## EE263 Homework 8

Fall 2023
7.1080. Hovercraft with limited range. We have a hovercraft moving in the plane with two thrusters, each pointing through the center of mass, exerting forces in the $\mathbf{x}$ and $\mathbf{y}$ directions with $100 \%$ efficiency. The hovercraft has mass 1 . The discretized equations of motion for the hovercraft are

$$
x(t+1)=\left[\begin{array}{llll}
1 & 1 & 0 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 1 & 1 \\
0 & 0 & 0 & 1
\end{array}\right] x(t)+\left[\begin{array}{cc}
\frac{1}{2} & 0 \\
1 & 0 \\
0 & \frac{1}{2} \\
0 & 1
\end{array}\right]\left[\begin{array}{l}
u_{1}(t) \\
u_{2}(t)
\end{array}\right]
$$

where $x_{1}$ and $x_{2}$ are the position and velocity in the $\mathbf{x}$-direction, and $x_{3}, x_{4}$ are the position and velocity in the $\mathbf{y}$-direction. Here

$$
u(t)=\left[\begin{array}{l}
u_{1}(t) \\
u_{2}(t)
\end{array}\right]
$$

is the force acting on the hovercraft for time in the interval $[t, t+1)$. Let the position of the vehicle at time $t$ be $q(t) \in \mathbb{R}^{2}$.
a) The hovercraft starts at the origin. We'd like to apply thrust to make it move through points $p_{1}, p_{2}, p_{3}$ at times $t_{1}, t_{2}, t_{3}$, where

$$
\begin{array}{lll}
p_{1}=\left[\begin{array}{c}
1 \\
-\frac{1}{2}
\end{array}\right] & p_{2}=\left[\begin{array}{l}
0 \\
1
\end{array}\right] & p_{3}=\left[\begin{array}{c}
-\frac{3}{2} \\
0
\end{array}\right] \\
t_{1}=6 & t_{2}=40 & t_{3}=50
\end{array}
$$

We will run the hovercraft on the time interval $[0,70]$. We'd like to apply a sequence of inputs $u(0), u(1), \ldots, u(70)$ to make the hovercraft position pass through the above sequence of points at the specified times.
We would like to find the sequence of inputs that drives the hovercraft through the desired points which has the minimum cost, given by the sum of the squares of the forces:

$$
\sum_{t=0}^{70}\|u(t)\|^{2}
$$

To do this, pick $A_{\text {hov }}$ and $y_{\text {des }}$ to set this problem up as an equivalent minimum-norm problem, where we would like to find the minimum-norm $u_{\text {seq }}$ which satisfies

$$
A_{\mathrm{hov}} u_{\mathrm{seq}}=y_{\mathrm{des}}
$$

where $u_{\text {seq }}$ is the sequence of force inputs

$$
u_{\mathrm{seq}}=\left[\begin{array}{c}
u(0) \\
u(1) \\
\vdots \\
u(70)
\end{array}\right]
$$

Plot the trajectory of the hovercraft using this input, and the way-points $p_{1}, \ldots, p_{3}$. Also plot the optimal $u$ against time.
b) Now we would like to compute the trade-off curve between the accuracy with which the mass passes through the waypoints and the norm of the force used. Let our two objective functions be

$$
J_{1}=\sum_{i=1}^{3}\left\|q\left(t_{i}\right)-p_{i}\right\|^{2}=\left\|A_{\mathrm{hov}} u_{\mathrm{seq}}-y_{\mathrm{des}}\right\|^{2}
$$

and

$$
J_{2}=\sum_{t=0}^{70}\|u(t)\|^{2}
$$

By minimizing the weighted sum

$$
J_{1}+\mu J_{2}
$$

for a range of values of $\mu$, plot the trade-off curve of $J_{1}$ against $J_{2}$ showing the achievable performance. This above trade-off curve shows how we can trade-off between how accurately the hovercraft passes through the waypoints and how much input energy is used.
c) For each of the following values of $\mu$

