

# Problem 1

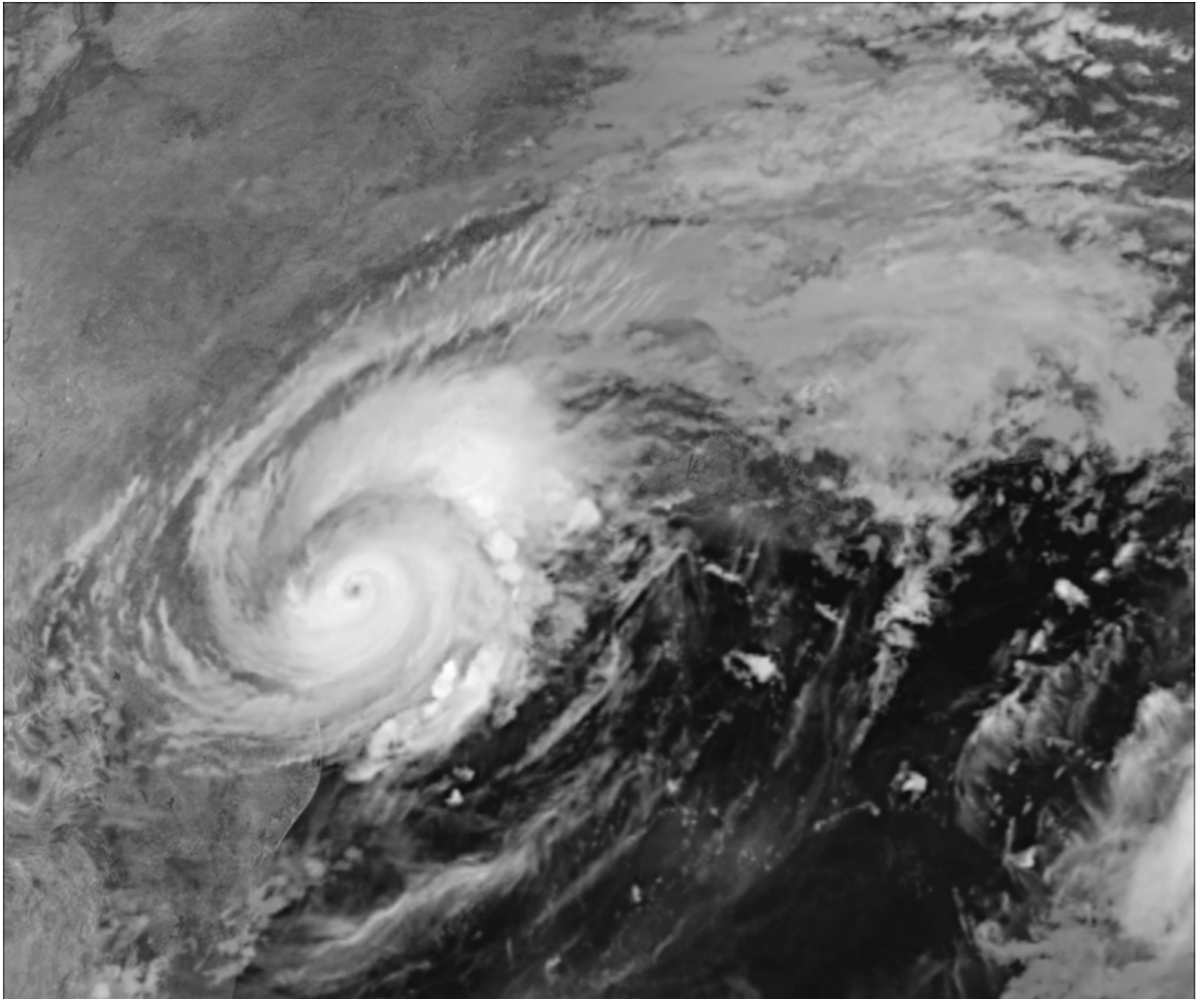
```
In [1]: using Images
        using Plots; pyplot()
        using LinearAlgebra
```

```
In [2]: img_path = "harvey-saturday-goes7am.jpg"
        img = load(img_path)
        size(img)
```

```
Out[2]: (1296, 1548)
```

```
In [3]: X = Gray.(img)
```

```
Out[3]:
```



```
In [4]: X = Float64.(X);
```

```
In [5]: function svd_compress(X, k)

    U,  $\sigma$ , V = svd(X, full=true); # slow to redo SVD, but not a big deal for homework
     $\tilde{X}$       = sum( $\sigma[i]$  * U[:, i] * V[:, i]' for i in 1:k)
    err      = norm(X -  $\tilde{X}$ )/norm(X)

    return  $\tilde{X}$ , err

end
```

