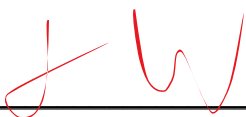
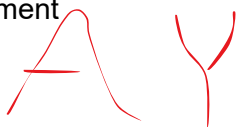


Homework Assignment 4 – 600.455/655 Fall 2025

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Signature (required) I/We have followed the rules in completing this assignment 	Signature (required) I/We have followed the rules in completing this assignment 

1. Remember that this is a graded homework assignment. It is the functional equivalent of a take-home exam.

2. You are to work **alone** or in **teams of two** and are not to discuss the problems with anyone other than the TAs or the instructor.

3. **IMPORTANT NOTE:** If you work in teams of two, you are **not** to split up the questions and each answer a subset individually. You are to work **together**. ~~I encourage~~ teaming on these problems because I believe that it encourages learning, not a way to reduce the required work for students taking the course. By signing this sheet you are asserting that each of you has contributed equally to each answer and can individually explain the answer as well as if you had answered the question alone. I view this as a question of trust and ethics.

4. It is otherwise open book, notes, and web. But you should cite any references you consult.

5. Please refer to the course organizational notes for a fuller listing of all the rules. I am not reciting them all here, but they are still in effect.

6. Unless I say otherwise in class, it is due before the start of class on the due date posted on the web.

7. Submit the assignment on GradeScope as a neat and legible PDF file. We will not insist on typesetting your answers, but we must be able to read them. We will not go to extraordinary lengths to decipher what you write. If the graders cannot make out an answer, the score will be 0.

8. Sign and hand in this page as the first sheet of your assignment. If you work with a partner, then you both should sign the sheet, but you should only submit one PDF file for both of you, using the GradeScope teaming feature. Indicate clearly who it is from.

This assignment has more than 100 points, but the most that will be applied to your grade is 100

HW 4

Wrist

\vec{a} : all joints $F_{RW} = [R_{RW}(\vec{a}), \vec{P}_{RW}(\vec{a})]$ $F_{RW}(a+\Delta a) = F_{RW}(\vec{a}) \Delta F_{RW}(\vec{a}, \Delta \vec{a})$

$\Delta F_{RW}(\vec{a}, \Delta \vec{a}) \approx \Delta F_{RW}(\vec{\eta}_{RW}) = [I + S_K(\vec{d}_{RW}), \vec{E}_{RW}]$ ^{twist}

$\vec{\eta}_{RW} = [\vec{d}_{RW}, \vec{E}_{RW}]^T = J_{RW}(\vec{a}) \Delta \vec{a}$

F_{WT} : wrist \rightarrow tool A_{tool} : tool shaft length

forces forces
wrench $\xi_w = [\phi_w^T, \tau_w^T]^T$ in F_w

Robot state: $[a, \dot{a}, \xi_w]$

$s(t) = [\overset{\text{joint}}{\vec{a}}, \overset{\text{joint}}{\dot{\vec{a}}}, \overset{\text{wrist}}{\xi_w}, \overset{\text{wrist}}{\tau_w}, \overset{\text{wrist}}{\phi_w}]$

1. Each loop of this algorithm is executed every Δt . However, the exact timing between each step is not constant due to computing times. This means that to maintain deterministic timing to avoid gaps or spikes in commands, which could be incompatible with low-level controllers, we must output the command immediately in step 1, not later on where timing becomes non-deterministic.

2. Mechanical constraints: $a_{min} \leq a \leq a_{max}, |\dot{a}| \leq \dot{a}_{max}, |\ddot{a}| \leq \ddot{a}_{max}$

To ensure these will not be violated by the ^{next} command output Δa_{cmd}

Check:

$a_{min} \leq a + \Delta a_{cmd} \leq a_{max}$
 $|\frac{1}{\Delta t} \Delta a_{cmd}| \leq \dot{a}_{max}$
 $|\frac{1}{\Delta t^2} \Delta a_{cmd} - \frac{1}{\Delta t} \dot{a}| \leq \ddot{a}_{max}$ ^{given robot state at time t and \dot{a} is known Δt}

3. With $\vec{\eta}_{RT}(\vec{a}, \Delta a) = [\vec{d}_{RT}, \vec{E}_{RT}]$ and $J_{RT}(a, \Delta a)$

$F_{RT}(a) = F_{RW}(a) F_{WT}$ $F_{RW}(a+\Delta a) = F_{RW}(a) \Delta F_{RW}(\vec{\eta}_{RW})$

