

```
In [15]: using Interact
using Gadfly
using Distributions
```

```
In [2]: set_default_plot_size(25cm,10cm)
```

# Chapter 1 - Numerical Differentiation and Integration

## 1. Numerical derivatives

### Exercise 1.

Let  $f$  be a continuous and twice differentiable function of  $x$  in the interval  $[0.05, 2]$ , such that:

$$f(x) = 2 - 0.5x^{-0.5} - 0.5x^{-0.2}$$

Compute the analytical derivative and numerical derivatives with both forward and two-sided approximation. Evaluate the three functions in the  $[0.05, 2]$  interval and plot the absolute values of the difference between the analytical derivative and each numerical approximation. What do you conclude?

```
In [3]: f(x) = 2-0.5x^(-0.5)-0.5x^(-.02);
```

Analytical derivative:

$$f'(x) = 0.25x^{-1.5} + 0.1x^{-1.2}$$

```
In [4]: df(x) = 0.25x^(-1.5)+0.1x^(-1.2);
```

Numerical derivative - forward approximation:

$$f'(x_0) \approx \frac{f(x_0 + h) - f(x_0 - h)}{h}, \quad h = x - x_0$$

```
In [5]: df_forward(x) = (f(x+h)-f(x))/h; h = 10.0^-10;
```

Numerical derivative - two-sided approximation:

$$f'(x_0) \approx \frac{f(x_0 + h) - f(x_0 - h)}{2h}, \quad h = x - x_0$$

```
In [6]: df_2sided(x) = (f(x+h)-f(x-h))/(2h); h = 10.0^-10;
```

Let us now plot the results for the interval [0.5, 2]:

```
In [7]: function ex1_Differentiation()

    x = 0.5:0.01:10

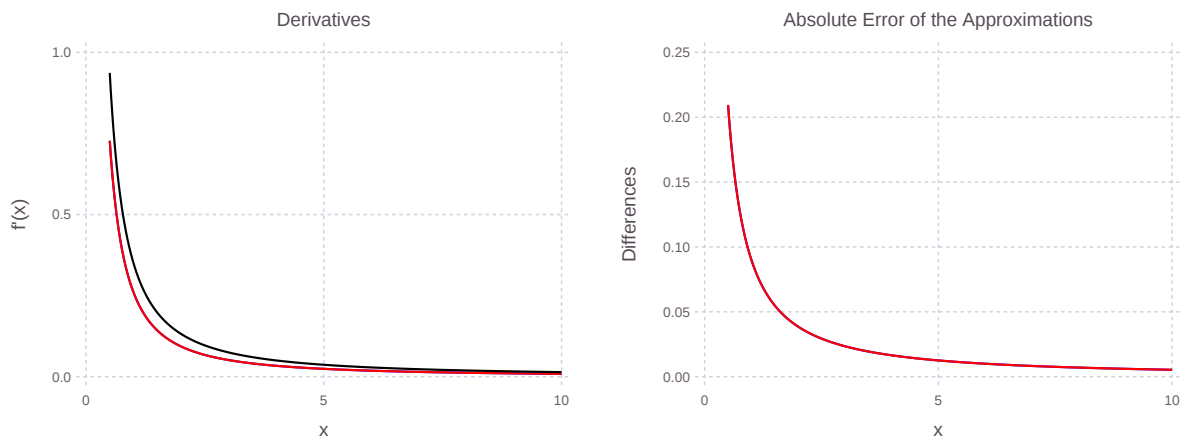
    results = plot(
        layer(x=x, y=df.(x), Geom.line, Theme(default_color=colorant"black")),
        layer(x=x, y=df_forward.(x), Geom.line, Theme(default_color=colorant"red")),
        layer(x=x, y=df_2sided.(x), Geom.line, Theme(default_color=colorant"blue")),
        Guide.Title("Derivatives"),
        Guide.xlabel("x"),
        Guide.ylabel("f'(x)")
    )

    difference = plot(
        layer(x=x, y=abs.(df.(x)-df_forward.(x)), Geom.line, Theme(default_color=colorant"red")),
        layer(x=x, y=abs.(df.(x)-df_2sided.(x)), Geom.line, Theme(default_color=colorant"blue")),
        Guide.Title("Absolute Error of the Approximations"),
        Guide.xlabel("x"),
        Guide.ylabel("Differences")
    )

    return hstack(results, difference)
end;
```

```
In [8]: ex1_Differentiation()
```

Out[8]:



```
In [9]: println("Sum of the absolute error of the forward approximation:", "\n", sum(a
println("Sum of the absolute error of the two-sided approximation:", "\n", sum
```

Sum of the absolute error of the forward approximation:  
23.046544815621964

Sum of the absolute error of the two-sided approximation:  
23.046518170269373

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## 2. Numerical Integrals

### Exercise 2.

Let  $f(x) = x^2$ . Compute  $\int_1^4 x^2 dx$  by making use of the left and right Riemann sums and the average of both. Compare the numerical approximations with the analytical solution for  $n = 50, 500, 5000$ .

