

# Nonlinear Optimization Exercise Session 5

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```
In [ ]: using LinearAlgebra
        using Plots
```

## 33. Variations of the Gradient Algorithm

Solve the minimization problem

$$\min_{x \in \mathbb{R}} f(x)$$

where

$$f: \mathbb{R}^n \rightarrow \mathbb{R}, f(x) = \begin{cases} \frac{1}{c} \|x\| - \frac{1}{2c^2}, & \text{if } \|x\| \geq \frac{1}{c} \\ \frac{1}{c} \|x\|^2, & \text{otherwise} \end{cases},$$

with the gradient algorithm and the fast gradient algorithm

- by considering different values for the dimension  $n \in \{1, 10, 50, 500, 5000\}$  and for the parameter  $c > 0$
- by using the Nesterov rule and the Chambolle-Dossal rule for the momentum parameters
- by using different values for the starting point  $x^0$

Run the two algorithms with step size  $\gamma = \frac{1}{L_{\nabla f}}$  for 200 iterations and display  $(\|x^k\|)_{k=0, \dots, 200}$  and  $(f(x^k))_{k=0, \dots, 200}$  as functions of the number of iterations  $k$ .

```

In [ ]: function gradientAlgorithmArmijo(
    f::Function,
    g::Function,
    x0::Vector,
    β::Real,
    σ::Real,
    k::Integer
)

    @assert k > 0 "Number of iterations is not positive, got $k."

    x = [x0]
    d = -g(x[1])

    for _ ∈ 1:k
        fx = f(x[end])
        σnorm = σ * norm(d)
        k = 0
        t = 1

        # iterate exponent k until Armijo condition is met
        while f(x[end] + t * d) > fx - t * σnorm
            k = k + 1
            t = β^k
        end

        push!(x, x[end] + t * d)
        d = -g(x[end])
    end

    return x
end
;

```

```

In [ ]: function gradientAlgorithmNesterov(
    f::Function,
    g::Function,
    x0::Vector,
    γ::Real,
    k::Integer
)

    @assert k > 0 "Number of iterations is not positive, got $k."

    x = [x0, x0]
    t_old = 1
    d = -g(x[1])

    for _ ∈ 2:k
        t_new = (1 + sqrt(4 * t_old^2 + 1)) / 2
        y = x[end] + (t_old - 1) / t_new * (x[end] - x[end-1])
        push!(x, y - γ * g(y))
        d = -g(x[end])
        t_old = t_new
    end

    return x
end
;

```

```

In [ ]: function gradientAlgorithmChambolleDossal(
    f::Function,
    g::Function,
    x0::Vector,
    γ::Real,
    α::Real,
    k::Integer
)

    @assert k > 0 "Number of iterations is not positive, got $k."

    x = [x0, x0]
    t_old = 1
    d = -g(x[1])

    for _ ∈ 2:k
        t_new = (length(x) + α - 2) / (α - 1)
        y = x[end] + (t_old - 1) / t_new * (x[end] - x[end-1])
        push!(x, y - γ * g(y))
        d = -g(x[end])
        t_old = t_new
    end

    return x
end
;

In [ ]: function plotIterations(ns::Vector{Int}, c::Real, β::Real, σ::Real, α::Real, k::Integer)
    function f(x::AbstractVector)
        nx = norm(x)
        return nx >= 1/c ? nx/c - 1/(2*c^2) : nx^2 / 2
    end

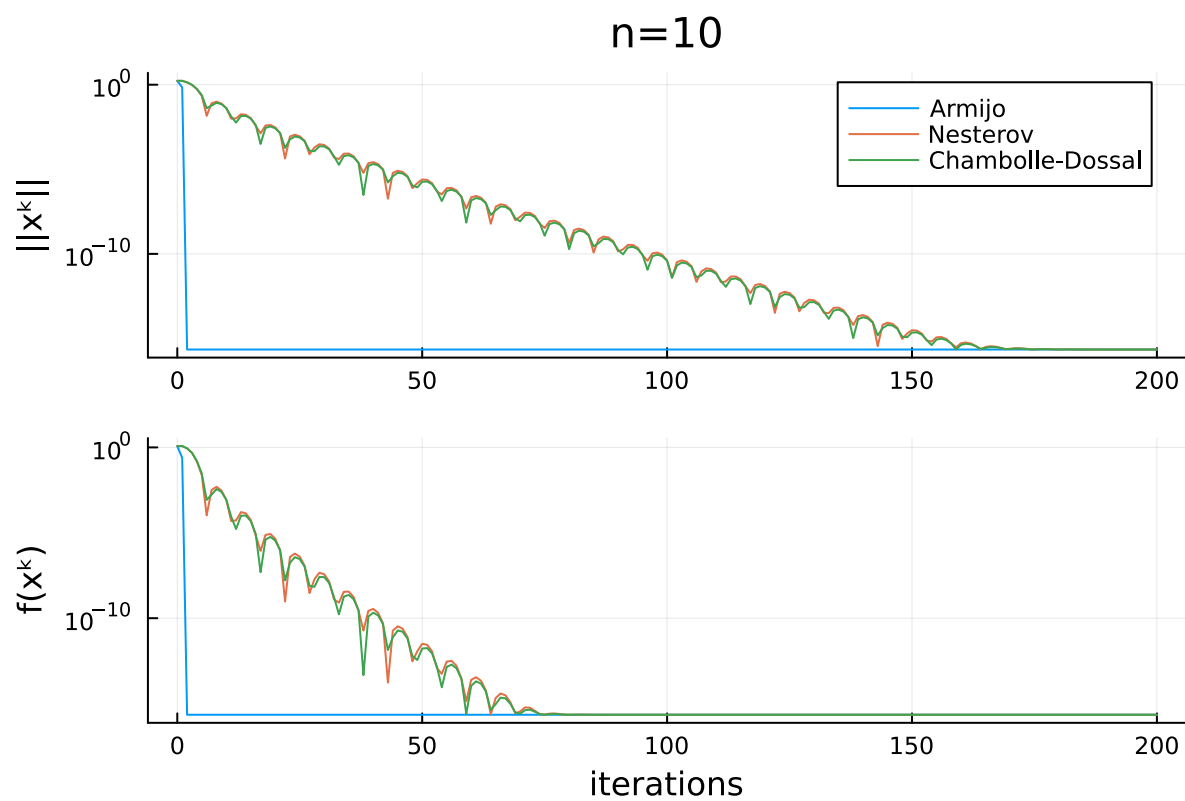
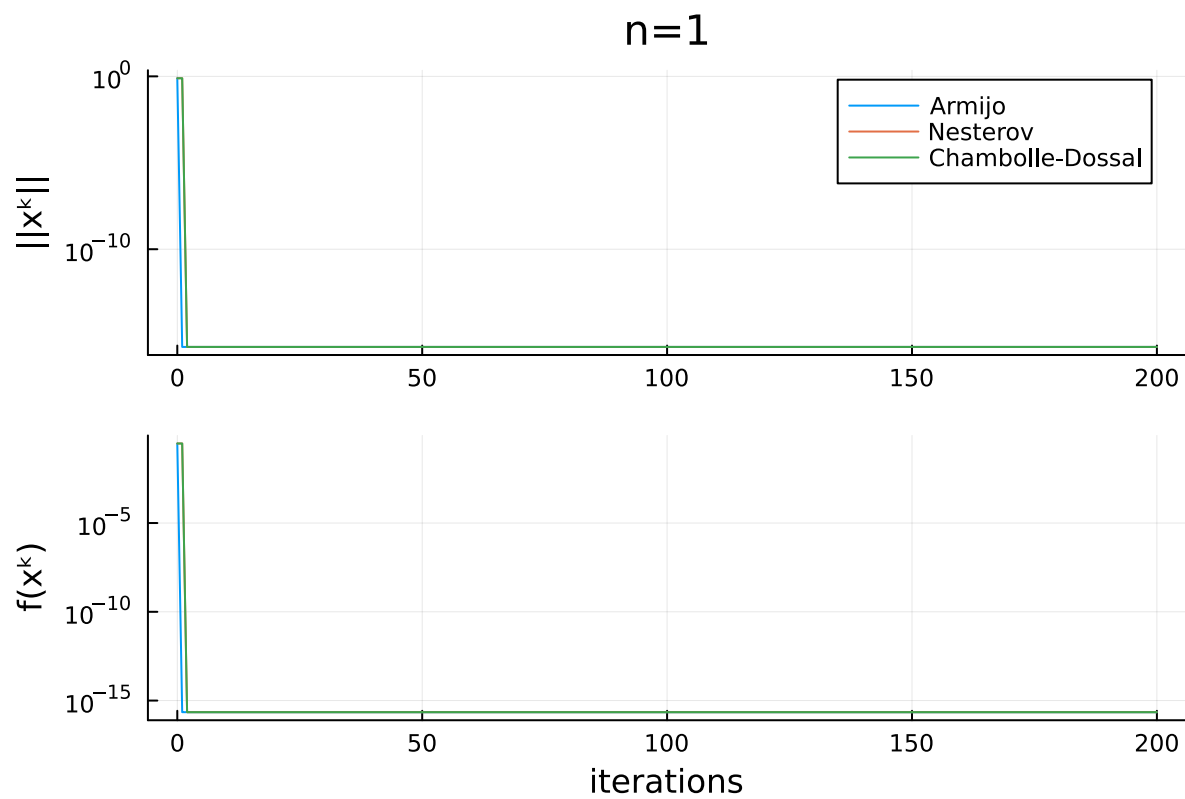
    function g(x::AbstractVector)
        nx = norm(x)
        nx >= 1/c ? x / nx / c : x
    end

