# 464-A2b-3643806

February 14, 2024

# 0.1 Assignment A2b: Photon Detection

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```
[]: using Distributions, Plots, Random
     # 1A
     function randtimes(N; t1, t2)
         return t1 .+ (t2 - t1) .* rand(N)
     end
     # 1A
     function plotflash(times; t1, t2, size=(600, 150))
         y = ones(length(times))
         scatter(times, y, label="Photon Arrival", color=:green, marker=:circle,
      ⇔size=size)
         for time in times
             plot!([time, time], [0, 1], color="#95B9E1", label="", legend=false)
         end
         #xlabel!("Time (s)")
         xlabel!("")
         #ylabel!("Photon Event")
         ylabel!("")
         title!("Photon Arrival Times")
         xlims!(t1, t2)
         ylims!(0, 1.2)
     end
     # 1B
     function randintervals(N; lambda=lambda, t1=t1)
         intervals = rand(Exponential(1/lambda), N)
         times = cumsum(intervals) .+ t1
         return times
     end
     # 2A
```

```
function poisson_probabilities(lambda, T, n_max)
    n_values = 0:n_max
    probabilities = [(lambda * T)^n / factorial(n) * exp(-lambda * T) for n in_{\sqcup}]
 \hookrightarrown_values]
    return n_values, probabilities
end
# 2B / 3B / 3C
function prob_detect_or_see(K; lambda=nothing, T=nothing, I=nothing, u
 ⇒alpha=nothing)
    effective_rate = if lambda !== nothing && T !== nothing
        lambda * T
    elseif I !== nothing && alpha !== nothing
        alpha * I
    else
        error("Invalid parameters. You must specify either (lambda and T) or (I_{\sqcup}

¬and alpha).")

    end
    poisson_dist = Poisson(effective_rate)
    prob = 1 - cdf(poisson_dist, K - 1)
    return prob
end
# 2B
function detectionprob(K; lambda=40, T=0.1)
    return prob_detect_or_see(K; lambda, T)
end
# 3A
function lightflash(lambda; f1=0.8, f2=2.2)
    n_photons = rand(Poisson(lambda * (f2 - f1)))
    photon_times = sort(f1 .+ (f2 - f1) .* rand(n_photons))
    return photon_times
end
# 3A
function apply_shutter(photon_times, s1, s2)
    return filter(t -> s1 <= t <= s2, photon_times)</pre>
end
# 3A
function detect_photons(photon_times, alpha)
    detected = filter(_ -> rand() < alpha, photon_times)</pre>
    return detected
end
```

```
# 3A
function photon_stream_fig(;lambda=100, alpha=0.06, f1=0.8, f2=2.2, s1=1, s2=2)
   photon_stream = lightflash(lambda; f1=f1, f2=f2)
   photons_through_shutter = apply_shutter(photon_stream, s1, s2)
   detected_photons = detect_photons(photons_through_shutter, alpha)
   p1 = scatter(photon_stream, ones(length(photon_stream)), label="Stream",
                title="Photon Stream", xlims=(f1, f2), ylims=(0.98, 1.01),
                yticks=false, marker=:circle, markersize=2, legend=:topright)
                for x in photon stream
                    plot!([x, x], [0, 1], color=:blue, legend=false) # Assuming_
 ⇔the color you want is blue
                end
   p2 = scatter(photons_through_shutter,__
 →ones(length(photons_through_shutter)), label="Through Shutter",
                title="Photons through Shutter", xlims=(f1, f2), ylims=(0.98, 1.
 ⇔01),
                yticks=false, marker=:circle, markersize=2, legend=:topright)
                for x in photons_through_shutter
                    plot!([x, x], [0, 1], color=:green, legend=false) #__
 Assuming the color you want is green
   p3 = scatter(detected_photons, ones(length(detected_photons)),
 ⇔label="Detected",
                title="Detected Photons", xlims=(f1, f2), ylims=(0.98, 1.01),
                yticks=false, marker=:circle, markersize=2, legend=:topright)
                for x in detected photons
                    plot!([x, x], [0, 1], color=:red, legend=false) # Assuming_
 → the color you want is red
                end
   plot(p1, p2, p3, layout=(3, 1), xlabel="", xticks=(f1:0.5:f2), size=(600, ____
 400),
       plot title="")
end
# 3B
function probseeing(I; alpha=0.06, K=6)
   return prob_detect_or_see(K; I=I, alpha=alpha)
end
# 3C
function plotdetectioncurve(;alphas=[0.5], Ks=[6], I_range=(0.01, 100), __
 ⇒points=100)
   xs = exp10.(range(log10(I_range[1]), log10(I_range[2]), length=points))
```