$$
\left\{\left.10^{\frac{p}{2}} \right\rvert\, p=-2,0,2, \ldots, 10\right\}
$$

plot the trajectories all on the same plot, together with the waypoints.
d) Now suppose we are controlling the hovercraft by radio control, and the maximum range possible between the transmitter and receiver is 2 (in whatever units we are using for distance.) Notice that, if we use the minimum-norm input then the hovercraft passes out of range, both when making its first turn and on the final stretch (between times 50 and 70).
We'd like to do something about this, but trading off the input norm as above doesn't do the right thing; if $\mu$ is large then the hovercraft stays within range, but misses the waypoints entirely; if $\mu$ is small then it comes close to the waypoints, but goes out of range. Notice that this is particularly a problem on the final stretch between times 50 and 70 ; explain why this is.
e) One remedy for this problem is to solve a constrained multiobjective least-squares problem. We would like to impose the constraint that

$$
A_{\mathrm{hov}} u_{\mathrm{seq}}=y_{\mathrm{des}}
$$

that is, achieve zero waypoint error $J_{1}=0$. We can attempt to keep the hovercraft in range by trading off the sum of the squares of the position

$$
J_{3}=\sum_{t=0}^{70}\|q(t)\|^{2}
$$

against input cost $J_{2}$ subject to this constraint. To do this, we'll solve

$$
\begin{aligned}
\operatorname{minimize} & J_{3}+\gamma J_{2} \\
\text { subject to } & A_{\mathrm{hov}} u_{\mathrm{seq}}=y_{\mathrm{des}}
\end{aligned}
$$

First, find the matrix $W$ so that the cost function is given by

$$
J_{3}+\gamma J_{2}=\left\|W u_{\text {seq }}\right\|^{2}
$$

f) Now we have a problem of the form

$$
\begin{array}{cl}
\text { minimize } & \|W u\|^{2} \\
\text { subject to } & A u=y_{\mathrm{des}}
\end{array}
$$

This is called a weighted minimum-norm solution; the only difference from the usual minimum-norm solution to $A u=y_{\text {des }}$ is the presence of the matrix $W$, and when $W=I$ the optimal $u$ is just given by $u_{\mathrm{opt}}=A^{\dagger} y_{\text {des }}$. Show that the solution for general $W$ is

$$
u_{\mathrm{opt}}=\Sigma^{-1} A^{T}\left(A \Sigma^{-1} A^{T}\right)^{-1} y_{\mathrm{des}}
$$

where $\Sigma=W^{T} W$. (One way to do this is using Lagrange multipliers.) Use this to solve the remaining parts of this problem.
g) For each of the following values of $\gamma$

$$
\left\{\left.10^{\frac{p}{2}} \right\rvert\, p=0,2,4, \ldots, 20\right\}
$$

Plot the trajectories all on the same plot, together with the waypoints. Explain what you see.
h) By trying different values of $\gamma$, you should be able to find a trajectory which just keeps the hovercraft within range. Plot the trajectory of the hovercraft; what is the corresponding value of $\gamma$ ? Is this the smallest-norm input $u$ that just keeps the hovercraft within range, and drives the hovercraft through the waypoints? Explain why, or why not.
i) For a range of values of $\gamma$, plot the trade-off curve of $J_{3}$ against $J_{2}$ showing the achievable performance.
13.2030. A method for rapidly driving the state to zero. We consider the discrete-time linear dynamical system