    for n ∈ ns
        x = 2*rand(n) .- 1
        γ = 1 / sqrt(n) # 1 / Lipschitz constant
        Armijo = gradientAlgorithmArmijo(f, g, x, β, σ, k)
        Nesterov = gradientAlgorithmNesterov(f, g, x, γ, k)
        ChambolleDossal = gradientAlgorithmChambolleDossal(f, g, x, γ, α, k)
        norms = plot(
            0:k,
            [norm.(Armijo).+eps(), norm.(Nesterov).+eps(), norm.(ChambolleDossal).+eps()],
            yscale=:log10,
            title="n=$n",
            ylabel="||x^k||",
            label=["Armijo" "Nesterov" "Chambolle-Dossal"],
        );
        fs = plot(
            0:k,
            [f.(Armijo).+eps(), f.(Nesterov).+eps(), f.(ChambolleDossal).+eps()],
            yscale=:log10,
            xlabel="iterations",
            ylabel="f(x^k)",
            legend=false
        );
        display(plot(norms, fs, layout=(2,1)))
    end
end
;

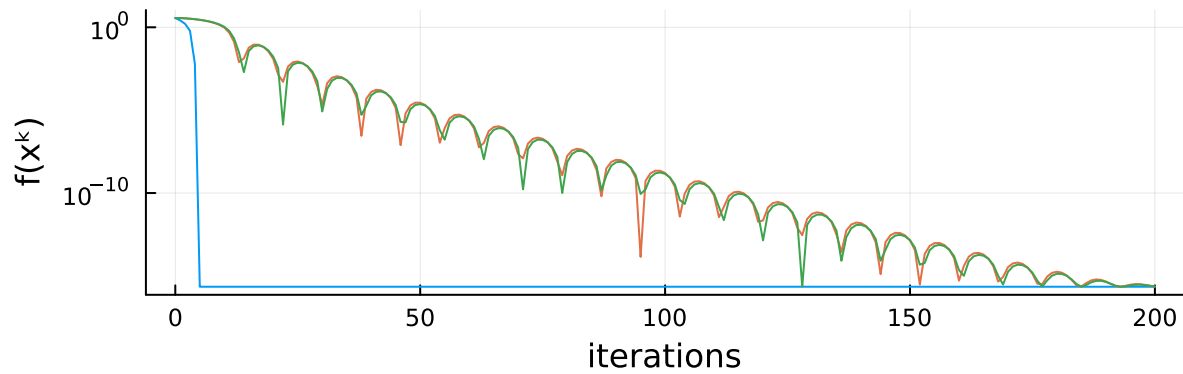
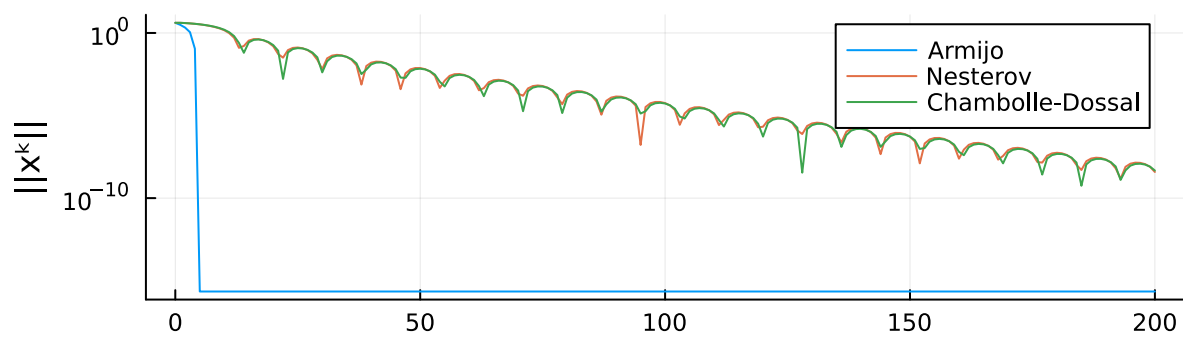
```