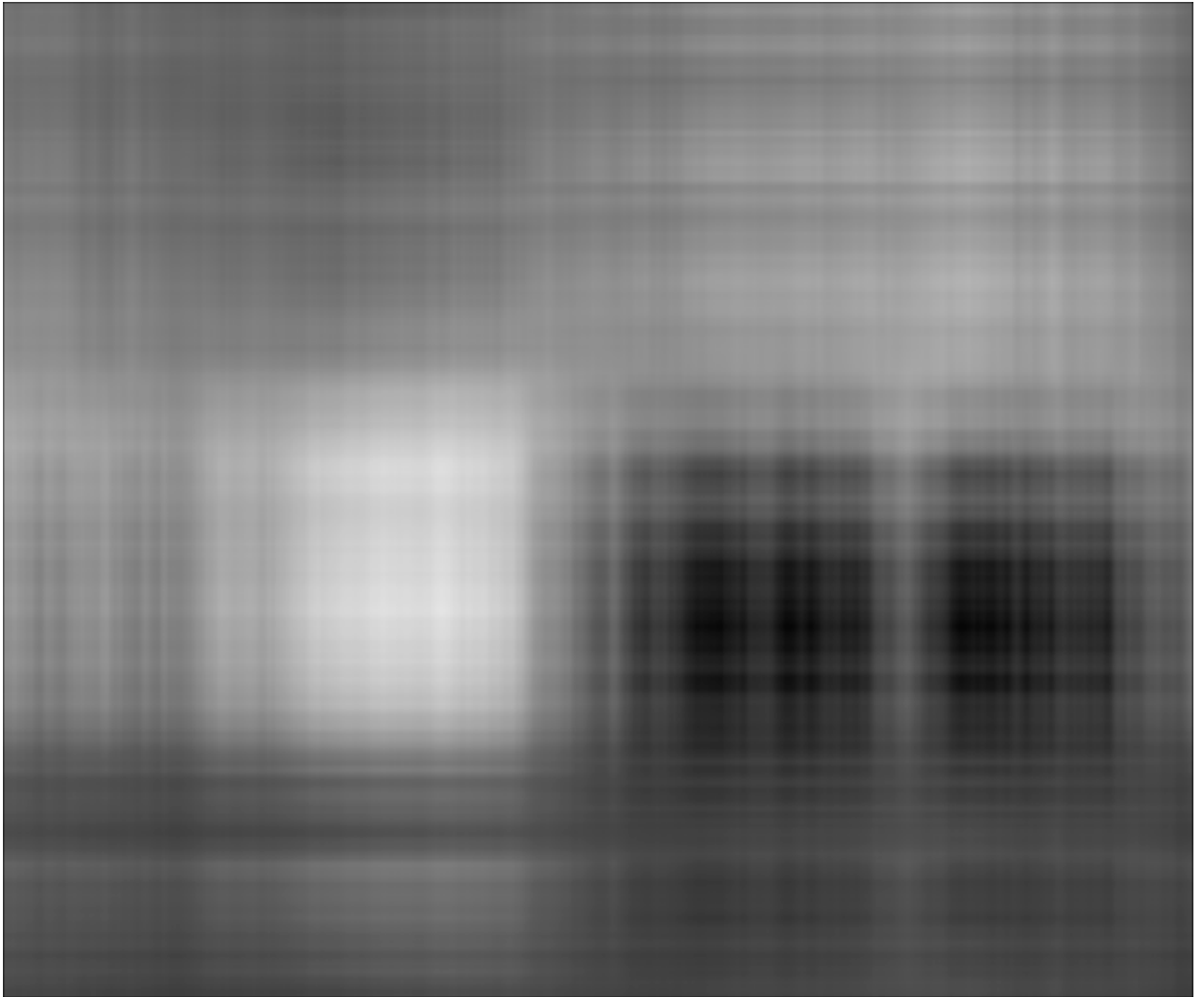
```
Out[5]: svd_compress (generic function with 1 method)
```

**Problem 1a** |  $k = 2$

```
In [6]:  $\tilde{X}$ , err = svd_compress(X, 2)
print(err)
#save("svd_k2.png", Gray.(\tilde{X}))
Gray.(\tilde{X})
```

0.2815103158754801

Out[6]:

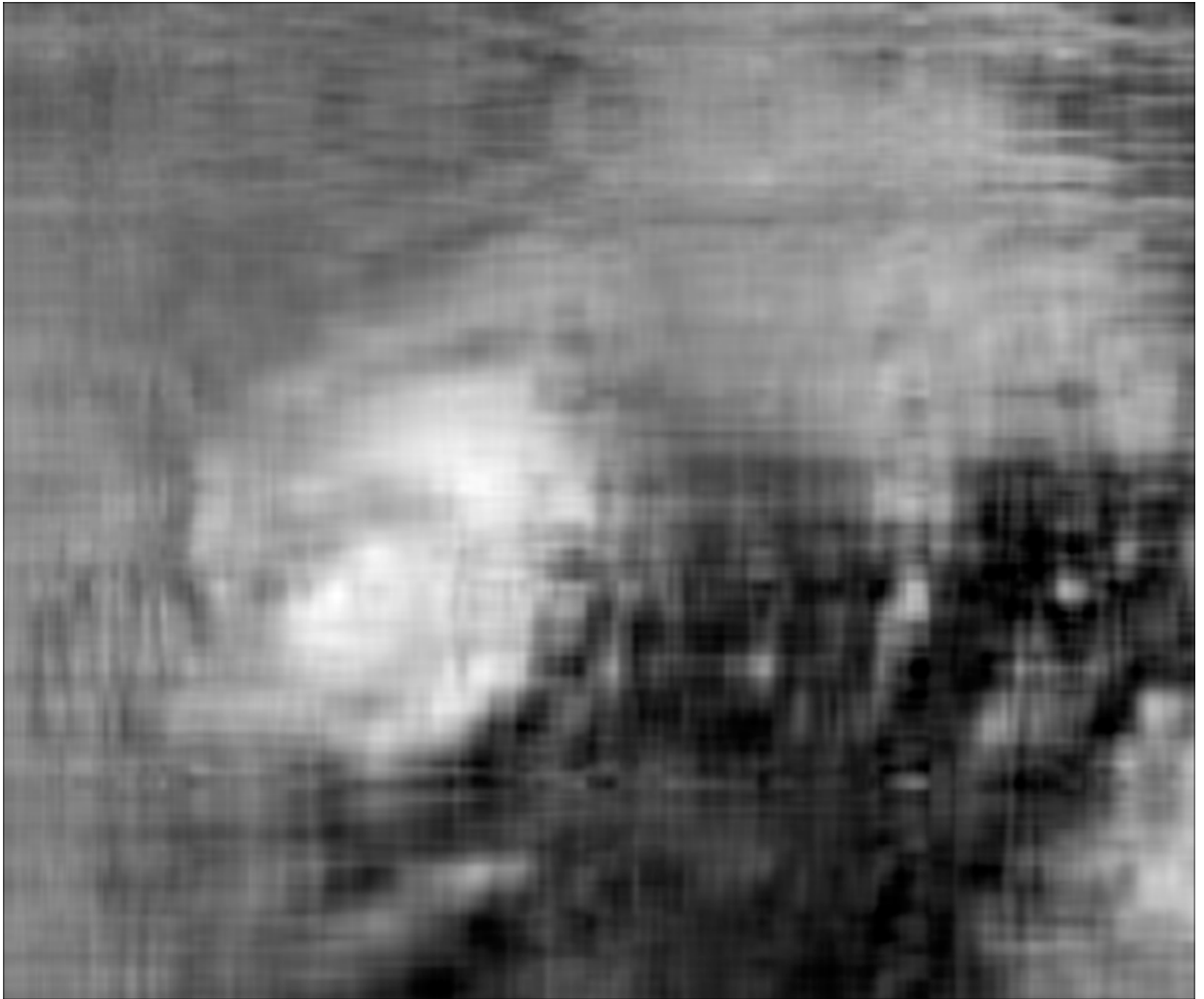


**Problem 1a** |  $k = 10$

```
In [7]:  $\tilde{X}$ , err = svd_compress(X, 10)
print(err)
#save("svd_k10.png", Gray.(\tilde{X}))
Gray.(\tilde{X})
```

0.15876586555633745

Out[7]:

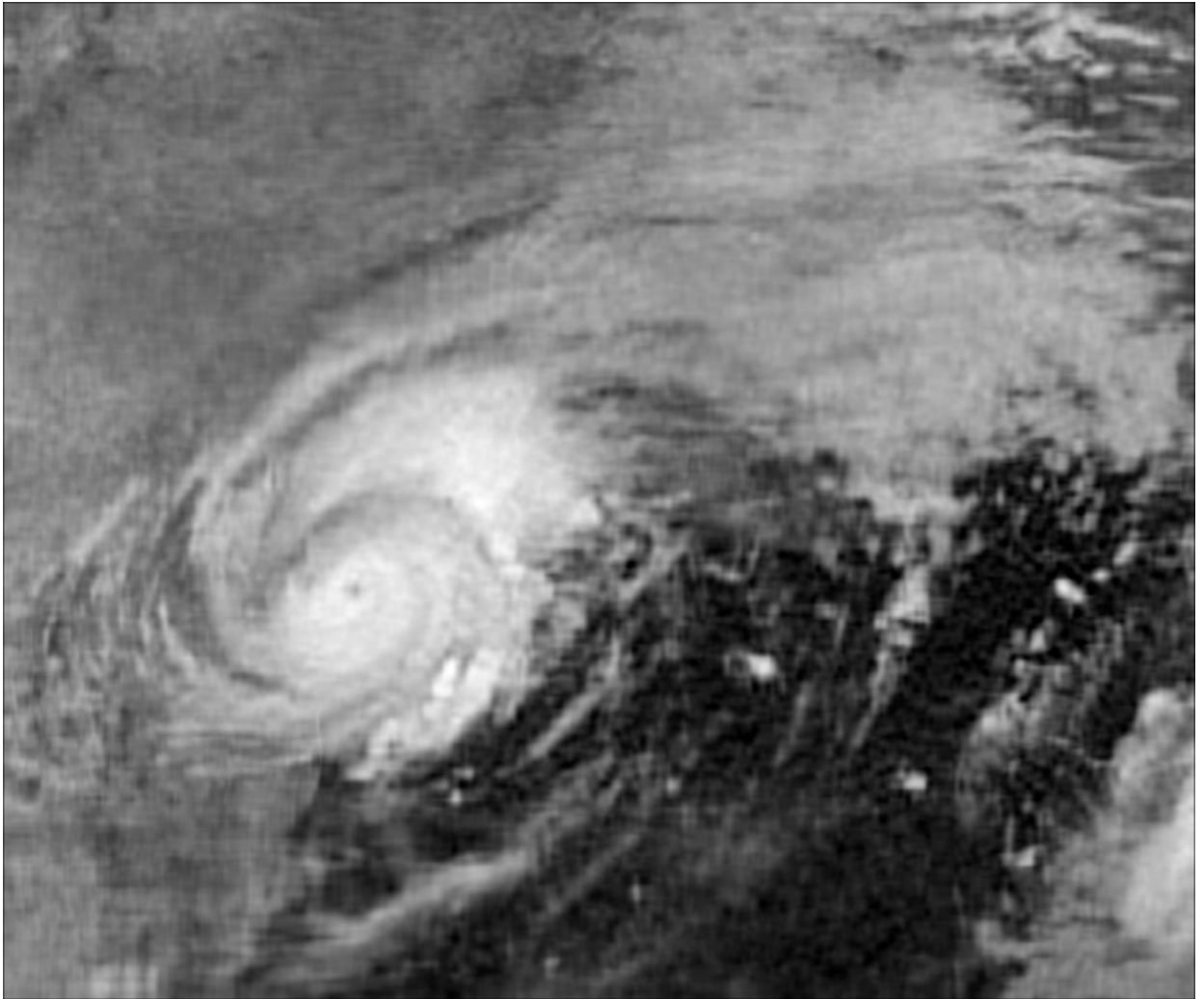


**Problem 1a** |  $k = 40$

```
In [8]:  $\tilde{X}$ , err = svd_compress(X, 40)
print(err)
#save("svd_k40.png", Gray.( $\tilde{X}$ ))
Gray.( $\tilde{X}$ )
```

0.08368458635729387

Out[8]:



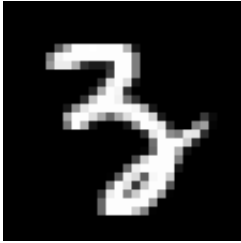
## Problem 4

```
In [9]: using MAT
using Statistics
```

```
In [10]: vars = matread("mnist_all.mat");
```

```
In [11]: test_3_9 = vars["test3"][9,:]  
img_as_mat(example) = permutedims(reshape(example, (28, 28)), [2, 1])  
Gray.(img_as_mat(test_3_9)/255)
```

Out[11]:



```
In [12]: function get_dataset(vars, classifier_digit)  
  
    X_train = zeros()  
    y_train = zeros()  
  
    X_test  = zeros()  
    y_test  = zeros()  
  
    for digit in 0:9  
  
        label = digit == classifier_digit ? 1 : 0  
  
        X_train = digit == 0 ? vars["train$(digit)"] : vcat(X_train, v  
ars["train$(digit)"])  
        y_train = digit == 0 ? label*ones(size(vars["train$(digit)"])[  
1], 1) : vcat(y_train, label*ones(size(vars["train$(digit)"])[1], 1))  
  
        X_test  = digit == 0 ? vars["test$(digit)"] : vcat(X_test, v  
ars["test$(digit)"])  
        y_test  = digit == 0 ? label*ones(size(vars["test$(digit)"])[1  
, 1) : vcat(y_test, label*ones(size(vars["test$(digit)"])[1], 1))  
  
    end  
  
    X_train = Float64.(X_train)  
    y_train = Float64.(y_train)  
  
    X_test  = Float64.(X_test)  
    y_test  = Float64.(y_test);  
  
    return X_train, y_train, X_test, y_test  
  
end
```