First find J_{WT} st. $\vec{\eta}_{RT} = J_{WT} \vec{\eta}_{RW}$

$F_{RT}(a+\Delta a) = F_{RW}(a+\Delta a) F_{WT}$
 $= F_{RW}(a) \Delta F_{RW}(\vec{\eta}_{RW}) F_{WT}$

$F_{RT}(a+\Delta a) = F_{RT}(a) \Delta F_{RT}(\vec{\eta}_{RT})$
 $= F_{RW}(a) F_{WT} \Delta F_{RT}(\vec{\eta}_{RT})$

The same notions, must be related by some adjoint matrix

3. now, we have

$$F_{RW}(a) \Delta F_{RW}(\eta_{RW}) F_{WT} = F_{RW}(a) F_{WT} \Delta F_{RT}(\eta_{RT})$$

$$\Delta F_{RW}(\eta_{RW}) F_{WT} = F_{WT} \Delta F_{RT}(\eta_{RT})$$

$$\Delta F_{RW}(\eta_{RW}) = F_{WT} \Delta F_{RT}(\eta_{RT}) F_{WT}^{-1}$$

Rewrite

$$\eta = \begin{bmatrix} d \\ \varepsilon \end{bmatrix} \rightarrow \hat{\eta} = \begin{bmatrix} \text{sk}(d) & \varepsilon \\ 0 & 0 \end{bmatrix} \text{ so now, } \Delta F = I + \hat{\eta}$$

$$I + \hat{\eta}_{RW} = F_{WT} (I + \hat{\eta}_{RT}) F_{WT}^{-1}$$

$$I + \hat{\eta}_{RW} = I + F_{WT} \hat{\eta}_{RT} F_{WT}^{-1}$$

$$\hat{\eta}_{RW} = F_{WT} \hat{\eta}_{RT} F_{WT}^{-1} \Rightarrow \eta_{RW} = \text{Adj}_{F_{WT}} \eta_{RT}$$

definition of adjoint

$$\begin{bmatrix} R_{WT} & P_{WT} \\ 0 & I \end{bmatrix} \begin{bmatrix} \text{sk}(d_{RT}) & \varepsilon_{RT} \\ 0 & 0 \end{bmatrix} \begin{bmatrix} R_{WT}^{-1} & -R_{WT}^{-1}P \\ 0 & I \end{bmatrix}$$

$$= \begin{bmatrix} R_{WT} \text{sk}(d_{RT}) - R_{WT} \varepsilon_{RT} & R_{WT} \varepsilon_{RT} \\ 0 & 0 \end{bmatrix}$$

$$\text{Adj}_{F_{WT}} = \begin{bmatrix} R_{WT} & R_{WT} \text{sk}(P_{WT}) \\ 0 & R_{WT} \end{bmatrix}$$

$$\eta_{RT} = \text{Adj}_{F_{WT}}^{-1} \eta_{RW}$$

so $J_{WT}(a) = \text{Adj}_{F_{WT}}^{-1}$
we know $J_{RT} = J_{WT} J_{RW}(a)$

$$J_{RT} = \text{Adj}_{F_{WT}}^{-1} J_{RW}(a)$$

where $\text{Adj}_{F_{WT}} = \begin{bmatrix} R_{WT} & R_{WT} \text{sk}(P_{WT}) \\ 0 & R_{WT} \end{bmatrix}$

4.

we want $|\eta_{RT}| \leq \eta_{RT}^{\max}$

Since $\eta_{RT} = \text{Adj}_{F_{WT}}^{-1} \eta_{RW}$, $|\text{Adj}_{F_{WT}}^{-1} \eta_{RW}| \leq \eta_{RT}^{\max}$

where $\text{Adj}_{F_{WT}}$ is defined above, and $F_{WT} = F_{RW}^{-1} F_{RT}$

6.

If the robot is deficient, we should use a least-squares objective function and terminate when residuals stop decreasing.

So

instead of $\eta_{RT}^{cmd} = J_{RT}(a) \Delta a_{cmd}$,

$$\Delta a_{cmd} = \underset{\Delta a_{cmd}}{\operatorname{argmin}} \left| \eta_{RT}^{desired} - J_{RT}(a) \Delta a_{cmd} \right|^2$$

and terminate when $\eta_{RT}^{desired} - J_{RT}(a) \Delta a_{cmd}$ stops decreasing

7.

If the robot has redundant joints, we can add a term to the objective function to encourage joints to stay near their midpoint.

