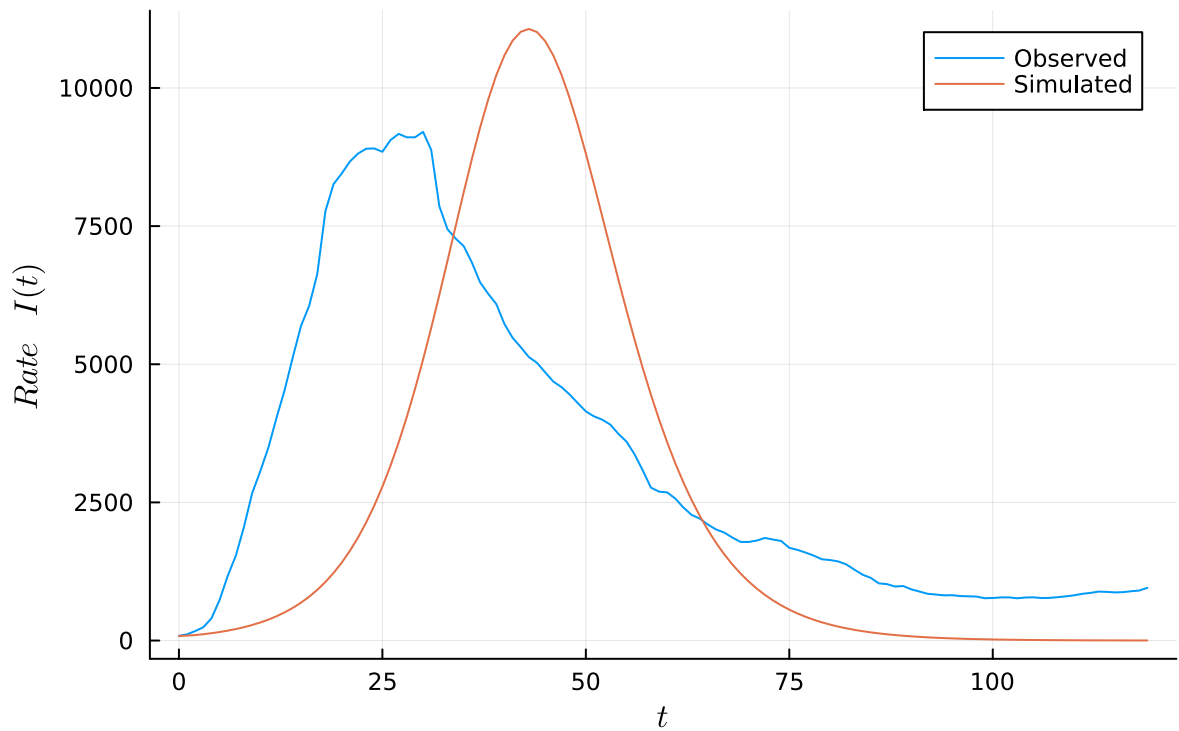
Rates of Active Infections for  $p = 2$



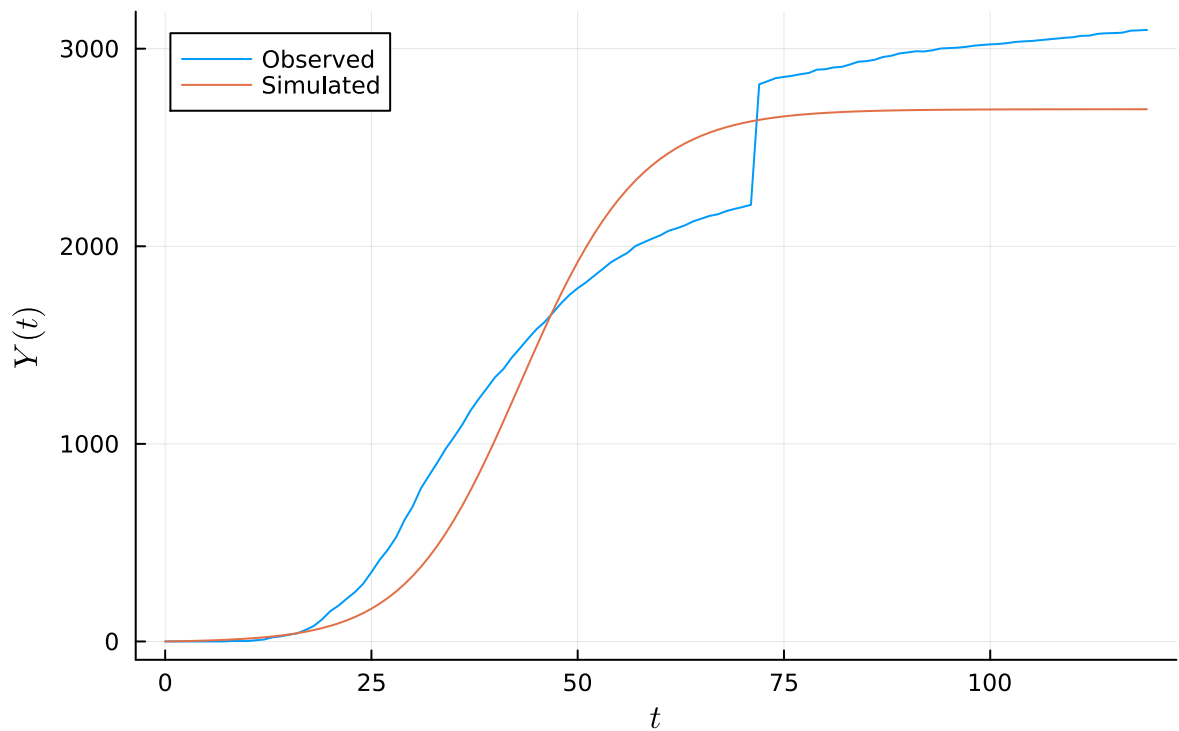
Cumulative Deaths for  $p = 2$



### Rates of Active Infections for $p = \infty$



### Cumulative Deaths for $p = \infty$



(4)

In all of the objective functions with alpha the low values of alpha make the function very large and hard to tell what is happening in the low part. So narrowing in on the optimal alpha and seeing the effect delta and beta have in this reduced range would be good.



For the graphs with beta and delta all three have there lowest point in the corner so it would be a good idea to increase the top of the range being searched for beta and delta. considering this some ranges to check would be alpha in  $[0.3, 0.4]$ ,  $R_0$  in  $[1.5, 3]$ , delta in  $[0.05, 1]$ ,  $N/\text{population}$  in  $[2\%, 10\%]$ .