## HW \#5

## Introduction

The project consists of a set of exercises and design problems covering most of the material presented in class. The use of the tools presented in class is strongly recommended, especially in the design problems.

## Reach sets

1. Consider the system Linear model with $A=[1,0 ; 0,1]$ and $B=[1,0]$. Build an approximation to the reach set at time $t=10$ when the system departs from the origin at time $t=0$.
2. Consider the following optimal control problem

$$
\begin{aligned}
& \text { Min c. } x(1) \text { (c is a vector in the unit ball) } \\
& d x / d t=A x+B u ; x(0)=x_{0} ; u \in \Omega
\end{aligned}
$$

Consider that you are given $R\left(x_{0} ; 0,1\right)$, the reach set of the system at time $t=1$ when departing from $x_{0}$ at time $t=0$.
a) Use $R\left(x_{0}, 0,1\right)$ to find $x^{*}(1)$, the final state of the solution to the optimal control problem Min c.z over all $z \in R\left(x_{0} ; 0,1\right)$
b) Use formula $x(1)=e^{A} x_{0}+0 \int_{0}^{1} e^{A(1-t)} B u(t) d t$ in order to show that the optimal control $u^{*}(t)$ maximizes the map $v \rightarrow c^{\top} e^{A(1-t)} B$ vover $\Omega$.
c) Check that $x^{*}(t)$ minimizes $p(t)$.z over all $z \in R\left(x_{0} ; 0, t\right)$, onde $p(t)^{\top}=c^{\top} e^{A(1-t)}$

Hint: Use the following facts:
i) The minimization in a) is also satisfied when the set $R\left(x_{0} ; 0,1\right)$ is replaced by $e^{A(1-t)}\left[R\left(x_{0} ; 0, t\right)-x^{*}(t)\right]+x^{*}(1)$.
Observe that $e^{A(1-t)}\left[R\left(x_{0} ; 0, t\right)-x^{*}(t)\right]+x^{*}(1) \subseteq R\left(x_{0} ; 0,1\right)$.
ii) $x^{*}(1)=e^{A(1-t)} x^{*}(t)+\int_{t}^{1} e^{A(1-s)} B u^{*}(s) d s$.
iii) $z \in A+v$ implies $z=z_{1}+v$ for some $z_{1} \in A$.
3. Consider the linear system: $d x / d t=A x+B u ; x(0)=x_{0}, u \in \Omega, t \in[0,1]$, where $x \in \mathbf{R}^{3}$ and $u \in \mathbf{R}^{1}$. Assume that the system is controllable.
a) Let $\Omega=[\alpha, \beta]$, with $\alpha<\beta$. Show that $R\left(x_{0} ; 0,1\right)$ is convex.
b) Show that if you replace $\Omega=[\alpha, \beta]$ by $\Omega=\{\alpha, \beta\}$, the reach set $R\left(x_{0} ; 0,1\right)$ does not change.

Observation:
(1) The points on the boundary of $R\left(x_{0} ; 0,1\right)$ can only be reached by values of the control on the boundary of $\Omega$.
(2) Points in the interior of $R\left(x_{0} ; 0,1\right)$ can be reached by values of the control on the boundary of $\Omega$ (convexifying effect of the integral).
Hint: Use the following facts (you may/should verify them):
i) $C$ is a convex set if $\mathrm{ac}_{1}+(1-a) \mathrm{c}_{2} \in \mathrm{C}$, for all $\mathrm{a} \in[0,1], \mathrm{c}_{1} \in \mathrm{C}$ and $\mathrm{c}_{2} \in \mathrm{C}$.
ii) Check the convexification role of the integral.
iii) Start with one dimensional state and, then, extend to dimension $2,3, \ldots n$.
4. Inner polyhedral approximation to the reachable set of linear systems.

Use the exercise in 9 to devise an algorithm to compute an inner polyhedral approximation of the reachable $R\left(x_{0} ; 0,1\right)$.for the linear system $d x / d t=A x+B u ; x(0)=x_{0}, u \in \Omega$.

Justify the following observations:
i) The solution to the minimization in 8.a) is always on the boundary of the reachable set.
ii) Given a set of $N \geq n+1$ points $Z=\left\{z_{1}, z_{2}, z_{3}, \ldots, z_{N}\right\}, z_{i} \in R^{n}, \operatorname{co}[Z] \subset R\left(x_{0}, 0,1\right)$.

Here co[Z] denotes the convex hull of the set $Z$. Consider the system in the plane, i.e., $n=2$, for which $A=[0,1 ; 2,1], B=[0,1]^{\mathrm{T}}$, and $\mathrm{U}=[-1,1]$ in order to devise a clever way of generating interesting parameters $c_{k}, k=1 \ldots P,\left|c_{k}\right|=1$ that, by replacing $c$ in the optimization problem in 8.a) will yield convenient vertices $x^{*}{ }_{k}(1)$ of the approximating polyhedral.

Hints:
a) The solutions to each one of the optimal control problems can easily be computed by hand as you can see from the optimal control notes.
b) Consider the following scheme:
a. Initial face: choose an arbitrary vector $c_{1},\left|c_{1}\right|=1$, and compute $x^{*}{ }_{1}(1)$. Then, let $\mathrm{C}_{2}=-\mathrm{c}_{1}$, and compute $\mathrm{x}^{*}{ }_{2}(1)$.
b. Let $c_{3},\left|c_{3}\right|=1$ be perpendicular to the vector $x^{*}{ }_{2}(1)-x^{*}{ }_{1}(1)$ and compute $x^{*}{ }_{3}(1)$. Then, let $\mathrm{C}_{4}=-\mathrm{C}_{3}$, and compute $\mathrm{x}^{\star}{ }_{4}(1)$.
c. Proceed as previously but now minding that the interior of the approximating polyhedron is nonempty and that the recursively generated vectors $c$ have to point outwards.
c) Construct an approximation with 8 points and discuss the "error" of the approximation, i.e., the "distance" (how should it be defined? Is Hausdorff metric ok?) between the sets

$$
\operatorname{co}\left\{x^{\star}{ }_{1}(1), x^{\star}{ }_{2}(1), \ldots, x^{*}(1)\right\} \text { and } R\left(x_{0} ; 0,1\right) .
$$

Observation: The Hausdorff distance between sets $A$ and $B$ is denoted by $d_{H}(A, B)$ and is defined by

$$
d_{H}(A, B)=: \max \left\{\sup \left\{d \_A(b): b \in B\right\}, \sup \left\{d \_B(a): a \in A\right\}\right\}
$$

where $d \_C(d)$ is the usual concept of distance between the point $d$ and the set $C$, that is

$$
\mathrm{d} \_C(\mathrm{~d})=: \inf \{\|\mathrm{c}-\mathrm{d}\|: \mathrm{c} \in \mathrm{C}\} .
$$

## Annex - Vehicle models

We consider several models of increasing complexity for the motions of our vehicles in $R^{2}$.
The state is $\mathrm{x}=\left(\mathrm{x}_{1}, \mathrm{x}_{2}\right) \in \mathrm{R}^{2}$
The general model is
$\mathrm{dx} / \mathrm{dt}=\mathrm{f}(\mathrm{x}, \mathrm{u})$

## Simple vehicle model

$\mathrm{dx} / \mathrm{dt} \in \mathrm{B}$ (unit closed ball in $\mathrm{R}^{2}$ )

## Simple vehicle model under adversarial behaviour

$d x / d t=u+v$ where $u \in B$ (unit closed ball in $R^{2}$ ) and $v \in 0.5 B$ (closed ball with radius 0.5 in $R^{2}$ )

## "Arrow model" <br> $\mathrm{dx}_{1} / \mathrm{dt} \in\{-1,0,1\}$ <br> $\mathrm{dx}_{2} / \mathrm{dt} \in\{-1,0,1\}$

## Unicycle model

This model has a third state $\Theta$ to model the orientation of the vehicle. The inputs are the speed $u_{s} \in$ $[0,1]$ and the angular velocity $r \in[-.1,1]$.
$\mathrm{dx}_{1} / \mathrm{dt}=\mathrm{u}_{\mathrm{s}} \cos \theta$
$\mathrm{dx}_{2} / \mathrm{dt}=\mathrm{u}_{\mathrm{s}} \sin \Theta$
$\mathrm{d} \Theta / \mathrm{dt}=\mathrm{r}$
Planar motions of autonomous vehicles are often described by this simplified model.

## Mode controlled unicycle model

The potential behavior of a vehicle modelled as a unicycle is the set of all smooth curves. This is too unstructured. We restrict its behavior to 4 discrete modes of motion:

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stop (u=0, Ө =0),
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straight ( $u=5, \Theta=0$ ),
left ( $u=0, \Theta=1$ ),
right $(u=0, \Theta=-1)$.
The vehicle is controlled by selecting a mode of motion.

## Linear model

$\mathrm{dx} / \mathrm{dt}=\mathrm{Ax}+\mathrm{Bu}, \mathrm{u} \in \mathrm{U}=[-1,1]$