Let

$$a_{mid} = \frac{1}{2}(a_{min} + a_{max})$$

$w_{desired}, w_{mid}$ = weights, $w_{desired} \gg w_{mid}$

Then we can do the same as before but objective function as:

$$\Delta a_{cmd} = \underset{\Delta a_{cmd}}{\operatorname{argmin}} \left(w_{desired} \left| \eta_{RT}^{desired} - J_{RT}(a) \Delta a_{cmd} \right|^2 + w_{mid} \left| (a + \Delta a_{cmd}) - a_{mid} \right|^2 \right)$$

Subject to same constraints as questions 2 and 4

8.

$$h \dot{\eta}_{RW} = K \tilde{\eta}_{RW} / \Delta t = \begin{bmatrix} k_{2x3}^{rot} & 0 \\ 0 & k_{2x3}^{vel} \end{bmatrix} \begin{bmatrix} \tilde{F}_w \\ \tilde{p}_w \end{bmatrix} / \Delta t$$

$$\Delta F_{RW}^{cmd} = [I + S k(d_{RW}^{cmd}), \epsilon_{RW}^{cmd}]$$

we want a Δa_{cmd} s.t.

$$J_{RW}(a) \Delta a_{cmd} = \eta_{RW}^{cmd}$$

So we want to optimize:

$$\Delta a_{cmd} = \underset{\Delta a_{cmd}}{\operatorname{argmin}} \left| \eta_{RW}^{cmd} - J_{RW}(a) \Delta a_{cmd} \right|^2$$

Subject to:

$$\left[\begin{array}{l} a_{min} \leq a + \Delta a \leq a_{max} \quad \left| \frac{1}{\Delta t} \Delta a \right| \leq \dot{a}_{max} \quad \left| \frac{1}{\Delta t^2} \Delta a - \frac{1}{\Delta t} \dot{a} \right| \leq \ddot{a}_{max} \end{array} \right] \star$$

Pseudocode:

$$\text{handOverHand}(a, a\text{-dot}, J_{RW}, \zeta_{RW}, K_{RW}, \Delta t) : \left[\begin{array}{l} |\eta_{RW}| \leq \eta_{RW}^{max} \end{array} \right]$$

$$\eta_{RW\text{-dot}} \leftarrow K_{RW} * \zeta_{RW} \quad // 6 \times 1$$

$$\eta_{RW} \leftarrow \eta_{RW\text{-dot}} * \Delta t \quad // 6 \times 1$$

$$\Delta a_{cmd} \leftarrow \underset{\Delta a_{cmd}}{\operatorname{argmin}} \left(\operatorname{abs}(\eta_{RW} - J_{RW}(a) * \Delta a_{cmd}) \right)$$

Subject to constraints in \star

Using a quadratic programming solver

$$\Delta a_{cmd\text{-dot}} \leftarrow \Delta a_{cmd} / \Delta t$$

return $\Delta a_{cmd\text{-dot}}$

CIS HW4.9-13

Justin Wang, Alan You

November 2025

1 Question 9

To achieve constrained admittance, we need to modify step 3 of the algorithm for some cases. Observe that for admittance, we only care about the translation of the tip not the rotation, so we only care about the velocity admittance weights.

For each control cycle we compute the tool tip position in CT frame:

$$p_{CT}(t)$$

For each structure $X \in \{A, B, C\}$ we get its SDF and gradient:

$$d_X(t) = \text{GetMap}(p_{CT}(t), \text{"SDF_X"}), \quad g_X(t) = \nabla d_X(p_{CT}(t)) = \text{GetGrad}(p_{CT}(t), \text{SDF_X})$$

Using Fig. 3, for each X we compute the distance-dependent virtual fixture admittance gains $\kappa_X^{(1)}(d_X)$ and $\kappa_X^{(2)}(d_X)$, leaving us with three possibilities we care about:

1. If $d_X(t) > d_X^{\text{thresh}}$, we use the unconstrained admittance from Question 8 (Step 3 is unmodified)
2. If $0 \leq d_X(t) \leq d_X^{\text{thresh}}$, we interpolate $\kappa_X^{(1)}(d_X)$ and $\kappa_X^{(2)}(d_X)$ according to Fig. 3.
3. If $d_X(t) < 0$, we additionally require that the commanded translational motion points outward.

If we hit possibility number 2, we have to calculate a new admittance weight matrix. We only care about constraining motion normal to the bodies, not tangentially.