```
In [ ]: ns = [1, 10, 50, 500, 5000]
         $\beta$  = 0.5 # parameter of Armijo rule
         $\sigma$  = 0.01 # parameter of Armijo rule
         $\alpha$  = 3.01 # parameter of Chambolle-Dossal rule
        k = 200 # number of iterations
        ;
```

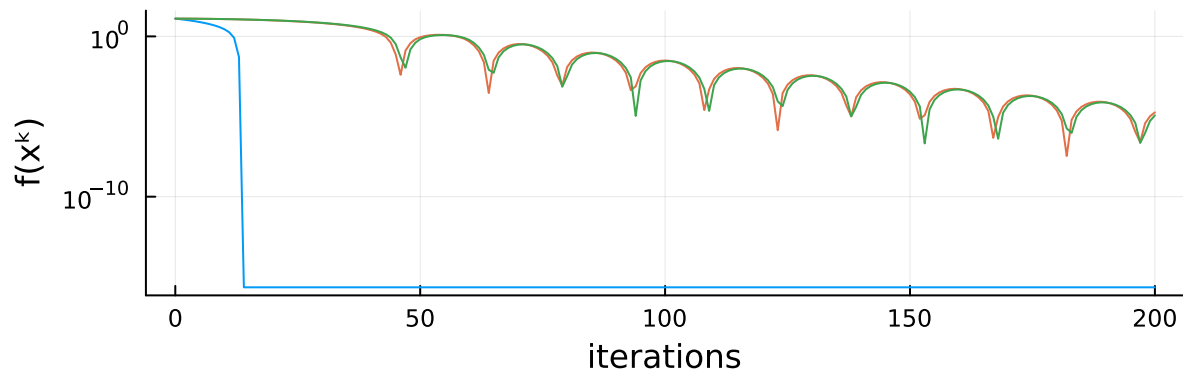
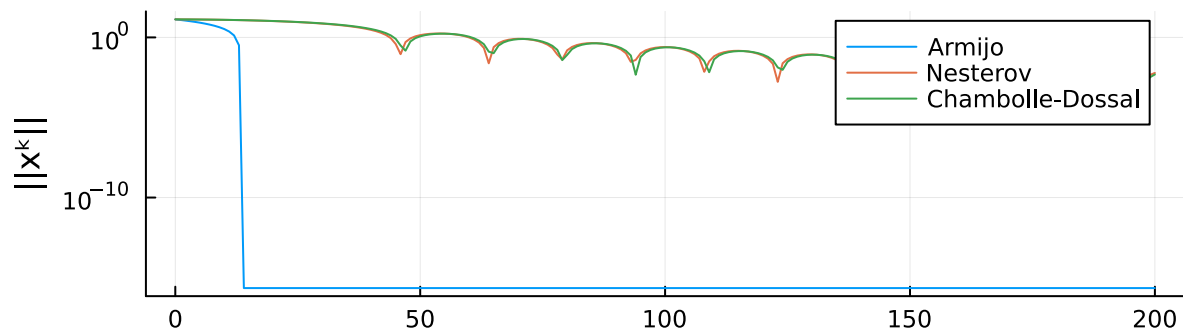
```
In [ ]: c = 1
        plotIterations(ns, c,  $\beta$ ,  $\sigma$ ,  $\alpha$ , k)
```



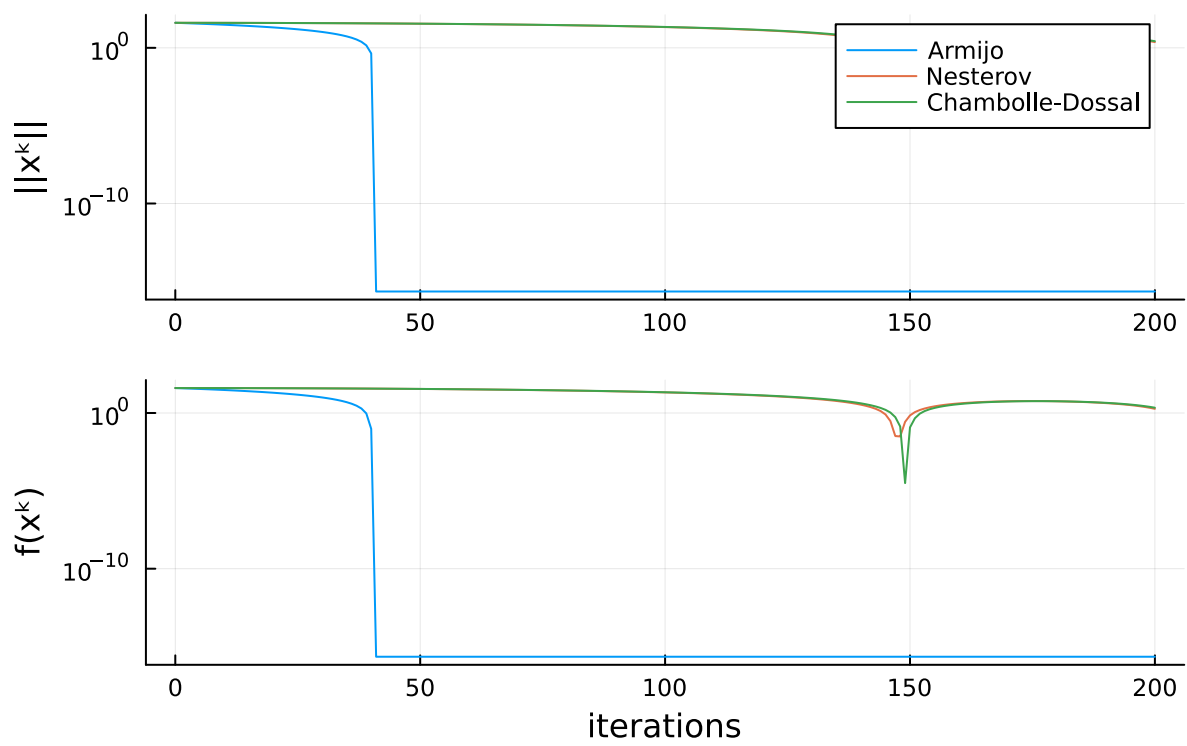
n=50



n=500

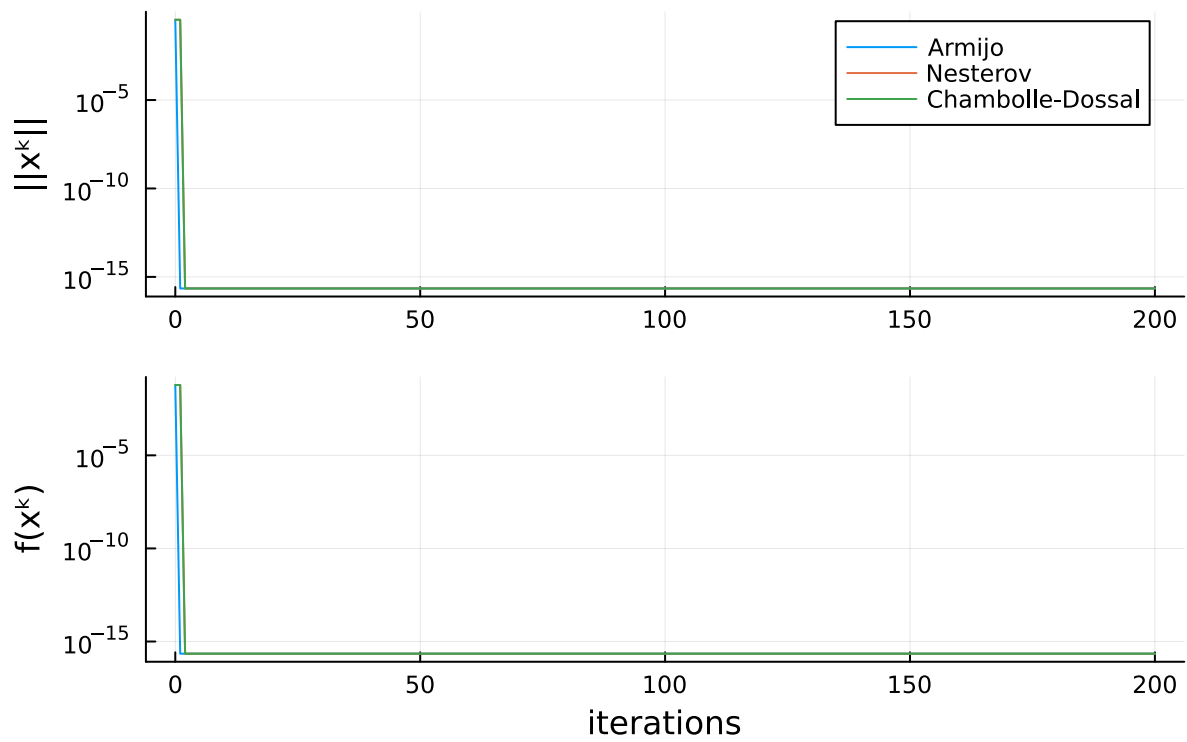


n=5000

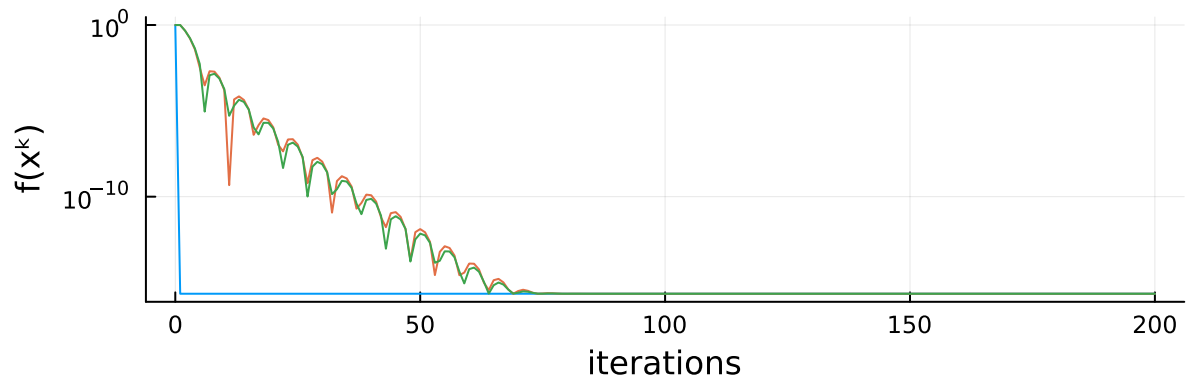
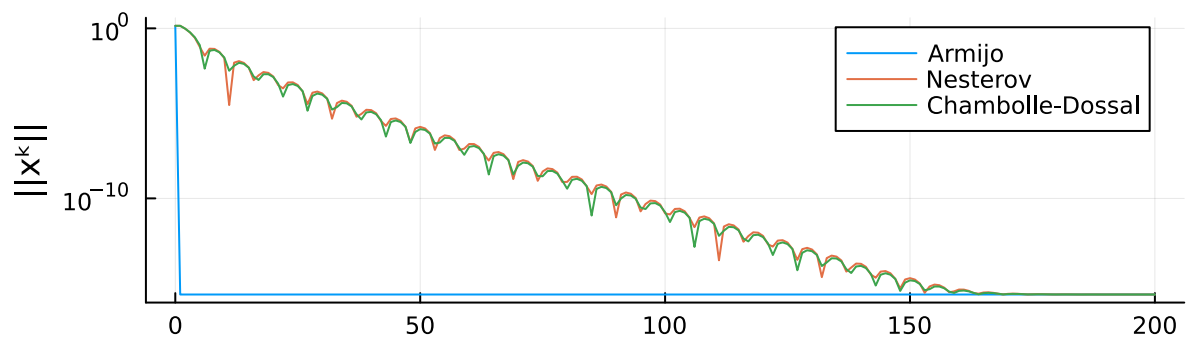