Out[12]: get\_dataset (generic function with 1 method)

```
In [13]: binarize(vec, threshold) = [v >= threshold ? 1 : 0 for v in vec]
zero_one_loss( $\hat{y}$ , y) = sum(  $\hat{y} \neq y$ )
accuracy(  $\hat{y}$ , y) = mean(  $\hat{y} == y$ )
squared_loss(  $\hat{y}$ , y) = mean(( $\hat{y} - y$ ).^2)
```

```
Out[13]: squared_loss (generic function with 1 method)
```

```

In [14]: digit = 2

X_train, y_train, X_test, y_test = get_dataset(vars, digit)

λ = 1E11

w = inv(X_train'*X_train + λ*I)*X_train'*y_train

ŷ_train = X_train * w
ŷ_test  = X_test  * w

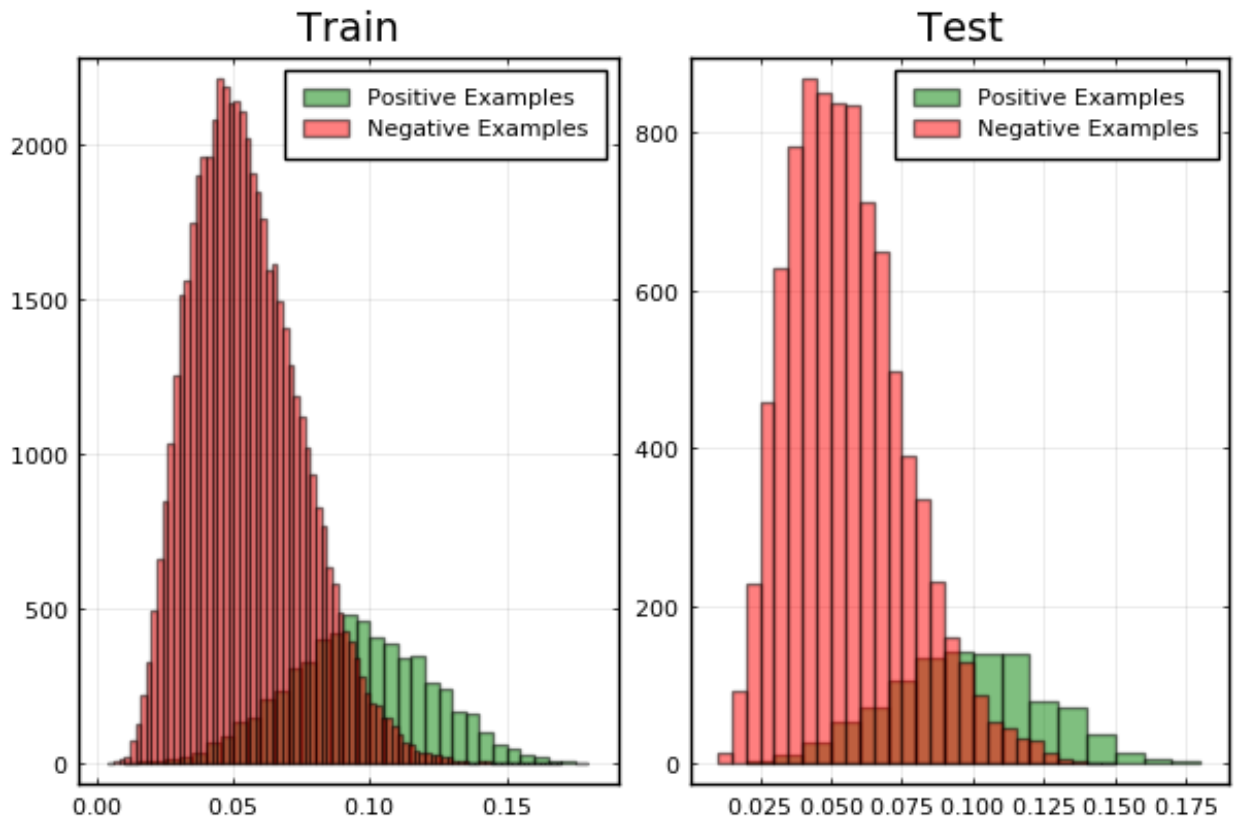
p1 = histogram( ŷ_train[y_train == 1], alpha=0.5, title="Train", label="Positive Examples", c=:green, box=:on)
      histogram!(ŷ_train[y_train == 0], alpha=0.5, label="Negative Examples", c=:red)

p2 = histogram( ŷ_test[ y_test == 1], alpha=0.5, title="Test", label="Positive Examples", c=:green, box=:on)
      histogram!(ŷ_test[ y_test == 0], alpha=0.5, label="Negative Examples", c=:red)

plot(p1, p2)

```

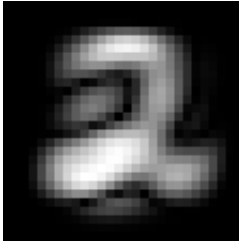
Out[14]:





```
In [15]: w = abs.(w)
w = (w .- minimum(w)) / (maximum(w) - minimum(w))
attention_map = Gray.(permutedims(reshape(w, (28, 28)), [2, 1]))
```

Out[15]:



```
In [16]: threshold = 0.1

ŷ_train_b = binarize(ŷ_train, threshold)
ŷ_test_b  = binarize(ŷ_test,  threshold)

@show train_zero_one_loss = zero_one_loss(ŷ_train_b, y_train)
@show train_squared_loss  = squared_loss( ŷ_train_b, y_train)
@show train_accuracy      = accuracy(      ŷ_train_b, y_train)

@show test_zero_one_loss  = zero_one_loss(ŷ_test_b,  y_test)
@show test_squared_loss   = squared_loss( ŷ_test_b,  y_test)
@show test_accuracy       = accuracy(      ŷ_test_b,  y_test);

train_zero_one_loss = zero_one_loss(ŷ_train_b, y_train) = 4637
train_squared_loss  = squared_loss(ŷ_train_b, y_train) = 0.07728333333333333
train_accuracy      = accuracy(ŷ_train_b, y_train) = 0.9227166666666666
test_zero_one_loss  = zero_one_loss(ŷ_test_b, y_test) = 819
test_squared_loss   = squared_loss(ŷ_test_b, y_test) = 0.0819
test_accuracy       = accuracy(ŷ_test_b, y_test) = 0.9181
```

## Problem 6

```
In [17]: using Distributions
```

```
In [18]: function rand_matrix_from_row_dist(dist, n, d)

    A = zeros(n, d)

    for i in 1:n
        A[i, :] = rand(dist)
    end

    return A

end
```

Out[18]: rand\_matrix\_from\_row\_dist (generic function with 1 method)

```
In [19]: n = 500
d = 50

μ = ones(d)
Σ = [2*0.5^abs(i-j) for i in 1:d, j in 1:d]

μt = zeros(d)

GA = MvNormal(μ, Σ)
T1 = MvTDist(1, μt, Σ)
T3 = MvTDist(3, μt, Σ)

A_GA = rand_matrix_from_row_dist(GA, n, d)
A_T3 = rand_matrix_from_row_dist(T3, n, d)
A_T1 = rand_matrix_from_row_dist(T1, n, d)

x_star = randn(d)

b_GA = A_GA * x_star + 0.01*rand(n)
b_T3 = A_T3 * x_star + 0.01*rand(n)
b_T1 = A_T1 * x_star + 0.01*rand(n);
```