```
plot(xscale=:log10, xlabel="Intensity I", ylabel="Percentage Detected",
        title="Detection Curve", legend=:topleft, xlims=I_range, ylims=(0,__
 →100))
   for alpha in alphas
       for K in Ks
           ys = [prob_detect_or_see(K; I=x, alpha=alpha) for x in xs] .* 100
           plot!(xs, ys, label=" =$alpha, K=$K")
       end
   end
   plot!()
end
# 3D
function plotfit(;alpha=0.01, K=6)
   I_{range} = (10, 250)
   points = 100
   xs = exp10.(range(log10(I_range[1]), log10(I_range[2]), length=points))
   p = plot(xscale=:log10, xlabel="photons at cornea", ylabel="% flashes_1

detected",
            title="", legend=:topleft, xlims=I_range, ylims=(0, 110),
            xticks=([10.0, 20.0, 30.0, 40.0, 50.0, 60.0, 70.0, 80., 90.0, 100.
 size=(400, 400))
   ys_new1 = [prob_detect_or_see(2; I=x, alpha=0.02) for x in xs] * 100
   plot!(p, xs, ys_new1, label="Model: =0.02, K=2", linewidth=2, linestyle=:

dash, color=:blue)

   ys = [prob_detect_or_see(K; I=x, alpha=alpha) for x in xs] * 100
   plot!(p, xs, ys, label="Model: =$alpha, K=$K", linewidth=2, linestyle=:
 ⇔solid, color=:orange)
   ys_new2 = [prob_detect_or_see(12; I=x, alpha=0.13) for x in xs] * 100
   plot!(p, xs, ys_new2, label="Model: =0.13, K=12", linewidth=2, linestyle=:
 →dash, color=:green)
   # HSP subject SS data
   I_{data} = [24.1, 37.6, 58.6, 91.0, 141.9, 221.3]
   percentage_seen = [0.0, 4.0, 18.0, 54.0, 94.0, 100.0]
   scatter!(p, I_data, percentage_seen, label="HSP SS Data", color=:white, __
 →markersize=3, markeralpha=0.6)
   return p
```

end

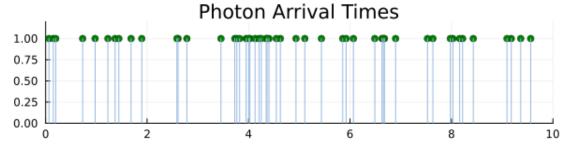
[]: plotfit (generic function with 1 method)

#### 0.1.2 1a. Random times

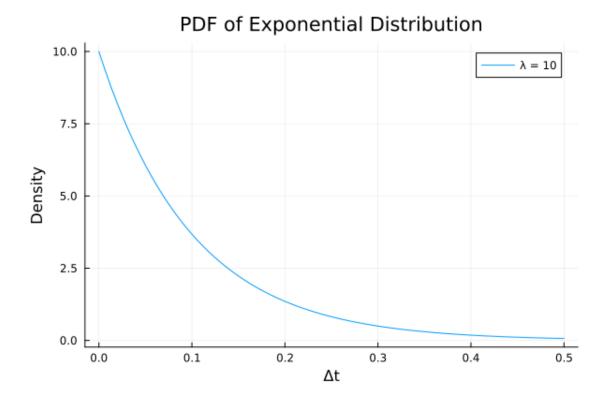
```
[]: # Sample data
N = 50
t1 = 0
t2 = 10
times = randtimes(N; t1=t1, t2=t2)