$$
x(t+1)=A x(t)+B u(t),
$$

where $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times k}, k<n$, is full rank. The goal is to choose an input $u$ that causes $x(t)$ to converge to zero as $t \rightarrow \infty$. An engineer proposes the following simple method: at time $t$, choose $u(t)$ that minimizes $\|x(t+1)\|$. The engineer argues that this scheme will work well, since the norm of the state is made as small as possible at every step. In this problem you will analyze this scheme.
a) Find an explicit expression for the proposed input $u(t)$ in terms of $x(t), A$, and $B$.
b) Now consider the linear dynamical system $x(t+1)=A x(t)+B u(t)$ with $u(t)$ given by the proposed scheme (i.e., as found in (a)). Show that $x$ satisfies an autonomous linear dynamical system equation $x(t+1)=F x(t)$. Express the matrix $F$ explicitly in terms of $A$ and $B$.
c) Now consider a specific case:

$$
A=\left[\begin{array}{ll}
0 & 3 \\
0 & 0
\end{array}\right], \quad B=\left[\begin{array}{l}
1 \\
1
\end{array}\right] .
$$

Compare the behavior of $x(t+1)=A x(t)$ (i.e., the orginal system with $u(t)=0$ ) and $x(t+1)=F x(t)$ (i.e., the original system with $u(t)$ chosen by the scheme described above) for a few initial conditions. Determine whether each of these systems is stable.
18.1250. Chasing a sea monster. A sea monster is loose in the Pacific Ocean! Your monster-chasing colleague has been measuring the sea monster's movements and has predicted it will surface at $m$ positions $p_{i} \in \mathbb{R}^{2}$ at times $s_{i}$. Here $p_{i}$ is the $i$ th column of the matrix $P$ given by

$$
P=\left[\begin{array}{cccccc}
1 & 1.75 & 2.4 & 2 & 0.5 & 0 \\
0.75 & 0.6 & 1.2 & 2.3 & 0.75 & 0
\end{array}\right]
$$

and the times $s=(2,5,8,11,17,20)$. You plan to observe the monster with a drone. Unfortunately the sea monster ate the last two drones you sent and you are almost out of research funding so your drone's sensors are not very good, and the drone must be exactly in the right position to observe the monster.
a) The dynamics of the drone are

$$
\ddot{q}=u
$$

where $q \in \mathbb{R}^{2}$ is the position of the drone, and $u \in \mathbb{R}^{2}$ is an input force. Write this as a linear dynamical system of the form

$$
\begin{aligned}
\dot{x} & =A x+B u \\
y & =C x
\end{aligned}
$$

where $y \in \mathbb{R}^{2}$ is the position of the drone.
b) We will use sample period $h$. Assume that the force input is piecewise constant on sample intervals, and construct the exact discretization

$$
\begin{aligned}
x_{d}(k+1) & =A_{d} x_{d}(k)+B_{d} u_{d}(k) \\
y_{d}(k) & =C_{d} x_{d}(k)
\end{aligned}
$$

where $x_{d}(k)=x(k h)$, and similarly for $y_{d}$ and $u_{d}$.
c) The drone starts at the origin with zero velocity, and we would like to move the drone so that $y\left(s_{i}\right)=p_{i}$ for $i=1, \ldots, m$. We will operate the drone on the time interval $[0, T]$
where $T=s_{m}$. For convenience, let $N=T / h$. Since drone batteries are limited, we would like to minimize

$$
J=\sum_{k=0}^{N-1}\left\|u_{d}(k)\right\|^{2}
$$

Explain in detail how you would solve this problem.
d) Use your method to compute the optimal input $u$, and plot $u$ versus time. Use $h=0.1$.
e) Report the optimal value of $J$ that you obtained.
f) Plot the trajectory of the drone. Use axes $q_{1}$ and $q_{2}$, so that the plot shows the path followed by the drone. Mark on your plot the points $p_{i}$ where the monster surfaces.
g) Draw a sea monster for 1 point of extra credit.
18.2890. Linear dynamical systems for portfolio management. We consider a portfolio of $n$ financial assets (like stocks) and cash, which we manage over $T$ time steps of unit length (e.g. one month). We call $x_{t} \in \mathbb{R}^{n+1}$ for $t=1, \ldots, T$ our state vector. The first $n$ elements are our positions in each of the assets, in dollars, and the last element is the dollar amount of cash we hold. Every element of $x$ can be either positive (for long positions) and negative (for short or borrowing). For $t=1, \ldots, T-1$, the transition from $x_{t}$ to $x_{t+1}$ is composed of two steps.

- First, the portfolio positions change value because of market returns. Let $\mu \in \mathbb{R}_{++}^{n}$ be the vector of returns, where $\mathbb{R}_{++}^{n}$ is the set of all vectors of length $n$ with strictly positive entries. We define the post-return portfolio $\tilde{x}_{t}$ to be

$$
\left(\tilde{x}_{t}\right)_{i}= \begin{cases}\mu_{i}\left(x_{t}\right)_{i} & i=1, \ldots, n \\ \left(x_{t}\right)_{i} & i=n+1 .\end{cases}
$$

(Intuitively, cash is unchanged, and the asset positions are multiplied by the corresponding element of the vector of returns.) For simplicity we assume that the vector of returns does not change in time.

- Then we trade. We can exchange any amount of cash for the corresponding amount of any of the assets. Note that the only valid trades are cash for asset. If you wish to trade some amount of an asset with the same amount of another asset, you have to perform two trades: trade the first asset with cash, and then trade cash with the second asset. (Think carefully about this definition of trade when you formulate the transaction costs.) For example, if we buy $c>0$ dollars of the first asset and sell $d>0$ dollars of the second asset the state evolves as

$$
x_{t+1}=\tilde{x}_{t}+\left[\begin{array}{c}
c \\
-d \\
0 \\
\vdots \\
0 \\
-(c-d)
\end{array}\right] .
$$

Finally, we define the portfolio value $v_{t} \in \mathbb{R}$ for $t=1, \ldots, T$ to be

$$
v_{t}=\mathbf{1}^{T} x_{t}
$$

a) Formulate the problem as a linear dynamical system of the form

$$
x_{t+1}=A x_{t}+B u_{t}, \quad t=1, \ldots, T-1 .
$$

The control vector $u_{t}$ should have dimension $n$.
b) Assume that our trades incur quadratic transaction costs with parameter $\rho>0$. For example, if at time $t$ we buy $c>0$ dollars of the first asset, and we sell $d>0$ dollars of the second asset (the example above), then the transaction costs for the transition $x_{t}$ to $x_{t+1}$ are

$$
\rho\left(c^{2}+d^{2}\right) .
$$

(Be careful, they are not $\rho(c+d)^{2}$.) Explain how to solve the problem of maximizing the final value of the portfolio $v_{T}$ minus the total transaction costs. (The sequence of controls $u_{1}, \ldots, u_{T-1}$ that achieves the maximum should be a function of $A, B$, and $\rho$ ). Use methods from EE263.
c) Apply your method to the following data.

```
T = 12;
x_1 = [1000, 1000, 0, 1000, 0, 0];
mu = [1.001, 1.003, 1.004, 1.006, 1.007];
rho = 0.0001;
```

What is the final value $v_{T}$ ? What are the total transaction costs? Plot the trajectories of the portfolio positions $x_{t}$. (On the same plot you should draw $n+1$ lines, one for each of the assets and cash, with time on the $x$-axis.)
d) Now assume that we aim to liquidate an initial portfolio, which means that at time $T$ we want to have zero positions in any of the $n$ assets and only hold cash. We thus impose the constraint $\left(x_{T}\right)_{i}=0$, for $i=1, \ldots, n$. Explain how to solve the problem of maximizing the final portfolio value (in this case, all cash) minus the transaction costs with this additional constraint. Use methods from EE263.
e) Apply your method to the data given above. What is the final value $v_{T}$ ? What are the total transaction costs? Plot the trajectories of the portfolio positions $x_{t}$. (On the same plot you should draw $n+1$ lines, one for each of the assets and cash, with time on the $x$-axis.)