The outward SDF normal is:

$$\mathbf{n}_X(t) = \mathbf{g}_X(t) / \|\mathbf{g}_X(t)\|$$

We define projection operators onto the normal direction and the tangent plane:

$$P_{\parallel}^{(X)}(t) = \mathbf{n}_X(t) \mathbf{n}_X^T(t), \quad P_{\perp}^{(X)}(t) = I - P_{\parallel}^{(X)}(t).$$

Now that we have the tangential and normal directions, we can construct an effective admittance weight matrix for each of the bodies of interest ($X \in A, B, C$):

$$K_{\text{vel}}^{(X)}(t) = \kappa_X^{(1)}(d_X(t)) P_{\parallel}^{(X)}(t) + \kappa_X^{(2)}(d_X(t)) P_{\perp}^{(X)}(t).$$

Once we have $K_{\text{vel}}^{(A)}(t), K_{\text{vel}}^{(B)}(t), K_{\text{vel}}^{(C)}(t)$, we need to construct an effective admittance weight matrix because we need to consider the distance of the tip relative to each body concurrently. This can be achieved by constructing an admittance weight matrix such that the translational motion is affected by the strongest "push-back" from each body. Observe that smaller weight value means stronger perceived "push-back". Also observe that we are only ever modifying the velocity weights never the rotation weights. We can construct:

$$K_{\text{vel}}^{\text{eff}}(t) = \min(K_{\text{vel}}, K_{\text{vel}}^{(A)}(t), K_{\text{vel}}^{(B)}(t), K_{\text{vel}}^{(C)}(t)),$$

So now we can construct

$$K_{RW}^{\text{eff}}(t) = \begin{bmatrix} K_{\text{rot}} & 0 \\ 0 & K_{\text{vel}}^{\text{eff}}(t) \end{bmatrix}.$$

where K_{rot} is the same as from Question 8. If $K_{RW}^{\text{eff}}(t)$ is used, then the combined admittance law can now be expressed as:

$$\dot{\eta}_{RW}(t) = K_{RW}^{\text{eff}}(t) \zeta_W(t) / \Delta t, \quad \eta_{RW}(t) = K_{RW}^{\text{eff}}(t) \zeta_W(t)$$

where $\eta_{RW}(t) = [\alpha_{RW}^T(t), \varepsilon_{RW}^T(t)]^T$. This can then be used in the same cost function as before:

$$\Delta \mathbf{q}_{\text{cmd}}(t) = \arg \min_{\Delta \mathbf{q}} \|\eta_{RW}(t) - J_{RW}(\mathbf{q}(t)) \Delta \mathbf{q}\|^2,$$

If we hit possibility number 3, we still use $K_{RW}^{\text{eff}}(t)$ the same as calculated above, but now we have another constraint we have to follow while optimizing:

$$\dot{\mathbf{p}}_{CT}^{\text{cmd}} \cdot \nabla d_X > 0$$

If we use

$$\dot{\mathbf{p}}_{CT}^{\text{cmd}} = J_{p,CT}(\mathbf{q}(t)) \Delta \mathbf{q}_{\text{cmd}} / \Delta t$$

We can rewrite it into this inequality

$$\nabla d_X J_{p,CT}(\mathbf{q}(t)) \Delta \mathbf{q}_{\text{cmd}}(t) > 0$$

2 Question 10

To also constrain a portion of the tool shaft, we can modify Question 9 by running this algorithm not only for the tip of the tool, but also for sampled points going up as far along the tool shaft from the tip as we care about. Then when constructing $K_{RW}^{\text{eff}}(t)$, we must take the element-wise minimums of all the $K_{\text{vel}}^{(X)}(t)$, $X \in (A, B, C)$ for each of the sampled points. So if we want to constrain the entire shaft of length ρ_{tool} we can run Step 3 described in Question 9 for a point sampled at every ρ_{tool}/n length along the shaft, where a good n would be 5 to 10.

3 Question 11

To ensure safety in this scenario, we can modify Question 9 by first incorporating all the modifications in Question 10 as well. Then, we can modify the calculations further by treating the areas as larger than they actually are to be safe. If there is some error in F_{RC} ,

$$F_{RC}^* = F_{RC} \Delta F_{RC}(\eta_{RC}), \quad \text{cov}(\eta_{RC}) = C_{RC},$$

Let J_{RC}^* be the Jacobian that maps the effect of η_{RC} in F_{RC}^* to the resulting error in tool tip position calculation. This means the true tool tip in CT frame also has some error:

$$p_{CT}^* = p_{CT} + J_{RC}^* \eta_{RC}$$

Therefore the tool-tip position has covariance

$$C_{p_{CT}} = J_{RC}^* C_{RC} J_{RC}^{*T}$$

For each of the structures of interest $X \in \{A, B, C\}$ the true distance from each of the bodies to the tip now also has some error:

$$d_X^* = d_X + \nabla d_X (\mathbf{p}_{CT}^* - \mathbf{p}_{CT}) = d_X + \nabla d_X J_{RC}^* \eta_{RC}$$

So the variance of the distance error is

$$C_{d_X} = \nabla d_X C_p \nabla d_X^T$$

If we want to guarantee that the tool tip contacts any of the structures given this uncertainty, we replace the distance d_X used in Question 9 with a conservative “safe” distance for all $X \in (A, B, C)$ and each point along the tool shaft:

$$d_X^{\text{safe}} = d_X - k C_{d_X}$$

where k is a chosen safety factor like 4 or 5. This calculation should be done as soon as possible so that all following calculations that depend on distance use d_X^{safe} .

4 Question 12

If we have a distance sensor, we can use the Kalman filter. Let's first assume we are only concerned about the tool tip to make the explanation simpler. This can generalize to any point on the shaft very easily. We know the following:

$$F_{RC}^* = F_{RC} \Delta F_{RC}(\eta_{RC}), \quad cov(\eta_{RC}) = C_{RC}$$

The twist η_{RC} represents uncertainty in F_{RC}^* .

Let p_{CT} be nominal tool-tip position in CT coordinates, and let $d_{\min} = \text{GetMap}(p_{CT}, \text{"SDF_X"})$ and $\nabla d_{\min} = \text{GetGrad}(p_{CT}, \text{SDF_X})$ for whatever X is associated with the smallest d_x . This is because it is safe to assume that whatever structure we think we are closest to based off of the preop CT scan, is most likely the structure that the distance sensor's measurement is based off of.

From Question 11,

$$\begin{aligned} p_{CT}^* &= p_{CT} + J_{RC}^* \eta_{RC}, \\ d_{\min}^{true} &= d_{\min} + \nabla d_{\min} J_{RC}^* \eta_{RC}, \end{aligned}$$

The sensor returns a measurement of the minimum distance to the nearest anatomic structure,

$$z = d_{\min}^{true} + \delta \quad \delta \sim \mathcal{N}(0, \sigma_\delta^2),$$

which can be used for cases when $d_X(p_{CT}) > d_{\text{thresh}}$ for all $X \in \{A, B, C, D\}$. We can now create:

$$z = d_{\min} + H \eta_{RC} + \delta, \quad H = \nabla d_{\min}^T J_{RC}^*$$

We can treat η_{RC} as the state we are trying to estimate using a Kalman filter. The measurement residual is

$$r = z - d_{\min}$$

So the Kalman gain is

$$K = C_{RC}^{prior} H^T (H C_{RC}^{prior} H^T + \sigma_\delta^2)^{-1}$$

So then we can update $\hat{\eta}_{RC}$ and C_{RC} :

$$\hat{\eta}_{RC}^{post} = \hat{\eta}_{RC}^{prior} + Kr, \quad C_{RC}^{post} = (I - KH) C_{RC}^{prior}.$$

Now we can also calculate an updated F_{RC} :

$$F_{RC}^{post} = F_{RC}^{prior} \Delta F_{RC}(\hat{\eta}_{RC}^{post}),$$

We can continue updating F_{RC} using this algorithm in each cycle.

5 Question 13

In this case, let's first assume we only care about the point at the tool tip. Points along the tool shaft can be generalized to easily based off of Question 10. The high level is that we now want admittances such that it is still very hard to move the tool tip into areas A, B, and C, but it is free to move within D, as well as to and from D and E but no other transition. To do this, we can add one high level condition to check first:

1. If the tool tip is inside of E and outside of D (and all other structures), proceed as normal.
2. If the tool tip is inside of D and outside E, still proceed as in Question 9 but instead of only finding K_{vel}^X for $X \in (A, B, C)$, also do it for D as well. Then, instead of calculating a twist to move the tool outwardly normal to structure D, instead calculate a twist to move the tool inwardly normal to structure D (flip the sign on its SDF gradient). However, if the tool is moving towards structure E, then do not move the tool inwardly normal to structure D.

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