## Problem 6(a)

```
In [20]: function uniform_distribution(A)

    n      = size(A)[1]
    dist = DiscreteNonParametric([i for i in 1:n], [1/n for i in 1:n])

end

function row_score_distribution(A)

    row_scores = [norm(row, 2)^2 for row in eachrow(A)]
    probs      = normalize(row_scores, 1)

    dist = DiscreteNonParametric([i for i in 1:size(A)[1]], probs)

end
```

Out[20]: row\_score\_distribution (generic function with 1 method)

```
In [21]: unif_dist_GA = uniform_distribution(A_GA)
unif_dist_T3 = uniform_distribution(A_T3)
unif_dist_T1 = uniform_distribution(A_T1)

rs_dist_GA   = row_score_distribution(A_GA)
rs_dist_T3   = row_score_distribution(A_T3)
rs_dist_T1   = row_score_distribution(A_T1);
```

```
In [22]: using ProgressMeter
```

```

In [23]: function random_sampling_approximate_least_squares(A, b, dist, r_max)

    d = size(A)[2]
    x = inv(A'*A)*A'*b
    errors = []

     $\tilde{A}$  = zeros(1, d)
     $\tilde{b}$  = zeros(1, 1)

    @showprogress for k in 1:r_max

        ik = rand(dist)

        if k == 1
             $\tilde{A}[k, :]$  = A[ik, :]
             $\tilde{b}[k]$  = b[ik]
        else
             $\tilde{A}$  = vcat( $\tilde{A}$ , reshape(A[ik, :], (1, d)))
             $\tilde{b}$  = vcat( $\tilde{b}$ , b[ik])
        end

        if k ≥ d
             $\tilde{x}$  = inv( $\tilde{A}'*\tilde{A}$ )* $\tilde{A}'*\tilde{b}$ 
            push!(errors, norm( $\tilde{x}$  .- x, 2) / norm(x, 2))
        end

    end

    return errors

end

```

```

Out[23]: random_sampling_approximate_least_squares (generic function with 1 method)

```



```
In [25]: r_max = 200

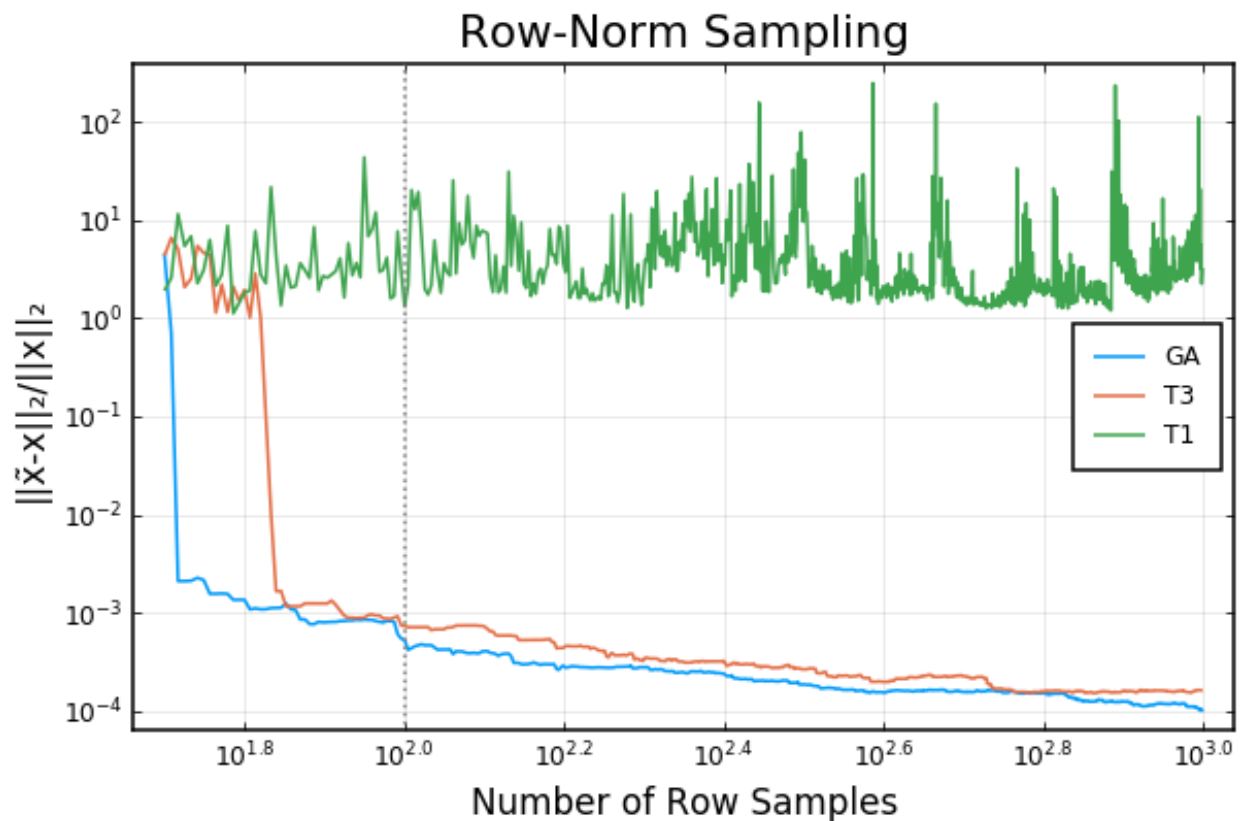
rs_errs_GA = random_sampling_approximate_least_squares(A_GA, b_GA, rs_
dist_GA, r_max)
rs_errs_T3 = random_sampling_approximate_least_squares(A_T3, b_T3, rs_
dist_T3, r_max)
rs_errs_T1 = random_sampling_approximate_least_squares(A_T1, b_T1, rs_
dist_T1, r_max)

plot( d:r_max, rs_errs_GA, label="GA", xscale=:log10, yscale=:log10, b
ox=:on, thickness_scaling=1.1,
      title="Row-Norm Sampling", xlabel="Number of Row Samples", yla
bel="|| $\tilde{x}-x$ ||2/|| $x$ ||2")
plot!(d:r_max, rs_errs_T3, label="T3")
plot!(d:r_max, rs_errs_T1, label="T1")
vline!([2d], c=:gray, ls=:dot, label=:none)

#png("als_error_rs_sampling.png")
```