```
In [ ]: c = 0.01
plotIterations(ns, c,  $\beta$ ,  $\sigma$ ,  $\alpha$ , k)
```

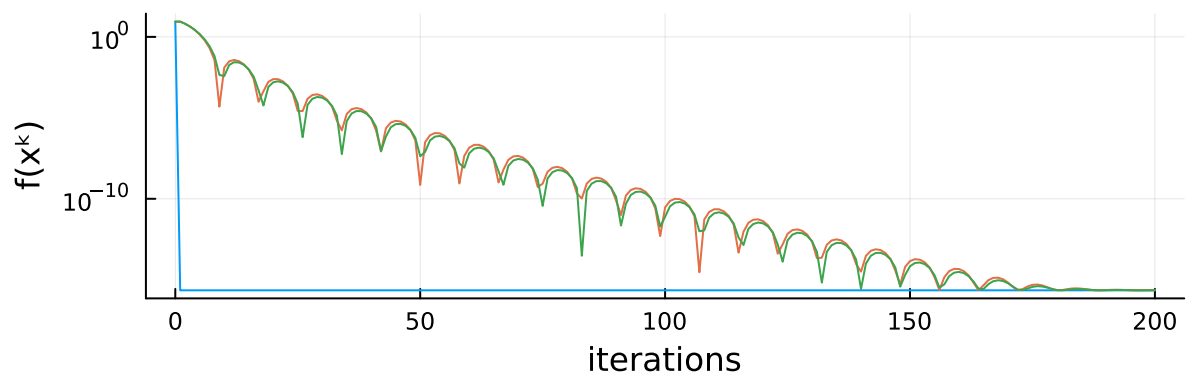
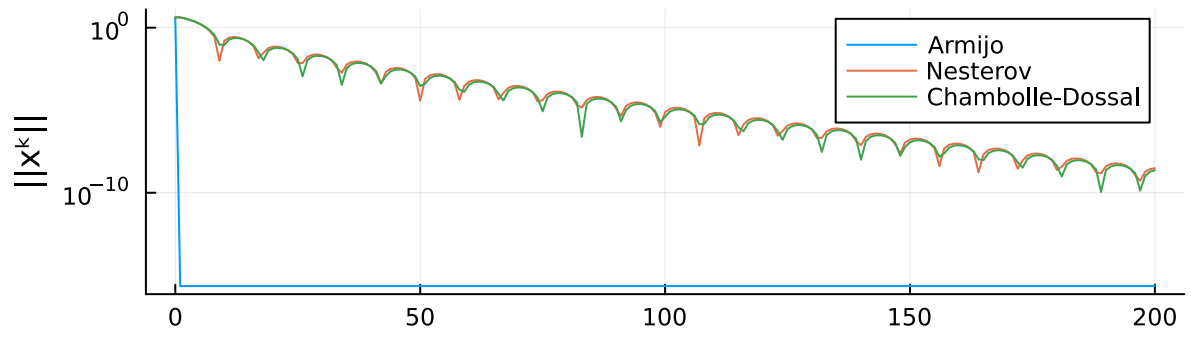
n=1



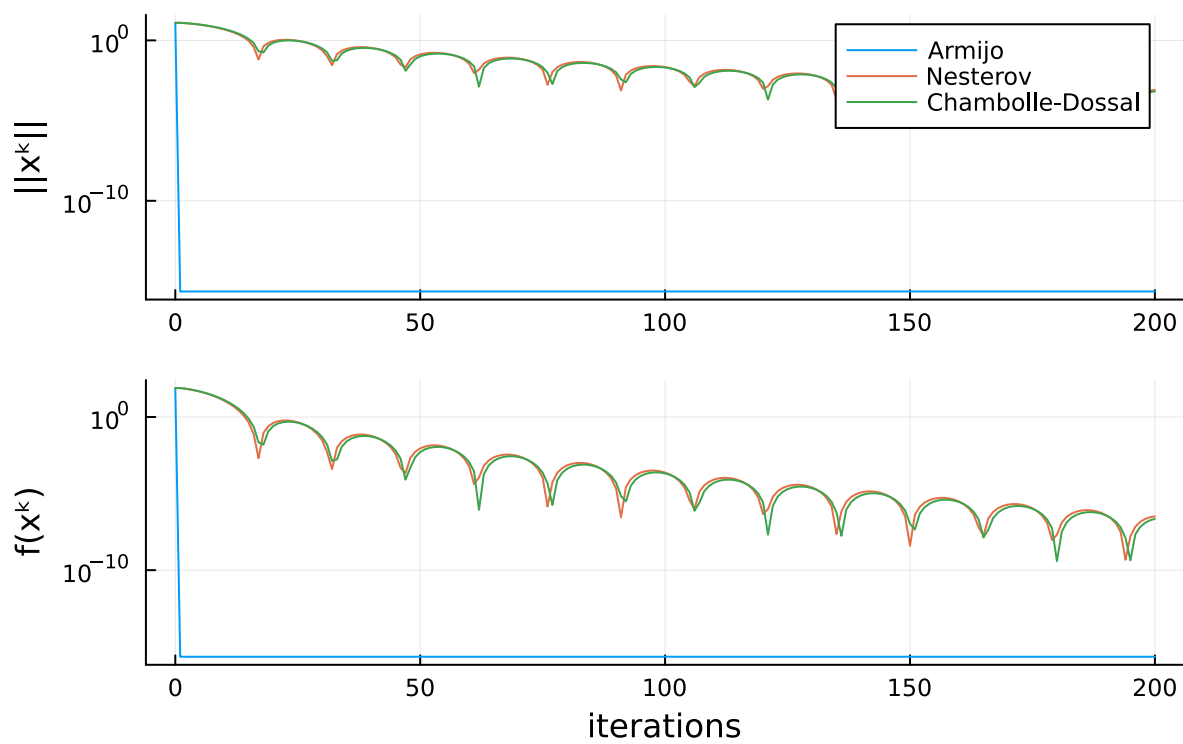
n=10



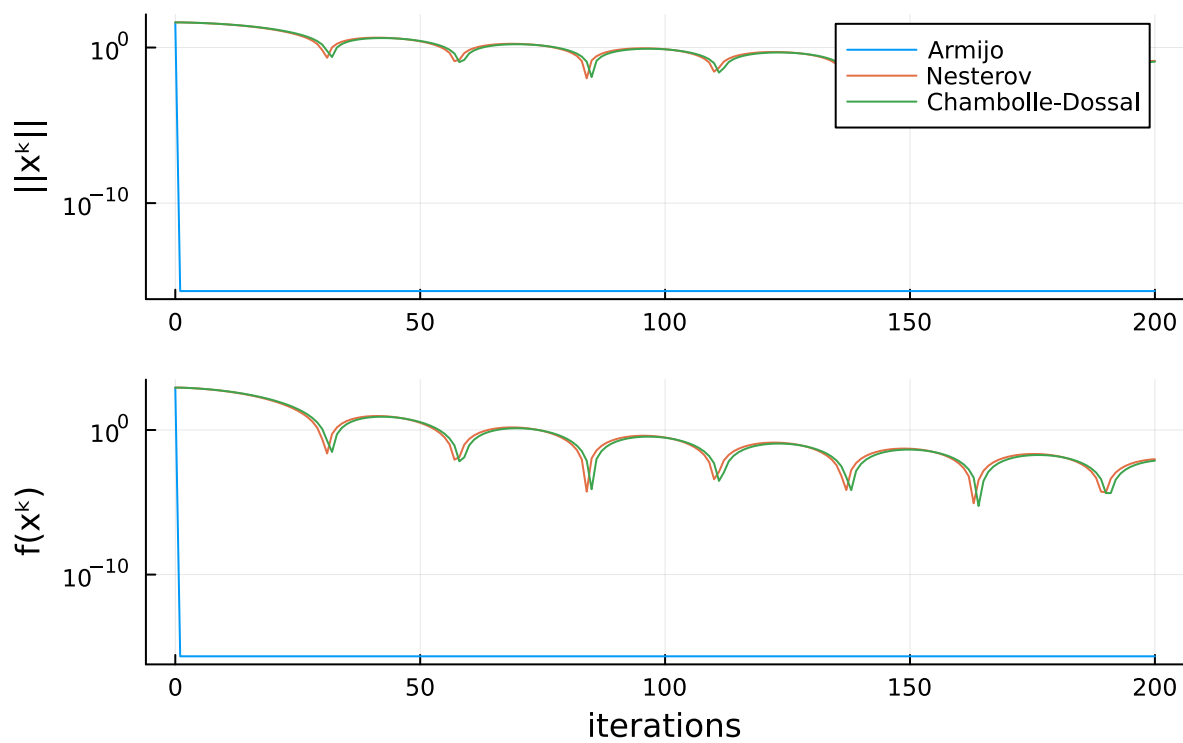
n=50



n=500



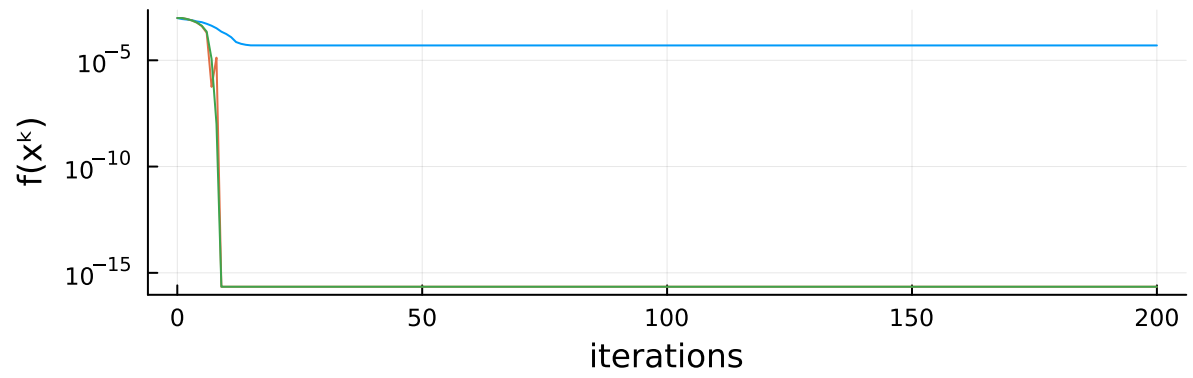
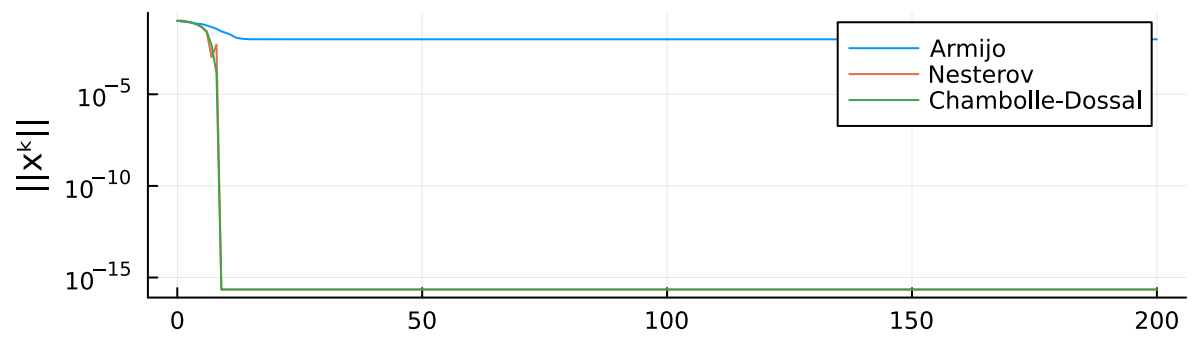
n=5000



