

Problem 2(a)

```
In [1]: using ProgressMeter
using LinearAlgebra
using MAT
using Plots; pyplot();

In [2]: vars = matread("spam_data.mat")
X_train = log.(vars["Xtrain"] .+ 0.1)
X_test = log.(vars["Xtest"] .+ 0.1)
y_train = vec(vars["ytrain"])
y_test = vec(vars["ytest"])

y_train[y_train .== 0] .= -1
y_test[y_test .== 0] .= -1

n, p = size(vars["Xtrain"]);

In [3]: O(a) = 1 / (1 + exp(-a))
g(w, X, y) = sum(O(y[i]*w'*X[i,:]) - 1)*y[i]*X[i,:] for i in 1:size(y)[1]
H(w, X, y) = sum(O(y[i]*w'*X[i,:])*(1 - O(y[i]*w'*X[i,:]))*X[i,:]*X[i,:]) for i in 1:size(y)[1]
loss(w, X, y) = sum(log.(1 .+ exp.(-y[i]*w'*X[i,:])) for i in 1:size(y)[1])[1];

In [4]: function gradient_descent(g, loss, X, y, x0; α=1E0, max_iter=1000, ε=1E0)
    tol = Inf
    hist, l_hist = [], []
    push!(hist, x0), push!(l_hist, loss(x0, X, y))

    for k in 1:max_iter
        tol < ε ? break : nothing;

        xk = hist[end]