[illegible]

Out[25]:



### Problem 6(b)

In [26]: **function** rademacher\_random\_projection\_matrix(m, d)

```
s = [rand() >= 0.5 ? 1 : -1 for _ in 1:m*d]
S = reshape(s, (m, d))

return S

end

function gaussian_random_projection_matrix(m, d)

s = [randn() for _ in 1:m*d]
S = 1/sqrt(m)*reshape(s, (m, d))

return S

end
```

Out[26]: gaussian\_random\_projection\_matrix (generic function with 1 method)

In [27]: **function** random\_projection\_approximate\_least\_squares(A, b, projection\_method, m\_max)

```
d = size(A)[2]
x = inv(A'*A)*A'*b
errors = []

@showprogress for m in d:m_max

    S = projection_method(m, n)

    SA = S*A
    Sb = S*b

     $\bar{x}$  = inv(SA'*SA)*SA'*Sb
    push!(errors, norm( $\bar{x}$  .- x, 2) / norm(x, 2))

end

return errors

end
```

Out[27]: random\_projection\_approximate\_least\_squares (generic function with 1 method)

```
In [28]: m_max = 10d

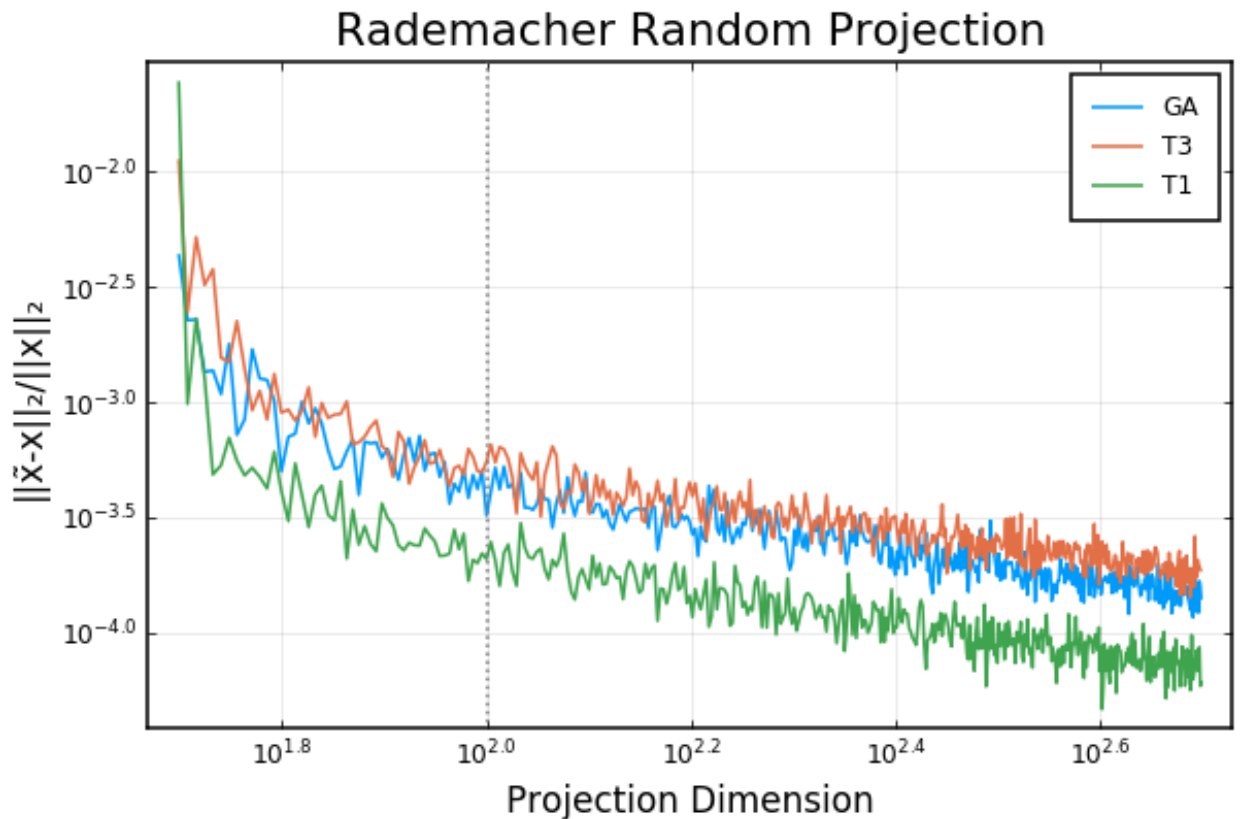
rade_errs_GA = random_projection_approximate_least_squares(A_GA, b_GA,
rademacher_random_projection_matrix, m_max)
rade_errs_T3 = random_projection_approximate_least_squares(A_T3, b_T3,
rademacher_random_projection_matrix, m_max)
rade_errs_T1 = random_projection_approximate_least_squares(A_T1, b_T1,
rademacher_random_projection_matrix, m_max)

plot( d:m_max, rade_errs_GA, label="GA", xscale=:log10, yscale=:log10,
box=:on, thickness_scaling=1.1,
      title="Rademacher Random Projection", xlabel="Projection Dimen
sion", ylabel="|| $\tilde{x}-x$ ||2/|| $x$ ||2")
plot!(d:m_max, rade_errs_T3, label="T3")
plot!(d:m_max, rade_errs_T1, label="T1")
vline!([2d], c=:gray, ls=:dot, label=:none)

#png("als_error_rade_projection.png")
```

```
Progress: 100%|████████████████████████████████████████████████████████████████████████████████| Time: 0:00:07
Progress: 100%|████████████████████████████████████████████████████████████████████████████████| Time: 0:00:04
Progress: 100%|████████████████████████████████████████████████████████████████████████████████| Time: 0:00:04
```