plotflash(times; t1, t2)
```

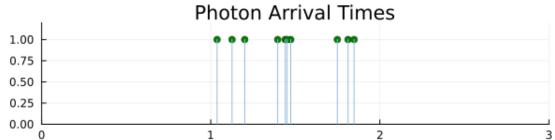
[]:



#### 0.1.3 1b. Random intevals







#### 0.1.4 1c. Seeing the flash

The two methods for simulating photon arrival times differ in how they model the process:

1. Uniform Distribution Method: Generates events uniformly across a time interval, sim-

plifying the process but not accurately reflecting the random nature of photon arrivals.

2. Exponential Distribution Method: Simulates the time between successive events using an exponential distribution, closely mirroring the actual random, independent occurrence of photons in a Poisson process.

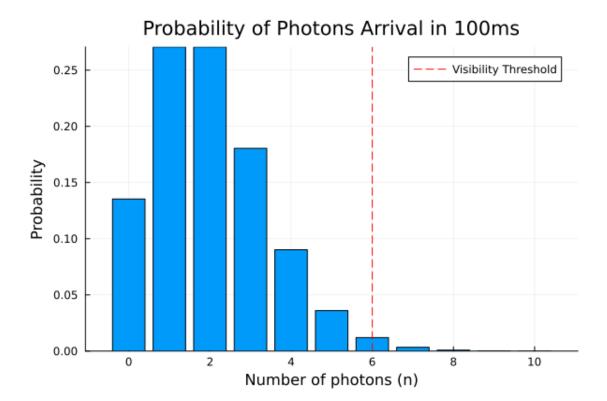
The detection of light flashes is probabilistic because:

- Quantum variability: Photons behave unpredictably; their arrival and detection are subject to quantum randomness.
- **Detection threshold**: A minimum number of photons must be detected within a certain timeframe to differentiate a light flash from background noise, due to the sensory system's limitations and internal noise.
- Random arrival: Photon arrival follows a Poisson process, making the number of photons detected in any interval inherently random.

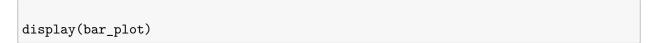
In summary, the probabilistic nature of detecting flashes arises from the random arrival of photons, quantum uncertainty, and the sensory system's need to distinguish signal from noise.

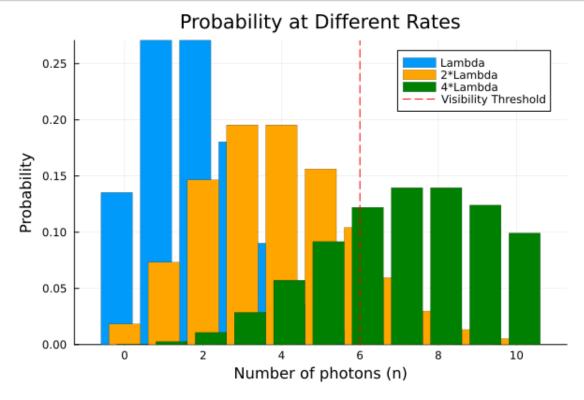
#### 0.1.5 2a. The probability of K photons

```
[]: # Sample Parameters
     lambda
     Т
                 = 0.1 # observation period in seconds
                 = 10
                        # maximum number of photons to calculate probabilities for
     n max
                        # (visible threshold) minimum number of photons for
     v_{threshold} = 6
      ⇔visibility
     n_values, probabilities = poisson_probabilities(lambda, T, n_max)
     bar_plot = bar(n_values, probabilities, legend=:topright, xlabel="Number of_"
      ⇒photons (n)", ylabel="Probability",
         title="Probability of Photons Arrival in 100ms", label=false, xticks=0:2:
      \hookrightarrown_max)
     visible_indices = findall(x -> x >= v_threshold, n_values)
     # Adding visibility threshold indicator
     vline!([v_threshold], linestyle=:dash, color=:red, label="Visibility Threshold")
     display(bar_plot)
```

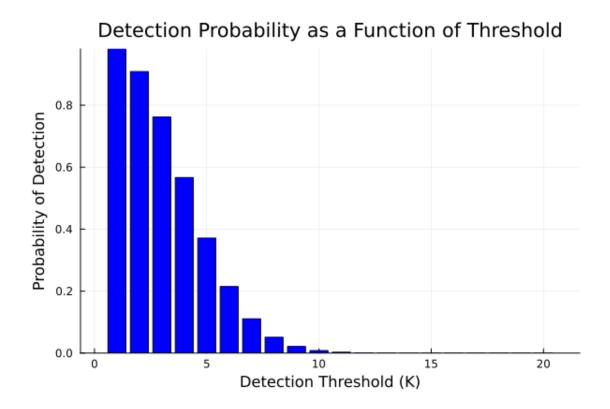