        gk = g(xk, X, y)

        xk1 = xk - α*gk

        push!(hist, xk1)
        push!(l_hist, loss(xk1, X, y))

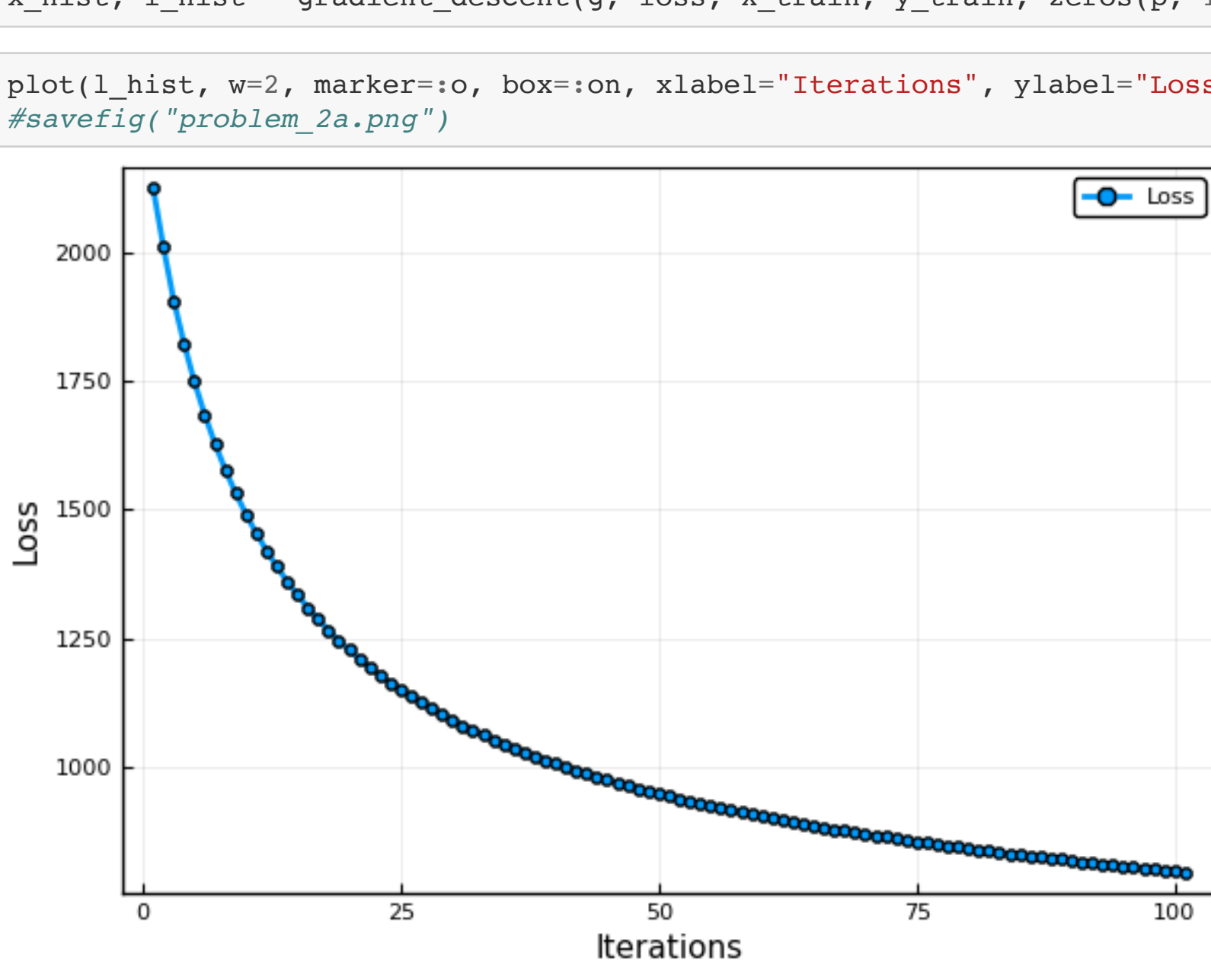
        tol = norm(g(xk1, X, y))
    end

    return hist, l_hist
end;

In [5]: x_hist, l_hist = gradient_descent(g, loss, X_train, y_train, zeros(p, 1), α=1E-5, max_iter=100);

In [6]: plot(l_hist, w=2, marker=:o, box=:on, xlabel="Iterations", ylabel="Loss", label="Loss")
savefig("problem_2a.png")

Out[6]:
```



Problem 2(b)

```
In [7]: function gradient_descent_w_momentum(g, loss, X, y, x0; α=1E0, β=0.9, max_iter=1000, ε=1E0)
    tol = Inf
    v0 = zeros(size(x0))
    v_hist, x_hist, l_hist = [], [], []
    push!(v_hist, v0), push!(x_hist, x0), push!(l_hist, loss(x0, X, y))

    for k in 1:max_iter
        tol < ε ? break : nothing;

        vk = v_hist[end]
        xk = x_hist[end]
        gk = g(xk, X, y)

        vk1 = β*vk + α*gk