Out[28]:





```
In [29]: m_max = 10d

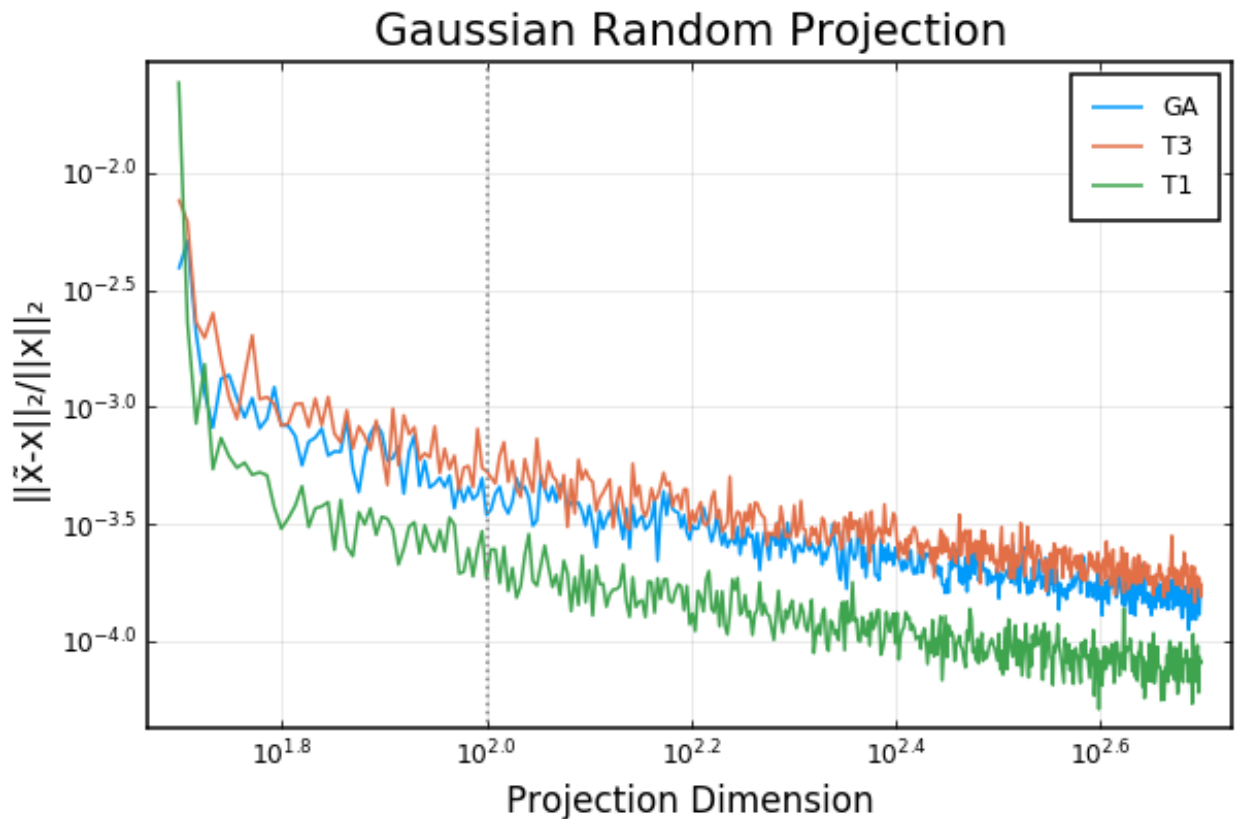
gaussian_errs_GA = random_projection_approximate_least_squares(A_GA, b
_GA, gaussian_random_projection_matrix, m_max)
gaussian_errs_T3 = random_projection_approximate_least_squares(A_T3, b
_T3, gaussian_random_projection_matrix, m_max)
gaussian_errs_T1 = random_projection_approximate_least_squares(A_T1, b
_T1, gaussian_random_projection_matrix, m_max)

plot( d:m_max, gaussian_errs_GA, label="GA", xscale=:log10, yscale=:lo
g10, box=:on, thickness_scaling=1.1,
      title="Gaussian Random Projection", xlabel="Projection Dimensi
on", ylabel="|| $\bar{x}-x$ ||2/|| $x$ ||2")
plot!(d:m_max, gaussian_errs_T3, label="T3")
plot!(d:m_max, gaussian_errs_T1, label="T1")
vline!([2d], c=:gray, ls=:dot, label=:none)

#png("als_error_gaussian_projection.png")
```

```
Progress: 100%|██████████████████████████████████████████| Time: 0:00:02
Progress: 100%|██████████████████████████████████████████| Time: 0:00:02
Progress: 100%|██████████████████████████████████████████| Time: 0:00:02
```

Out[29]: Gaussian Random Projection



### Problem 6(c)

```
In [30]: function sparse_gaussian_random_projection_matrix(m, d, q=0.5)

    s = [rand() <= q ? 0 : (1-q)*randn() for _ in 1:m*d]
    S = 1/sqrt(m)*reshape(s, (m, d))

    return S

end
```