```
[]: # Sample Parameters
     lambda
                 = 0.1 # observation period in seconds
                 = 10
                        # maximum number of photons to calculate probabilities for
                        # (visible threshold) minimum number of photons for
     v_threshold = 6
      ⇔visibility
     # Calculate probabilities for the initial, double, and quadruple rates
     n_values, probabilities_initial = poisson_probabilities(lambda * 1, T, n_max)
     _, probabilities_double
                                     = poisson_probabilities(lambda * 2, T, n_max)
     _, probabilities_quadruple
                                     = poisson_probabilities(lambda * 4, T, n_max)
     # Plotting
     bar_plot = bar(n_values .- 0.2, probabilities_initial, width=0.2,__
      ⇔label="Lambda", legend=:topright,
                    xlabel="Number of photons (n)", ylabel="Probability",
                    title="Probability at Different Rates", xticks=0:2:n_max)
     bar!(n_values, probabilities_double, width=0.2, label="2*Lambda", color=:orange)
     bar!(n_values .+ 0.2, probabilities_quadruple, width=0.2, label="4*Lambda",__
      ⇔color=:green)
     # Adding visibility threshold indicator
     vline!([v_threshold], linestyle=:dash, color=:red, label="Visibility Threshold")
```





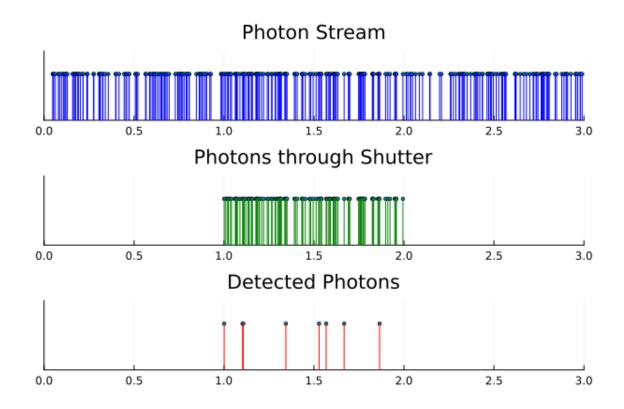
#### 0.1.6 2b. The probability of K or more photons



# 0.1.7 3a. Simulating the photon stream

```
[]: # Sample Parameters
lambda = 100  # Photon rate per millisecond
f1, f2 = 0.0, 3.0  # Start and stop times in milliseconds for the photon stream
s1, s2 = 1.0, 2.0  # Shutter open and close times in milliseconds
alpha = 0.06  # Detection probability

photon_stream_fig(lambda=lambda, alpha=alpha, f1=f1, f2=f2, s1=s1, s2=s2)
[]:
```



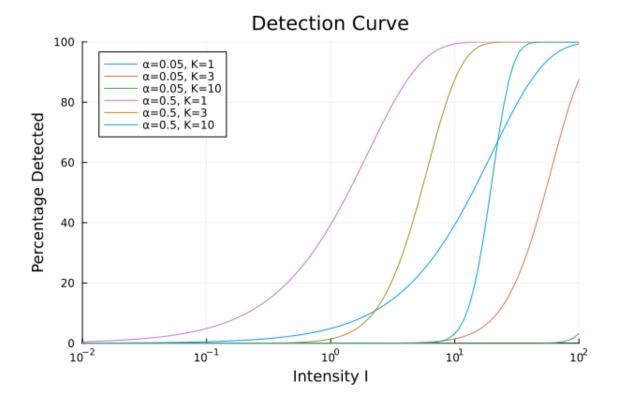
# 0.1.8 3b. Probability of seeing

```
[]: # Sample parameters
I = 300
prob = probseeing(I)
println("Probability of seeing: ", prob)
```

Probability of seeing: 0.9996760065488989

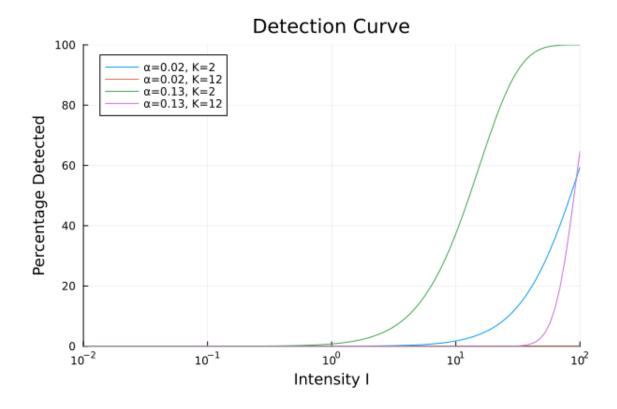
# 0.1.9 3c. Plotting % detected vs light intensity for different parameters

```
[]: plotdetectioncurve(alphas=[0.05, 0.5], Ks=[1, 3, 10], I_range=(.01, 100))
[]:
```

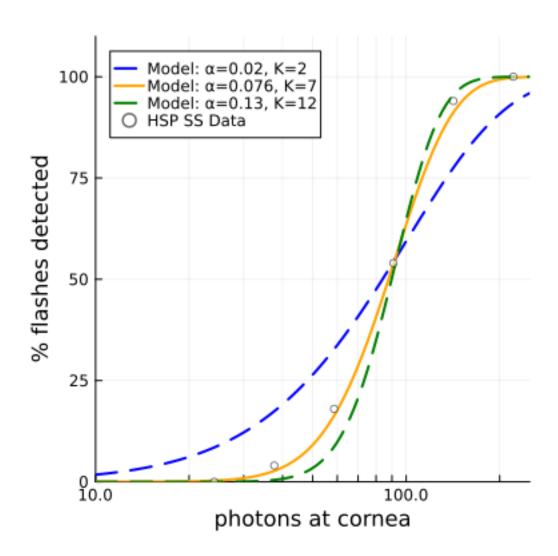


0.1.10 3d. Fitting parameters to experimental data

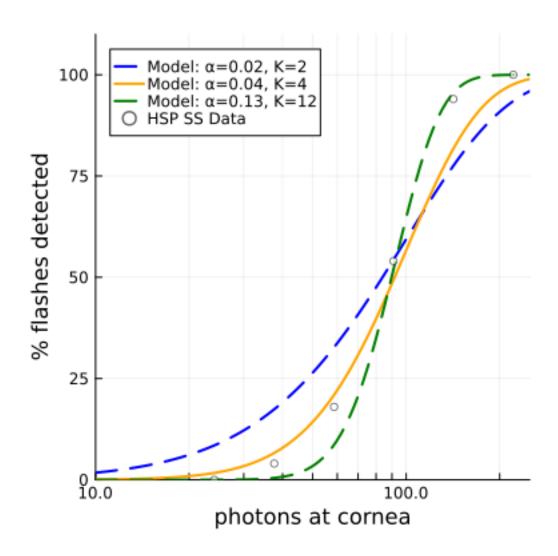
```
[]: plotdetectioncurve(alphas=[0.02, 0.13], Ks=[2, 12], I_range=(.01, 100))
[]:
```

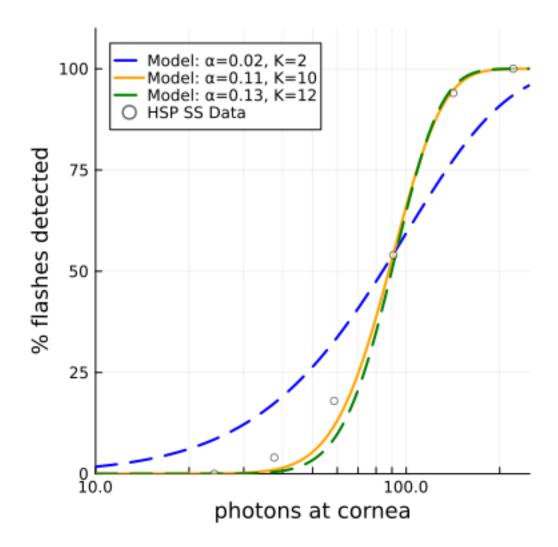


```
[]: # Optimal values plot
display(plotfit(alpha=0.076, K=7))
```



```
[]: # Sub-optimal values plots
display(plotfit(alpha=0.04, K=4))
display(plotfit(alpha=0.11, K=10))
```





Seeing a dim flash of light involves the following key points:

- 1. **Photon Arrival**: A certain number of photons must reach the eye. Not all photons emitted by a source will make it to the retina.
- 2. **Absorption**: Photoreceptor cells (rods) in the eye must absorb these photons. Only a fraction of arriving photons are absorbed, determined by the absorption rate  $(\alpha)$ .
- 3. **Detection Threshold**: There's a minimum number of photons (K) that must be absorbed within a short time frame for the flash to be perceived. This threshold varies but can be very low, often just a few photons under ideal conditions.
- 4. **Statistical Nature**: The process is probabilistic, governed by factors like the absorption rate and the physiological noise in the eye's photoreceptors.

In essence, detecting a dim flash of light is a combination of the number of photons reaching the eye, the efficiency of the eye in absorbing those photons, and the statistical likelihood of those

photons being enough to surpass the sensory threshold for perception.

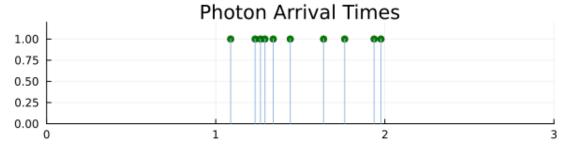
```
[]: # Question 1