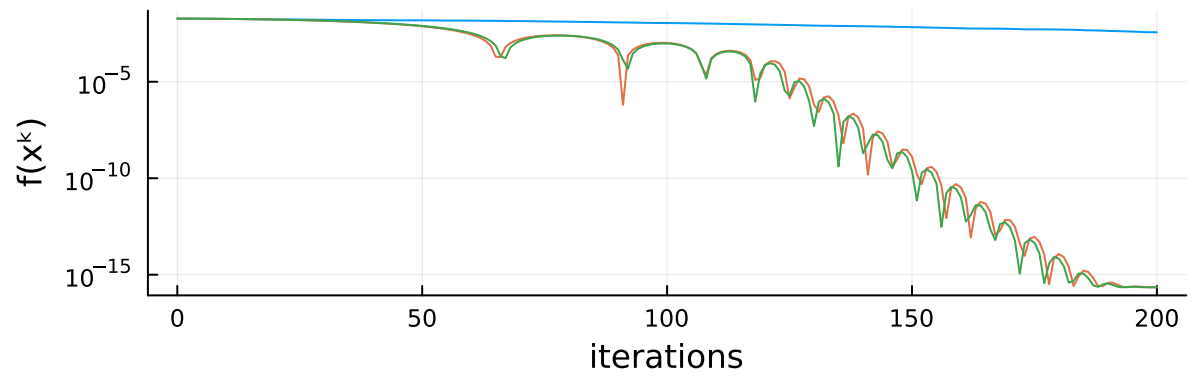
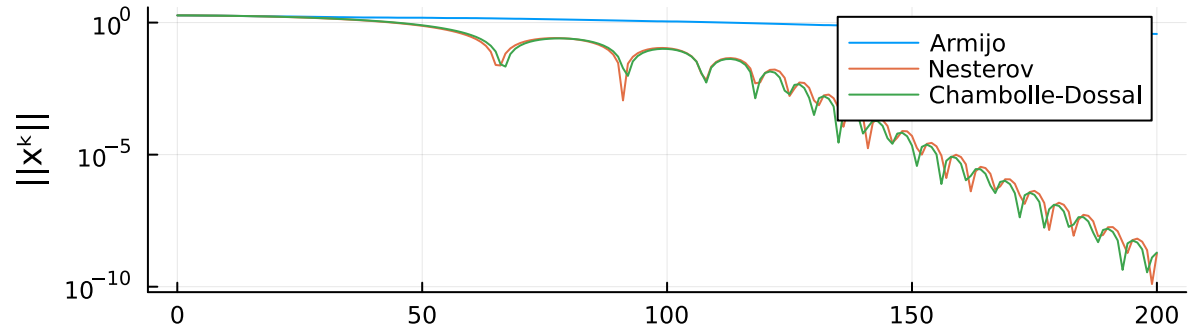
```
In [ ]: c = 100
plotIterations(ns, c,  $\beta$ ,  $\sigma$ ,  $\alpha$ , k)
```



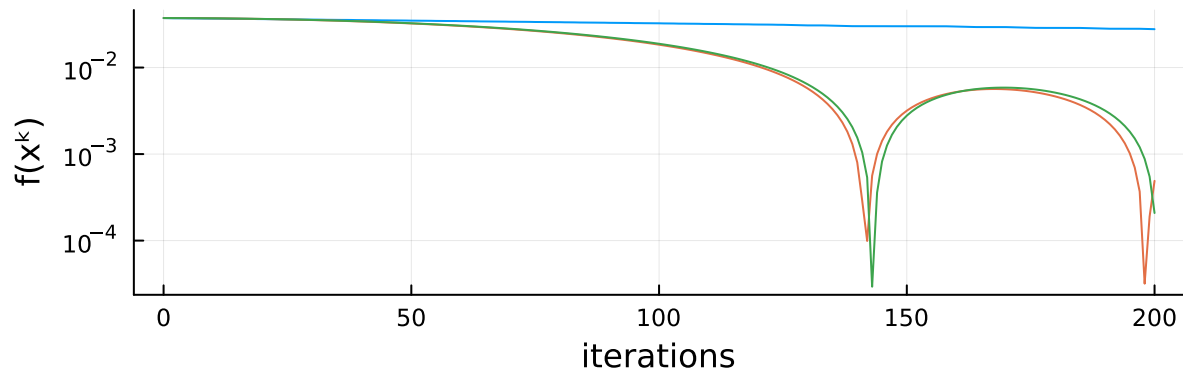
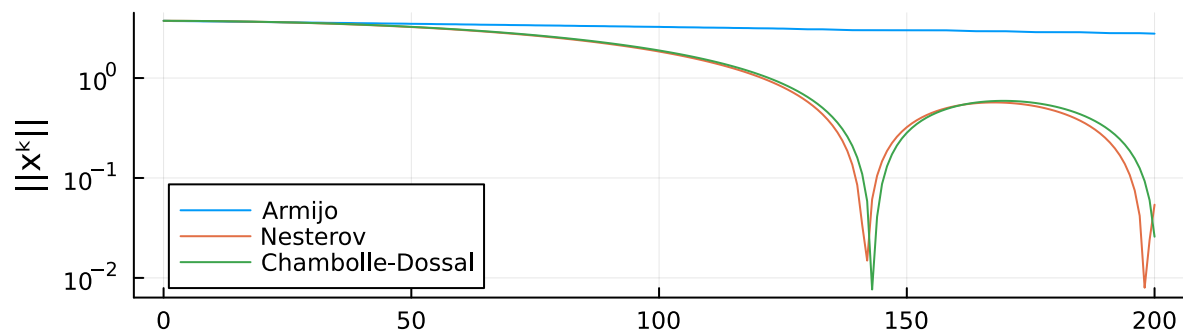
$n=1$



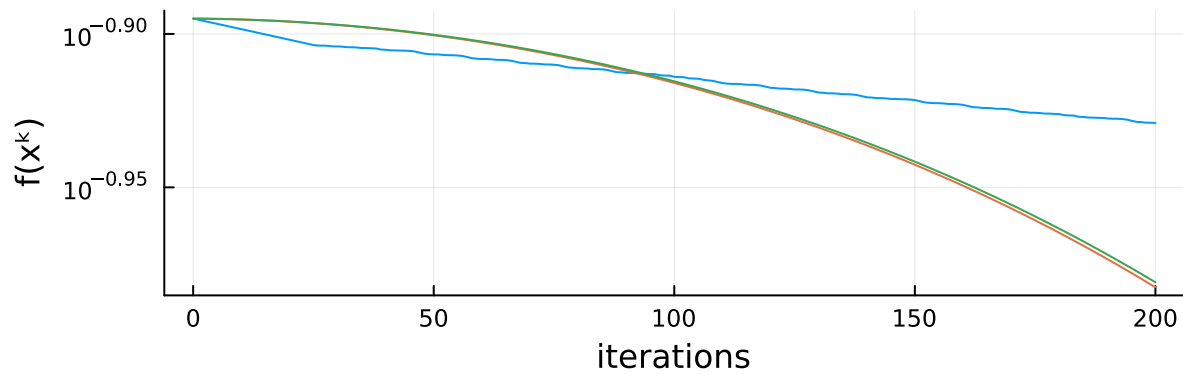
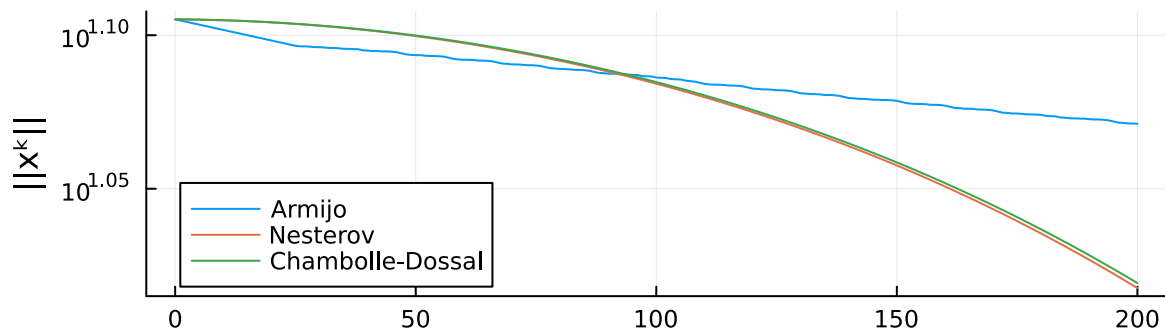
$n=10$



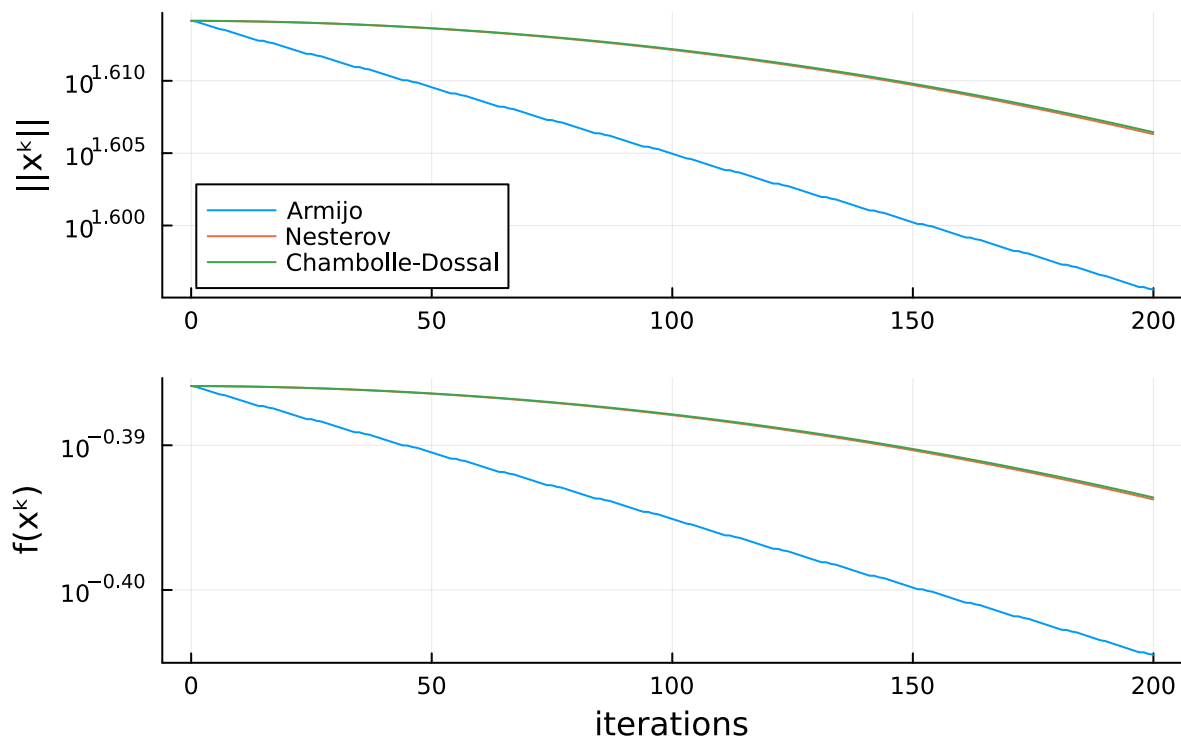
n=50



n=500



n=5000



### 34. Implementation of CG Algorithm

Implement the CG algorithm for linear systems (Algorithm 8.2). Use as input data the symmetric and positive definite matrix  $A$ , the vector  $b$ , the starting vector  $x^0$  and the parameter for the stopping criterion  $\varepsilon$ . The solution  $x^*$  and the number of performed iterations should be returned.