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Analytical solution:

$$\int_1^4 x^2 dx = \left[ \frac{x^3}{3} \right]_1^4 = \frac{64}{3} - \frac{1}{3} = 21$$

Numerical approximation:

```
In [10]: function ex2_RiemannSums()

    a = 1; b = 4; f(x) = x.^2
    results = ["n" "l_node" "r_node" "mean"; zeros(3, 4)]

    for (i, n) in enumerate([50, 500, 5000])
        dx = (b-a)/n
        l_int = dx*f.(a:dx:4-dx)
        r_int = dx*f.(a+dx:dx:4)
        results[i+1,:] = [n sum(l_int) sum(r_int) sum(.5*(l_int+r_int))]
    end

    return results
end;
```

```
In [11]: ex2_RiemannSums()
```

```
Out[11]: 4x4 Array{Any,2}:  
          "n"      "l_node"      "r_node"      "mean"  
    50.0    20.5518    21.4518    21.0018  
   500.0    20.955     21.045     21.0  
  5000.0    20.9955    21.0045    21.0
```

### Exercise 3.

The manager of a supermarket has to decide each day how much bread to buy. For every bread that it is sold, there is a profit of 60 cents. However, for every one not sold, there is a loss of 40 cents. The demand is uniformly distributed in the interval [80, 140]. How much bread should the manager order, in order to maximize the profit? What if the demand follows the Gamma distribution, with parameters  $k = 2$  and  $\theta = 55$ ?

Formalizing the problem. Let  $P$  denote the profit. Then:

$$P = \begin{cases} 0.6Q, & \text{if } D \geq Q \\ 0.6D - 0.4(Q - D), & \text{if } D < Q \end{cases}$$

Uniform distribution - analytical solution:

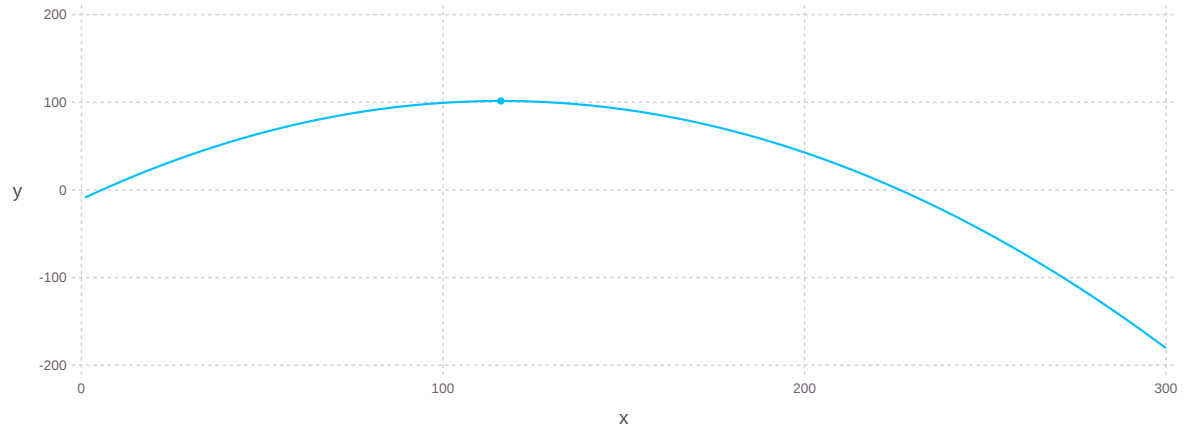
$$\max_Q E[P] = \max_Q \left[ \int_{80}^Q [0.6x - 0.4(Q - x)] \frac{1}{60} dx + \int_Q^{140} 0.6Q \frac{1}{60} dx \right]$$

$$Q^* = 116 \quad E[P] = 101.47$$

```
In [12]: function ex3_Uniform(Q)
          return (-1/120)Q^2+(29/15)Q-32/3
        end

        plot(layer(x=1:300, y=ex3_Uniform.(1:300), Geom.line),
              layer(x=[116], y=[101.47], Geom.point)
        )
```

Out[12]:



Gamma distribution - analytical formulation:

$$\max_Q E[P] = \max_Q \left[ \int_{80}^Q [0.6x - 0.4(Q - x)] \frac{1}{\Gamma(k)\theta^k} x^{k-1} dx + \int_Q^{140} 0.6Q \frac{1}{\Gamma(k)\theta^k} x^{k-1} dx \right]$$

Gamma distribution - Numerical approximation:

```

In [13]: function ex3_MonteCarlo()

    k = 2;  $\theta$  = 55

    results = ["n" "Optimal Q" "E[P]"]

    for n in [5 500 5000]
        Q = quantile.(Gamma(k, $\theta$ ), rand(n))

        gQ = pdf.(Gamma(k, $\theta$ ),Q)
        P = zeros(n)
        D = rand(Gamma(k, $\theta$ ),n)
        mP = zeros(n)

        for i in 1:n
            for j in 1:n
                if D[j] >= Q[i]
                    P[j] = 0.6*Q[i]
                else
                    P[j] = 0.6*D[j]-0.5(Q[i]-D[j])
                end
            end
            mP[i] = mean(P)
        end

        Pfunc = [Q mP]
        Q_EP = []
        max_pair = Pfunc[findfirst(Pfunc[:,2], maximum(Pfunc[:,2])),:]

        results = [results; n max_pair[1] max_pair[2]]

    end

    return results
end;

```

```

In [14]: ex3_MonteCarlo()

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Out[14]: 4×3 Array{Any,2}:
           "n"      "Optimal Q"      "E[P]"
    5.0      69.3069      17.9061
   500.0     98.7182      32.4981
  5000.0    101.292      33.3719

```