Out[30]: sparse\_gaussian\_random\_projection\_matrix (generic function with 2 methods)

```
In [31]: m_max = 10d

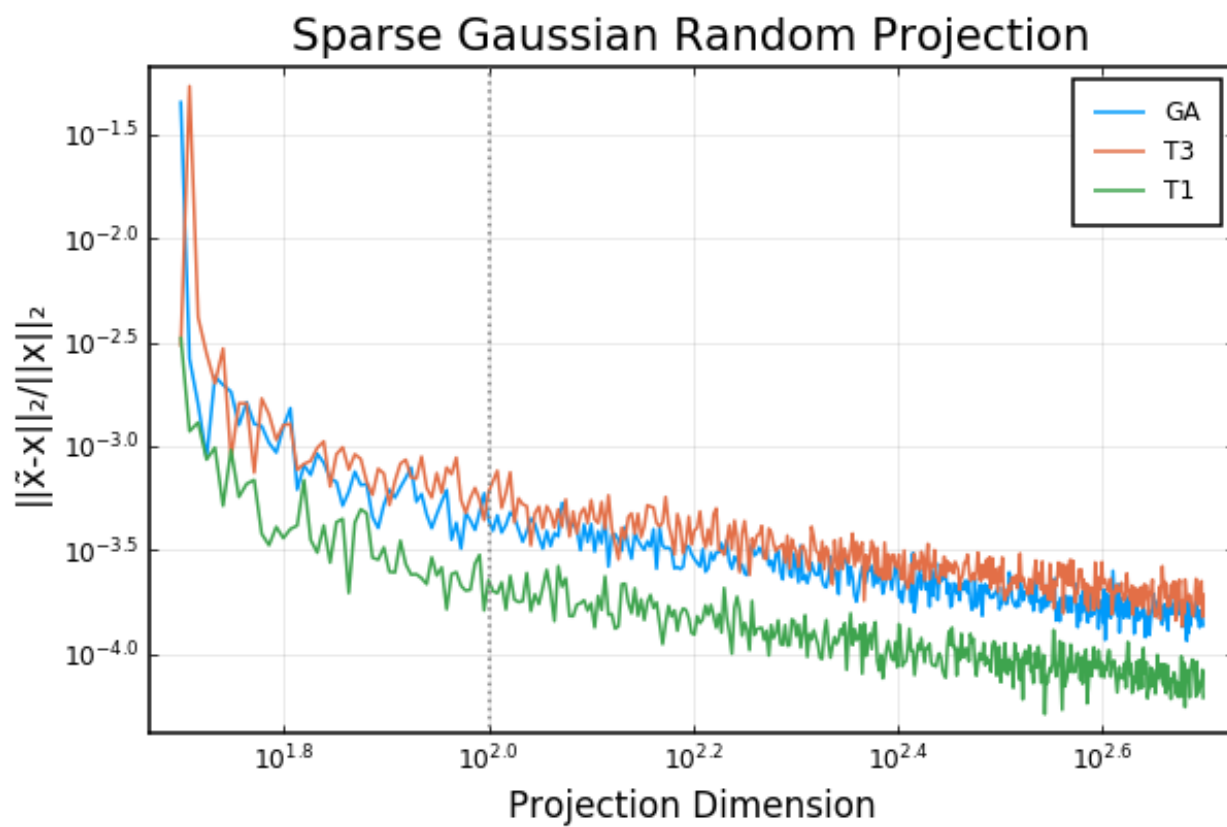
sparse_gaussian_errs_GA = random_projection_approximate_least_squares(
A_GA, b_GA, sparse_gaussian_random_projection_matrix, m_max)
sparse_gaussian_errs_T3 = random_projection_approximate_least_squares(
A_T3, b_T3, sparse_gaussian_random_projection_matrix, m_max)
sparse_gaussian_errs_T1 = random_projection_approximate_least_squares(
A_T1, b_T1, sparse_gaussian_random_projection_matrix, m_max)

plot( d:m_max, sparse_gaussian_errs_GA, label="GA", xscale=:log10, yscale=:log10, box=:on, thickness_scaling=1.1,
      title="Sparse Gaussian Random Projection", xlabel="Projection Dimension", ylabel="|| $\tilde{x}-x$ ||2/||x||2")
plot!(d:m_max, sparse_gaussian_errs_T3, label="T3")
plot!(d:m_max, sparse_gaussian_errs_T1, label="T1")
vline!([2d], c=:gray, ls=:dot, label=:none)

#png("als_error_sparse_gaussian_projection.png")
```

```
Progress: 100% | Time: 0:00
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Progress: 100% | Time: 0:00
:14
Progress: 100% | Time: 0:00
:17
```

Out[31]:



```

In [32]: function random_projection_approximate_least_squares_vary_q(A, b, projection_method, m)

    d = size(A)[2]
    x = inv(A'*A)*A'*b
    qvals = []
    errors = []

    @showprogress for q in 0.01:0.01:0.99

        S = projection_method(m, n, q)

        SA = S*A
        Sb = S*b

        try
             $\tilde{x}$  = inv(SA'*SA)*SA'*Sb
            push!(qvals, q)
            push!(errors, norm( $\tilde{x}$  .- x, 2) / norm(x, 2))
        catch
        end

    end

    return qvals, errors

end

```

Out[32]: random\_projection\_approximate\_least\_squares\_vary\_q (generic function with 1 method)

```

In [33]: m = d

q_GA, sparse_gaussian_errs_q_GA = random_projection_approximate_least_squares_vary_q(A_GA, b_GA, sparse_gaussian_random_projection_matrix, m)
q_T3, sparse_gaussian_errs_q_T3 = random_projection_approximate_least_squares_vary_q(A_T3, b_T3, sparse_gaussian_random_projection_matrix, m)
q_T1, sparse_gaussian_errs_q_T1 = random_projection_approximate_least_squares_vary_q(A_T1, b_T1, sparse_gaussian_random_projection_matrix, m)

p1 = plot(q_GA, sparse_gaussian_errs_q_GA, label="GA", ylabel=:log10, box=:on,
           title="Sparse Gaussian Proj. (m = $m)", xlabel="Sparsity Parameter, q", ylabel="|| $\tilde{x}$ -x||2/||x||2")
plot!(q_T3, sparse_gaussian_errs_q_T3, label="T3")
plot!(q_T1, sparse_gaussian_errs_q_T1, label="T1")

m = Int(round(1.5d))

```

```

q_GA, sparse_gaussian_errs_q_GA = random_projection_approximate_least_
squares_vary_q(A_GA, b_GA, sparse_gaussian_random_projection_matrix, m
)
q_T3, sparse_gaussian_errs_q_T3 = random_projection_approximate_least_
squares_vary_q(A_T3, b_T3, sparse_gaussian_random_projection_matrix, m
)
q_T1, sparse_gaussian_errs_q_T1 = random_projection_approximate_least_
squares_vary_q(A_T1, b_T1, sparse_gaussian_random_projection_matrix, m
)

p2 = plot(q_GA, sparse_gaussian_errs_q_GA, label="GA",yscale=:log10,
box=:on,
         title="Sparse Gaussian Proj. (m = $m)", xlabel="Sparsity Param
eter, q", ylabel="|| $\tilde{x}-x$ ||2/||x||2")
plot!(q_T3, sparse_gaussian_errs_q_T3, label="T3")
plot!(q_T1, sparse_gaussian_errs_q_T1, label="T1")

m = 2d

q_GA, sparse_gaussian_errs_q_GA = random_projection_approximate_least_
squares_vary_q(A_GA, b_GA, sparse_gaussian_random_projection_matrix, m
)
q_T3, sparse_gaussian_errs_q_T3 = random_projection_approximate_least_
squares_vary_q(A_T3, b_T3, sparse_gaussian_random_projection_matrix, m
)
q_T1, sparse_gaussian_errs_q_T1 = random_projection_approximate_least_
squares_vary_q(A_T1, b_T1, sparse_gaussian_random_projection_matrix, m
)

p3 = plot(q_GA, sparse_gaussian_errs_q_GA, label="GA",yscale=:log10,
box=:on,
         title="Sparse Gaussian Proj. (m = $m)", xlabel="Sparsity Param
eter, q", ylabel="|| $\tilde{x}-x$ ||2/||x||2")
plot!(q_T3, sparse_gaussian_errs_q_T3, label="T3")
plot!(q_T1, sparse_gaussian_errs_q_T1, label="T1")

plot(p1, p2, p3, layout=(1,3), size=(1000, 400))
#png("als_error_sparse_gaussian_projection.png")

```

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:01		

Out[33]:

