ECE 133A HW 6

Lawrence Liu

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Exercise A10.1

(a)

Let $u = [u(0), u(1), \cdots, u(N-1)]^T$, then we have that $x = u$ we want to minimize the energy or

$$||u||^2$$

Furthermore we can express

$$s_1(N) = 0.1u(N-2)+(0.95+1)\cdot 0.1u(N-3)+(0.95^2+0.95+1)\cdot 0.1u(N-4)+\cdots+(\sum_{i=0}^{N-2} 0.95^i)\cdot 0.1u(0)$$

$$s_2(N) = 0.1u(N-1)+(0.95)\cdot 0.1u(N-2)+(0.95^2)\cdot 0.1u(N-3)+\cdots+(0.95^{N-1})\cdot 0.1u(0)$$

So therefore we have that

$$C = \begin{bmatrix} 0.1\sum_{i=0}^{N-2} 0.95^i & \cdots & 0.1(0.95+1) & 0.1 & 0 \\ 0.1(0.95^{N-1}) & \cdots & 0.1(0.95^2) & 0.1(0.95) & 0.1 \end{bmatrix}$$

And

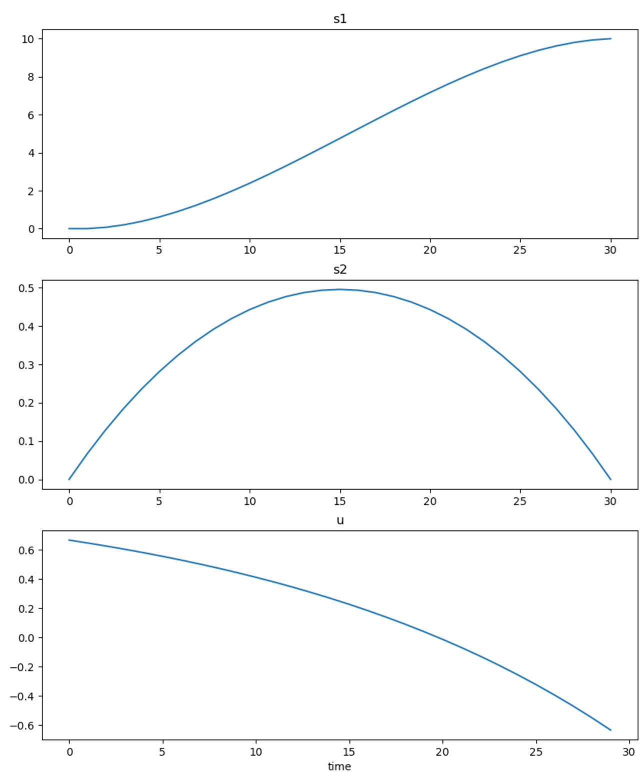
$$d = \begin{bmatrix} 10 \\ 0 \end{bmatrix}$$

(b)

```
using PyPlot

function create_C(N)
    C = zeros(2,N)

    #make the first row of C
    for i=1:N-1
        for j=1:N-i
            C[1,i]+=0.1*(0.95^(j-1))
        end
    end
    # C[1,N-1]=0.1

    #make the second row of C
    for i=1:N
        C[2,i] = 0.1*(0.95^(N-i))
    end
    return C
end

# calculate u
N=30
C = create_C(N)
d=[10,0]
u = C\d
s1=zeros(N+1)
s2=zeros(N+1)
for i=1:N
    s1[i+1]=s1[i]+s2[i]
    s2[i+1]=0.95*s2[i]+0.1*u[i]
end

fig,axs=subplots(3,1,figsize=(10,12))
axs[1].plot(s1)
axs[1][:set_title]("s1")
```

```
axs[2][:plot](s2)
axs[2][:set_title]("s2")
axs[3][:plot](u)
axs[3][:set_title]("u")
axs[3][:set_xlabel]("time")

savefig("problem2a.png")
close()
```

We get the following plot



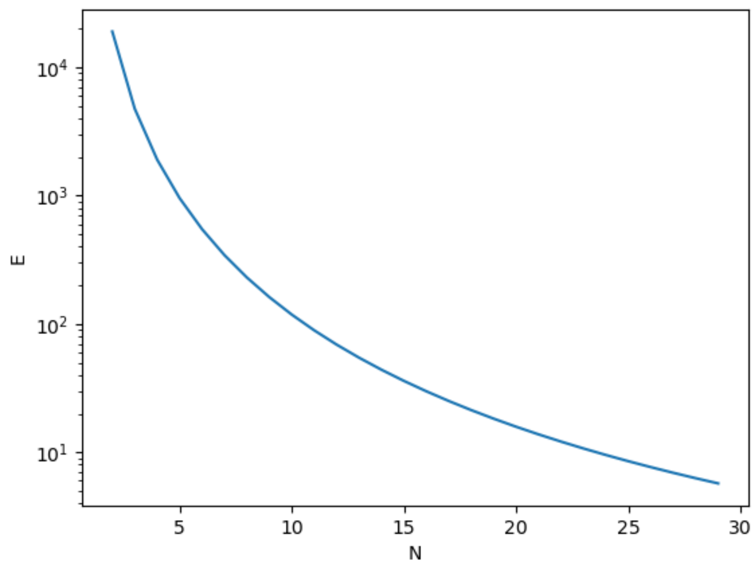
(c)

With the following code:

```
using PyPlot

function create_C(N)
    C = zeros(2,N)

    #make the first row of C
    for i=1:N-1
        for j=1:N-i
            C[1,i]+=0.1*(0.95^(j-1))
        end
    end
    # C[1,N-1]=0.1

    #make the second row of C
    for i=1:N
        C[2,i] = 0.1*(0.95^(N-i))
    end
    return C
end

N=2:29
println(N)
E=zeros(length(N))
for i=1:length(N)
    C = create_C(N[i])
    d=[10,0]
    u = C\d
    E[i]=sum(u.^2)
end
plot(N,E)
xlabel("N")
ylabel("E")
yscale("log")
```

```
savefig("problem2b.png")
close()
```

We get the following plot



Exercise A10.9

(a)

We use Langrage multipliers to solve this problem.

$$L(x) = ||Ax - b||^2 + \lambda e_i^T x$$

$$
\begin{aligned}
\nabla L(x) &= 0 \\
2A^T(Ax - b) + \lambda e_i &= 0 \\
2A^TAx &= 2A^Tb - \lambda e_i \\
A^TAx &= A^Tb - \frac{\lambda}{2}e_i \\
x &= (A^TA)^{-1}A^Tb - \frac{\lambda}{2}(A^TA)^{-1}e_i \\
x &= \hat{x} - \frac{\lambda}{2}(A^TA)^{-1}e_i
\end{aligned}
$$

subsituting this back into the constraint we get

$$
\begin{aligned}
e_i^T x &= 0 \\
e_i^T\left(\hat{x} - \frac{\lambda}{2}(A^TA)^{-1}e_i\right) &= 0 \\
\hat{x}_i - \frac{\lambda}{2}(A^TA)^{-1}_{ii} &= 0 \\
\hat{x}_i &= \frac{\lambda}{2}(A^TA)^{-1}_{ii}\lambda = \frac{2\hat{x}_i}{(A^TA)^{-1}_{ii}}
\end{aligned}
$$

And thus we get that

$$
x = \hat{x} - \frac{\hat{x}_i}{(A^TA)^{-1}_{ii}}(A^TA)^{-1}e_i
$$

(b)

Calculating the QR factorization of A costs $2mn^2$ flops. And solving $\hat{x}$ costs an additional $2mn + n^2$ flops. Then we can solve for $(A^TA)^{-1}e_i$ using backwards and forwards substitution which costs $2n^2$ flops, and then finding $(A^TA)^{-1}_{ii}$ is just finding the value for the ith index in the vector $(A^TA)^{-1}e_i$ which costs 0 flops. This is the same for $\hat{x}_i$. Then calculating $\frac{\hat{x}_i}{(A^TA)^{-1}_{ii}}$ costs 1 flops, and then multiplying that to every value of $(A^TA)^{-1}e_i$ costs n flops. And then subtracting the resulting vector from $\hat{x}$ will cost n flops. So the total cost is $\boxed{2mn^2 + 2mn + 3n^2 + 2n + 1}$ flops.