        xk1 = xk - vk1

        push!(v_hist, vk1)
        push!(x_hist, xk1)
        push!(l_hist, loss(xk1, X, y))

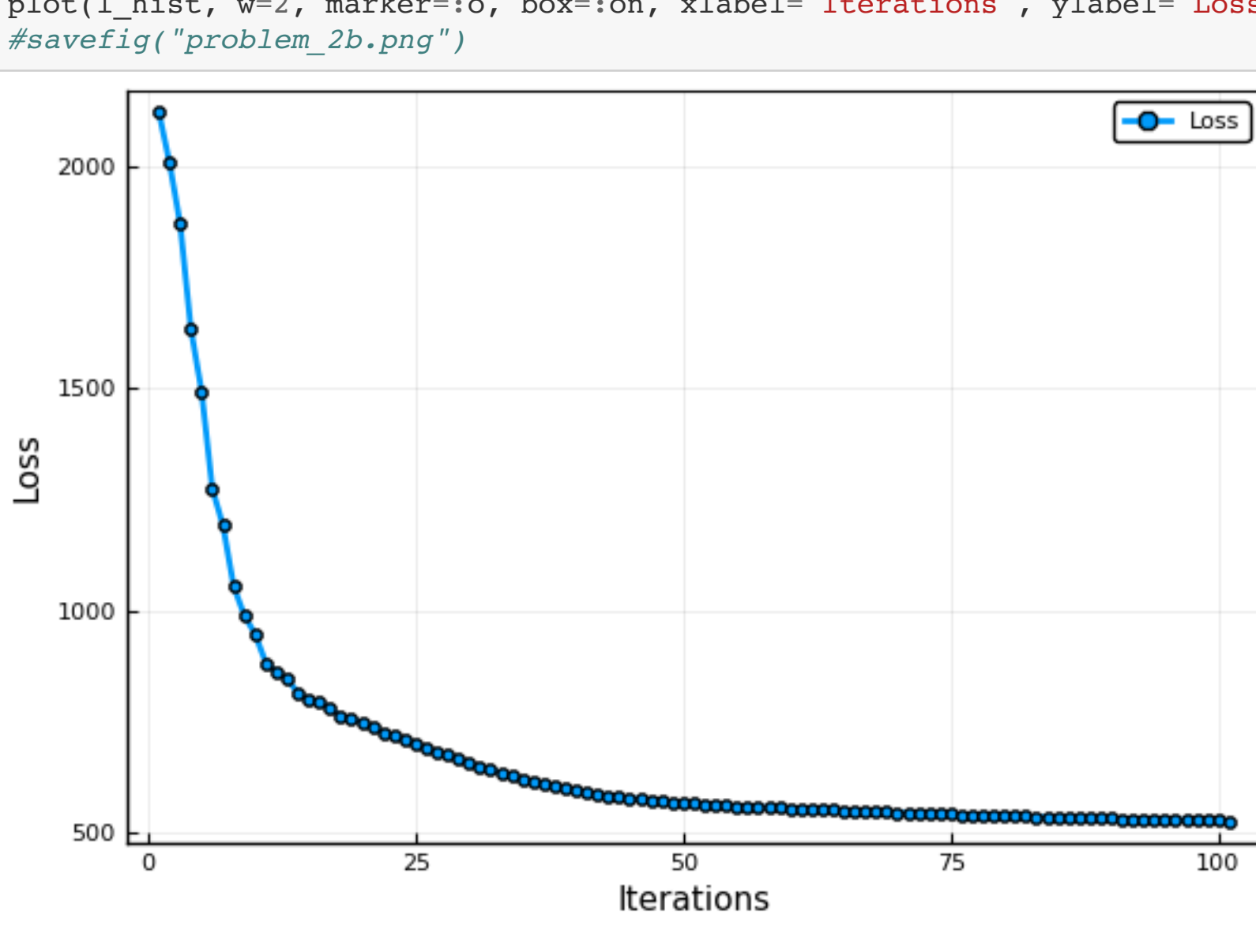
        tol = norm(g(xk1, X, y))
    end

    return x_hist, l_hist
end;

In [8]: x_hist, l_hist = gradient_descent_w_momentum(g, loss, X_train, y_train, zeros(p, 1), α=1E-5, β=0.9, max_iter=100);

In [9]: plot(l_hist, w=2, marker=:o, box=:on, xlabel="Iterations", ylabel="Loss", label="Loss")
savefig("problem_2b.png")

Out[9]:
```



Problem 2(c)

Armijo line search

Problem 4(a)

```
In [11]: function newtons_method(g, H, loss, X, y, x0; α=1E0, max_iter=100, ε=1E0)
    tol = Inf
    hist, l_hist = [], []
    push!(hist, x0), push!(l_hist, loss(x0, X, y))

    for i in 1:max_iter
        tol < ε ? break : nothing;

        xk = hist[end]
        gk = g(xk, X, y)
        Hk = H(xk, X, y)

        xk1 = xk - α*inv(Hk)*gk

        push!(hist, xk1)