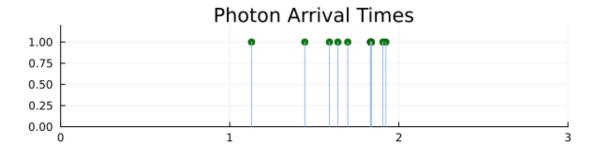
# Parameters
N = 10
t1 = 1
t2 = 2

t = randtimes(N; t1=t1, t2=t2)
p1 = plotflash(t; t1=0, t2=3)

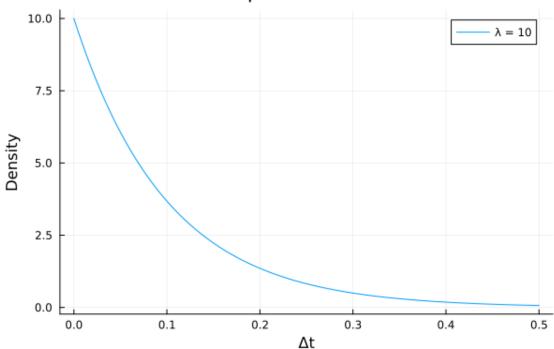
t = randtimes(N; t1=t1, t2=t2)
p2 = plotflash(t; t1=0, t2=3)

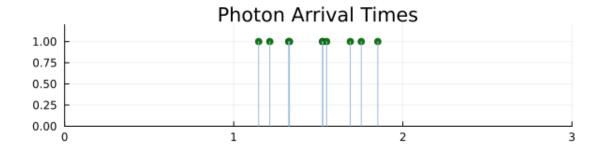
display(p1)
display(p2)
```

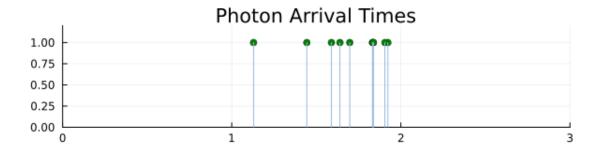




# PDF of Exponential Distribution







```
Probability of receiving 0 photons in 0.1 seconds: 0.36787944117144233 Probability of receiving 1 photons in 0.1 seconds: 0.36787944117144233 Probability of receiving 2 photons in 0.1 seconds: 0.18393972058572117 Probability of receiving 3 photons in 0.1 seconds: 0.061313240195240384
```

```
[]: # Question 5
lambda = 20
T = 0.1 # observation period in seconds
n_max = 10 # maximum number of photons to calculate probabilities for
v_threshold = 6 # (visible threshold) minimum number of photons for
v_visibility

n_values, probabilities = poisson_probabilities(lambda, T, n_max)
```

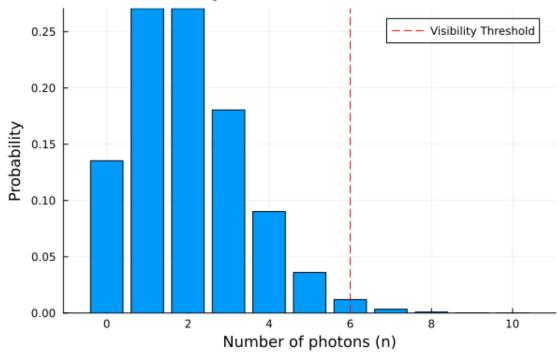
```
bar_plot = bar(n_values, probabilities, legend=:topright, xlabel="Number of_ophotons (n)", ylabel="Probability",
   title="Probability of Photons Arrival in 100ms", label=false, xticks=0:2:
   on_max)

visible_indices = findall(x → x >= v_threshold, n_values)

# Adding visibility threshold indicator
vline!([v_threshold], linestyle=:dash, color=:red, label="Visibility Threshold")

display(bar_plot)
```





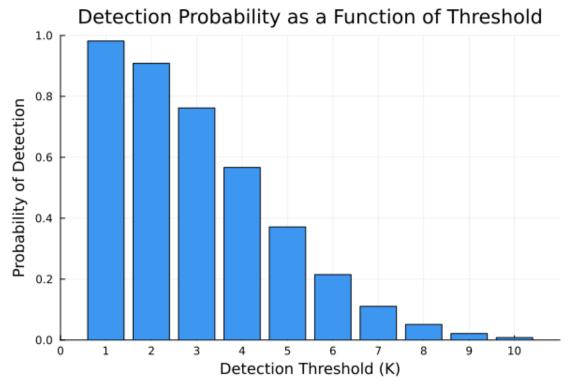
```
[]: # Question 6 detectionprob(6; lambda=40, T=0.1)
```

#### []: 0.21486961296959461

```
# Calculate detection probabilities for thresholds from 1 to K_max
K_values = 1:K_max
probabilities = [detectionprob(K; lambda=lambda, T=T) for K in K_values]

# Plotting
bar_plot = bar(K_values, probabilities, legend=false, xlabel="Detection_"
Threshold (K)", ylabel="Probability of Detection",
    title="Detection Probability as a Function of Threshold", color="#3D97F1",
    ylim=(0,1),
    xlims=(0,11),
    xticks=0:1:10)

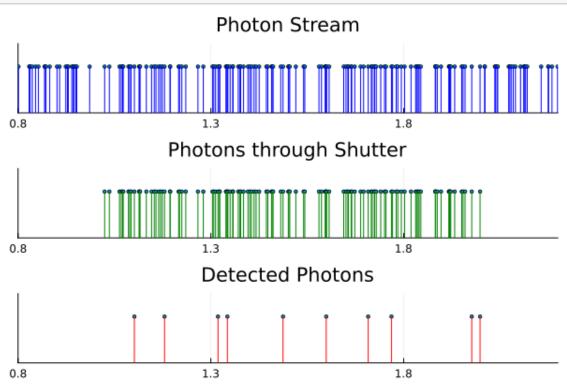
display(bar_plot)
```

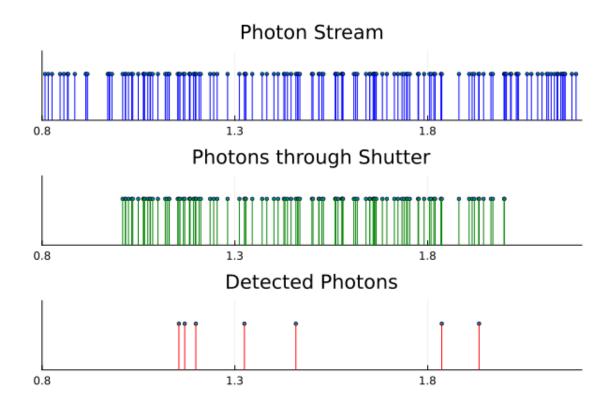


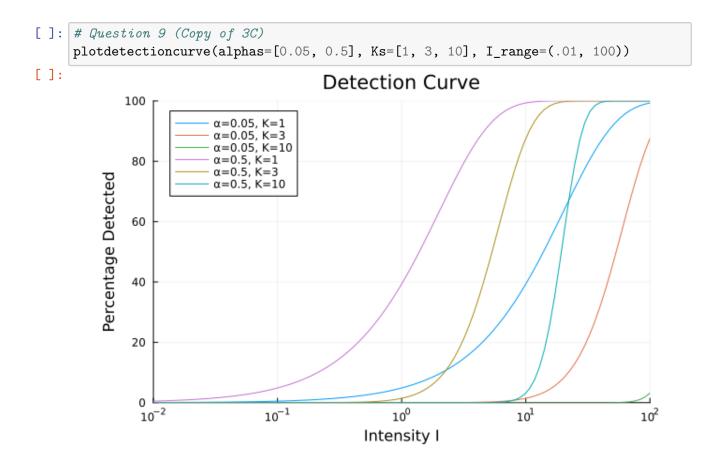
```
# Question 8

# Parameters
lambda = 100  # Photon rate per millisecond
f1, f2 = 0.8, 2.2  # Start and stop times in milliseconds for the photon stream
s1, s2 = 1.0, 2.0  # Shutter open and close times in milliseconds
alpha = 0.06  # Detection probability
```

```
p1 = photon_stream_fig(lambda=lambda, alpha=alpha, f1=f1, f2=f2, s1=s1, s2=s2)
p2 = photon_stream_fig(lambda=lambda, alpha=alpha, f1=f1, f2=f2, s1=s1, s2=s2)
display(p1)
display(p2)
```



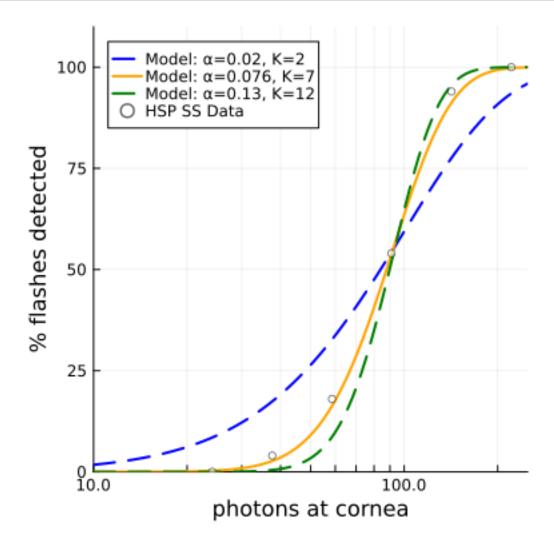


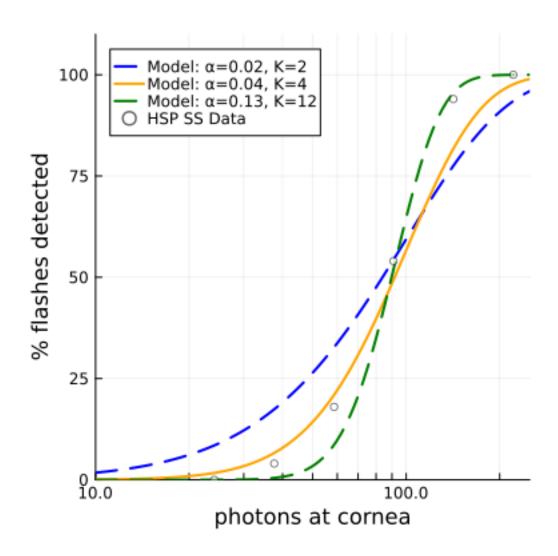


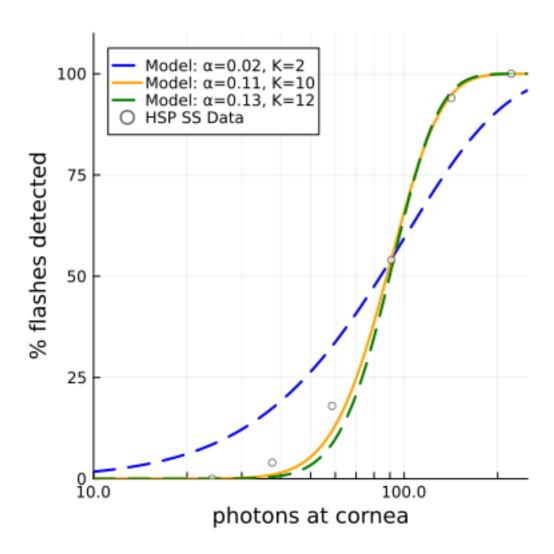
```
[]: # Question 10 (Copy of 3D)

# Optimal values plot
display(plotfit(alpha=0.076, K=7))

# Sub-optimal values plots
display(plotfit(alpha=0.04, K=4))
display(plotfit(alpha=0.11, K=10))
```

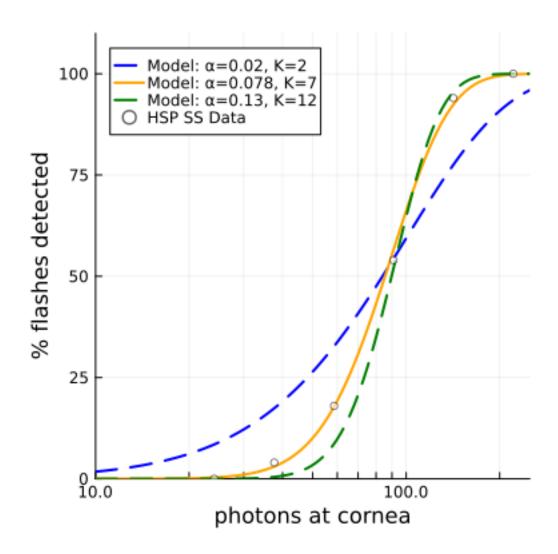






```
[]: # Question 11

# Optimal values plot
display(plotfit(alpha=0.078, K=7))
```



```
[]: # Question 12

# Optimal values plot
display(plotfit(alpha=0.078, K=7))
```