```
In [ ]: function CG(A::AbstractMatrix, b::AbstractVector, x0::AbstractVector, tol::Real)

    n = size(A, 1)

    @assert size(A, 2) == n "Matrix has to be square, got $n by $(size(A, 2))."
    @assert size(b, 1) == n "System vector has to be of size $n, got $(size(b, 1))."
    @assert size(x0, 1) == n "Starting vector has to be of size $n, got $(size(x0, 1))"
    @assert tol >= 0 "Tolerance has to be greater than zero, got $tol."

    x_old = x0
    g_old = A * x_old - b
    norm_old = norm(g_old)^2
    d_old = -g_old
    tol = tol^2
    n_iterations = 0

    while norm_old > tol && n_iterations < n
        # calculate new iterates
        Ad = A * d_old
        t = norm_old / dot(d_old, Ad)
        x_new = x_old + t * d_old
        g_new = g_old + t * Ad
        norm_new = norm(g_new)^2
        β = norm_new / norm_old
        d_new = -g_new + β * d_old

        # update old variables
        x_old = x_new
        g_old = g_new
        norm_old = norm_new
        d_old = d_new
        n_iterations += 1
    end

    return x_old, n_iterations
end
;
```

Test the algorithm on the following optimization problems and input data values:

(a) Minimize the function

$$f(x_1, x_2) = 2x_1^2 + x_2^2 - 4x_1 - 2x_2 + 3$$

over  $\mathbb{R}^2$ , for  $x^0 = (5, -5)^\top$  and  $\varepsilon = 10^{-3}$ .

```
In [ ]: A = [4 0; 0 2]
b = [4; 2]
x0 = [5; -5]
tol = 1e-3
;
```

```
In [ ]: x, n_iterations = CG(A, b, x0, tol)

([1.0, 1.0000000000000004], 2)
```

```
In [ ]: # optimality condition
all(isapprox.(A * x - b, 0, atol=1e-8))

true
```

```
In [ ]: # minimum value
dot(x, A, x) / 2 - dot(b, x) + 3

0.0
```

(b) Minimize the function

$$f(x_1, x_2, x_3) = x_1^2 + 0.3x_1x_2 + 0.975x_2^2 + 0.01x_1x_3 + x_3^2 + 3x_1 - 4x_2 + x_3$$

over  $\mathbb{R}^3$ , for  $x^0 = (0, 0, 0)^\top$  and  $\varepsilon = 10^{-8}$ .

```
In [ ]: A = [2 0.3 0.01; 0.3 0.975*2 0; 0.01 0 2]
b = [-3; 4; -1]
x0 = [0; 0; 0]
tol = 1e-8
;
```

```
In [ ]: x, n_iterations = CG(A, b, x0, tol)

([-1.8478819339864998, 2.3355715795876666, -0.4907605903300675], 3)
```

```
In [ ]: # optimality condition
all(isapprox.(A * x - b, 0, atol=1e-8))

true
```

```
In [ ]: # minimum value
dot(x, A, x) / 2 - dot(b, x)

-4.688346355320117
```

```
In [ ]:
```