        push!(l_hist, loss(xk1, X, y))

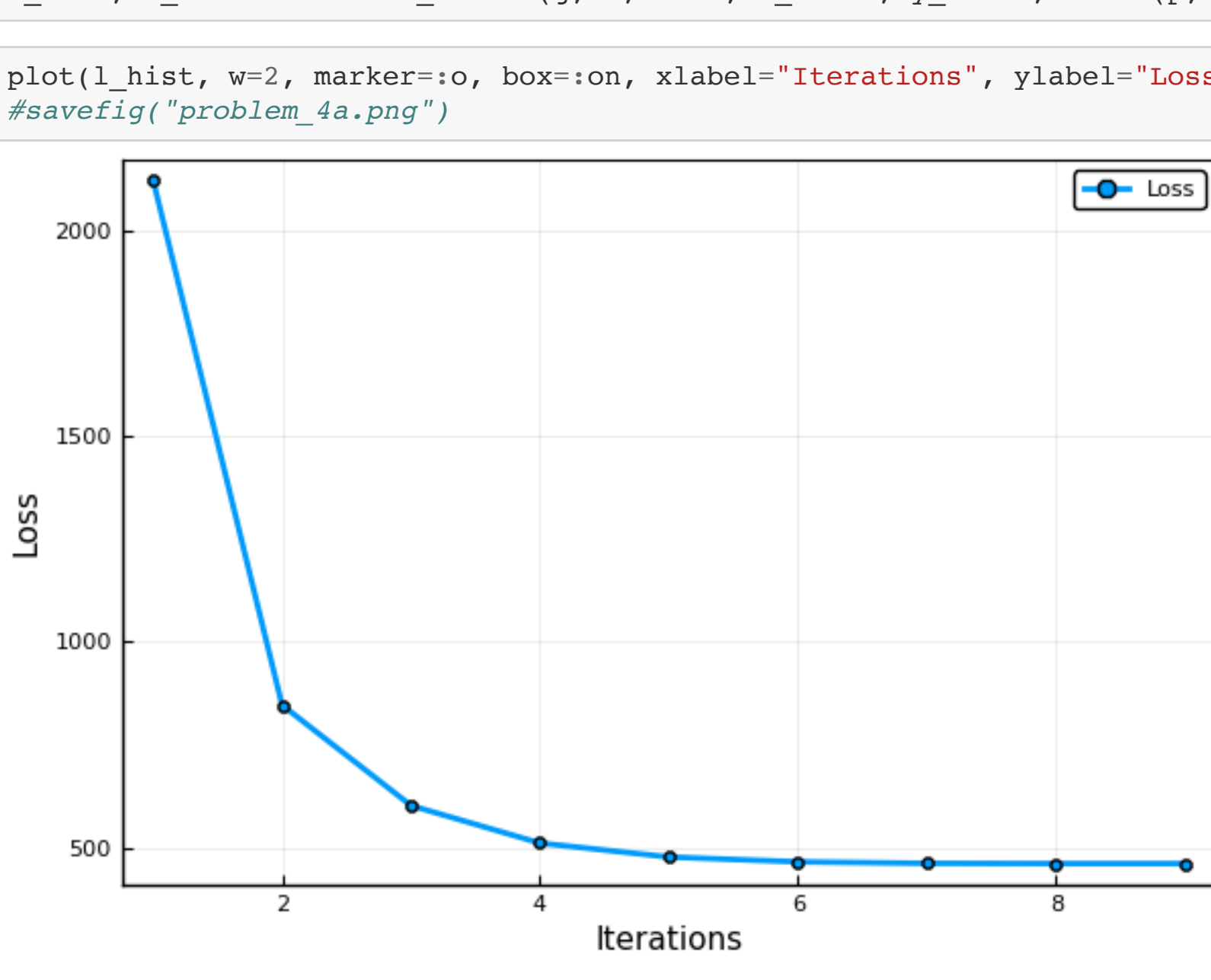
        tol = norm(g(xk1, X, y))
    end

    return hist, l_hist
end;

In [12]: x_hist, l_hist = newtons_method(g, H, loss, X_train, y_train, zeros(p, 1), α=1);

In [13]: plot(l_hist, w=2, marker=:o, box=:on, xlabel="Iterations", ylabel="Loss", label="Loss")
savefig("problem_4a.png")

Out[13]:
```



Problem 4(b)

```
In [14]: Asqrt(w, X, y) = H(w, X, y)^(1/2);

In [15]: function randomized_newtons_method(g, A, d, loss, X, y, x0; α=1E0, max_iter=100, ε=1E0)
    tol = Inf
    hist, l_hist = [], []
    push!(hist, x0), push!(l_hist, loss(x0, X, y))

    for i in 1:max_iter
        tol < ε ? break : nothing;

        xk = hist[end]
        gk = g(xk, X, y)
        Ak = A(xk, X, y)

        n = size(Ak)[1]
        Sk = zeros(d, n)
        for i in 1:d
            Sk[i, rand(1:n...)] = 1/sqrt(n/d)
        end

        Hk = Ak'*Sk'*Sk*Ak

        xk1 = xk - α*inv(Hk)*gk

        push!(hist, xk1)
        push!(l_hist, loss(xk1, X, y))

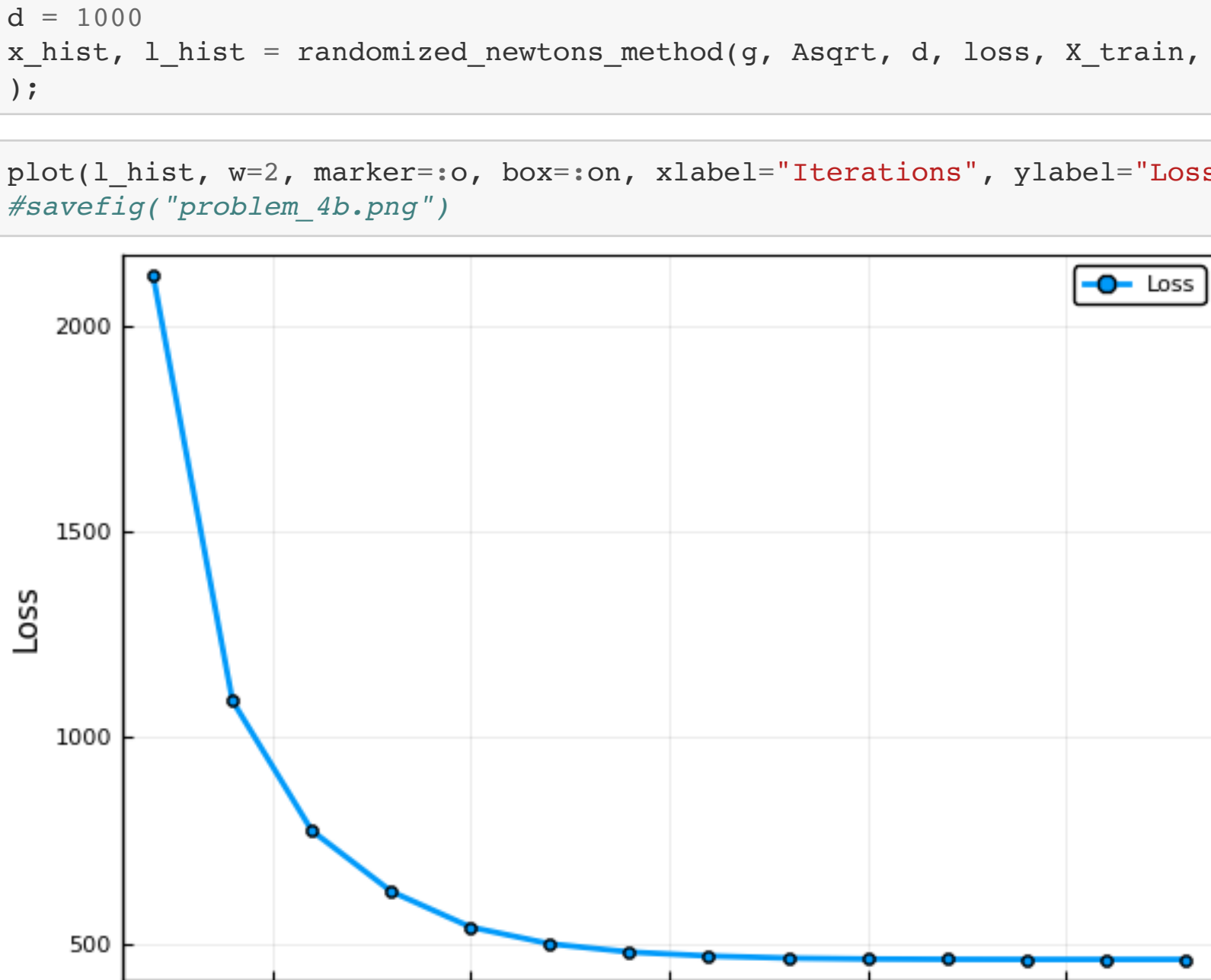
        tol = norm(g(xk1, X, y))
    end

    return hist, l_hist
end;

In [16]: d = 1000
x_hist, l_hist = randomized_newtons_method(g, Asqrt, d, loss, X_train, y_train, zeros(p, 1), α=2E2, ε=1E0, max_iter=50);

In [17]: plot(l_hist, w=2, marker=:o, box=:on, xlabel="Iterations", ylabel="Loss", label="Loss")
savefig("problem_4b.png")

Out[17]:
```



Problem 5(a)

```
In [18]: m = 128
n = 1024;

In [19]: function hadamard(d)
    if d == 1
        return [[1, 1] [1, -1]]
    else
        return hcat(vcat(hadamard(d-1), hadamard(d-1)), vcat(hadamard(d-1), -hadamard(d-1)))
    end
end

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

In [20]: function fjlt(d, m, n)
    H = hadamard(d)
    D = diagm([rand([-1, +1]) for _ in 1:n])
    P = zeros(m, n)
    for i in 1:m
        j = rand(1:n...)
        P[i, j] = 1
    end
    P' *= sqrt(n/m)

    S = 1/sqrt(n)*P'*H*D;

    return S
end

Out[20]: fjlt (generic function with 1 method)

In [21]: S = fjlt(10, m, n);
```

Problem 5(b)

```
In [22]: @show norm(S'*S - I);

norm(S' * S - I) = 89.7997772825746
```

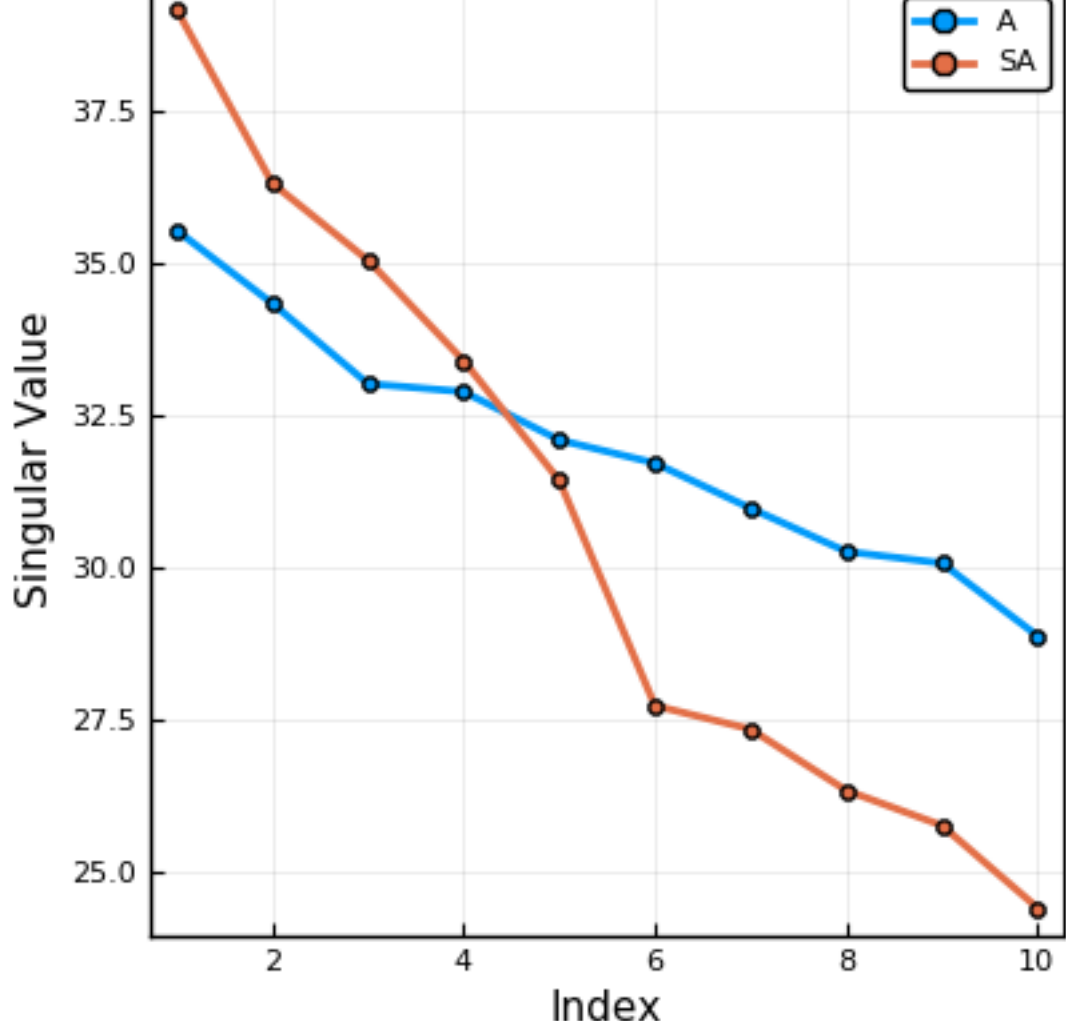
Problem 5(c)

```
In [23]: A = [randn() for i in 1:1024, j in 1:10];

In [24]: UA, ZA, VA = svd(A)
USA, ZSA, VSA = svd(S*A)

plot(ZA, w=2, marker=:o, label="A", box=:on, xlabel="Index", ylabel="Singular Value", size=(400,400))
savefig("problem_5c.png")

Out[24]:
```



Problem 5(d)

```
In [25]: function jl_epsilon(A, S)
    A_dists = zeros(10,10)
    SA_dists = zeros(10,10)

    for i in 1:10, j in 1:10
        A_dists[i, j] = norm(A[:, i] - A[:, j])^2
        SA_dists[i, j] = norm(S*A[:, i] - S*A[:, j])^2
    end

    ε = abs.(SA_dists ./ A_dists .- 1)
    ε[isnan.(ε)] .= 0

    return maximum(ε)
end;

In [26]: @show jl_epsilon(A, S);

jl_epsilon(A, S) = 0.29705755463730754

In [27]: using ProgressMeter
using Statistics

Es = []

@showprogress for i in 1:100
    S = fjlt(10, m, n)
    push!(Es, jl_epsilon(A, S))
end

@show minimum(Es)
@show mean(Es)
@show maximum(Es);

Progress: 100% ██████████ Time: 0:00:48

minimum(Es) = 0.17793509316963618
mean(Es) = 0.2960596802090196
maximum(Es) = 0.6668913151454343
```

Problem 5(e)

```
In [28]: A = [randn() for i in 1:1024, j in 1:10]
b = [randn() for i in 1:1024, j in 1:1]

x̂ = (A'*A)\A'*b

SA = S'A
Sb = S*b

x̂ = (SA'*SA)\SA'*Sb;

In [29]: @show norm(x̂ - x)
@show norm(A*x̂ - A*x)
@show norm(A*x̂ - b)^2 / norm(A*x̂ - b)^2;

norm(x̂ - x) = 0.2828462927729973
norm(A * x̂ - A * x) = 5.240542031042262
norm(A * x̂ - b) ^ 2 / norm(A * x̂ - b) ^ 2 = 1.0870096195456453
```