Assignment 2 - Kinematic walking controller

Hand-in: Apr. 1, 2022, 18:00 CEST

Leave your name, student ID, ETH email address and URL link to demo video here.

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In this assignment, we implement a kinematic walking controller for a legged robot!

Let's see the figure below.

figure: overview

Figure 1: The control pipeline of kinematic walking controller: 1) trajectory planning 2) computing desired joint angles by IK.

We start from five trajectories: 1 for base, and other 4 for feet. We plan the trajectories given target velocity of the robot base (body) and a timeline of the foot contacts (i.e. when and how long does a foot contact with a ground.) The details of how we plan the timeline of foot contacts, and how we generate the target trajectories, are out of scope of this assignment. But in Ex.3, we will have a sneak peek of trajectory planning procedure for the robot's base.

Our goal is tracking all the five target trajectories at the same time. We will simplify this problem by assuming the robot's base (somehow...) always perfectly tracks a target base trajectory. Then, we want to find **desired joint angles** for each leg that allow the foot (i.e. the end effector of individual leg) to reach a target position. We can effectively formulate this as an IK problem. Since the robot has four legs, by solving four IK problems, we can obtain desired configuration of the robot.

figure: overview2

Figure 2: Don't freak out!, this is a skeletal visualization of our Dogbot. Assuming the base is at the target position, we want to find **desired joint angles** for each leg that allow the foot (i.e. the end effector of individual leg) to reach its corresponding target position. Note. You can render skeletal view by Main Menu > Draw options > Draw Skeleton.

Hand-in and Evaluation

Once you complete this assignment you should hand in

- code pushed to your github repository.
- a short video demonstrating Ex.5 implementation. (Upload your video to YouTube and add a link to README.md)

The grading scheme is as follows

- baseline (80 %): based on your code. Each exercise is counted as 20%.
 - Ex.1-3 will be evaluated by an auto grading system after the deadline. (the test sets are not visible to you)
- advanced (20 %): based on unit test, code and demo videos. Each exercise is counted as 10%.
 - Ex.4 will be evaluated by an auto grading system after the deadline.
 (the test sets are not visible to you)
 - Ex.5 will be evaluated based on your demo video.

IMPORTANT: If your code is not built successfully, you will get **zero** point from this assignment. So make sure your code is built without any build/compile error.

IMPORTANT: If the system detect a suspected plagiarism case, you will get **zero** point from this assignment.

Please leave your questions on GitHub, so your colleagues also can join our discussions.

Exercises

Okay now let's do this step-by-step:)

Ex.1 Forward Kinematics (baseline)

In order to formulate an IK problem, firstly, we have to express the positions of a foot as functions of joint angles (and the base position). Formally speaking, given a *generalized coordinates* vector

equation: generalized coordinate

which is a concatenated vector of position of robot's base, orientation of robot's base and nj joint angles, we need to find a map between \mathbf{q} vector and end effector position $\mathbf{p} \text{EE}$ expressed in world coordinate frame.

equation: forward kinematics

In the previous lecture, we learned how to find this map by **forward kinematics**.

Code:

- Files:
 - $-\ {\tt src/libs/simAndControl/robot/GeneralizedCoordinatesRobotRepresenetation.cpp}$
- Functions:

- P3D getWorldCoordinates(const P3D &p, RB *rb)

Task:

• Implement forward kinematics for point **p**.

Details:

- GeneralizedCoordinatesRobotRepresentation represents the generalized coordinate vector \mathbf{q} .
- P3D getWorldCoordinates(const P3D &p, RB *rb) returns position of a point in world coordinate frame. The arguments p is a position of the point in rigidbody rb's coordinate frame.
- You may want to implement getCoordsInParentQIdxFrameAfterRotation(int qIndex, const P3D &pLocal) first: this function returns position of a point in the coordinate frame of the parent of qIdx after the DOF rotation has been applied.

Once you implement getWorldCoordinates function correctly, you will see green spheres around feet of the robot.

figure: forward kinematics

Figure 3: Check if your implementation is correct. You should see the green spheres around the robot's feet.

Ex.2-1 Inverse Kinematics - Jacobian by Finite Difference (baseline)

Okay, now we can express the position of the feet as a function of joint angles. It's time to formulate an IK problem: we want to find a generalized coordinate vector \mathbf{q} desired given end effector target position \mathbf{p} EEtarget.

equation: inverse kinematics

In the last assignment, we learn how to formulate the inverse kinematics problem as an optimization problem.

equation: inverse kinematics as optimization

We can solve this problem using gradient-descent method , Newton's method, or Gauss-Newton method. Whatever optimization method you choose, we need a Jacobian matrix of the feet point. Remember, Jacobian is a matrix of a vector-valued functions's first-order partial derivatives.

equation: jacobian

For now, we will use finite-difference (FD) for computing Jacobian. The idea of finite difference is simple. You give a small perturbation h around jth component of \mathbf{q} , and compute the (\mathbf{i}, \mathbf{j}) component as follows.

equation: jacobian by finite difference

Code:

- Files:
 - src/libs/simAndControl/robot/GeneralizedCoordinatesRobotRepresentation.cpp
- Functions:
 - void estimate_linear_jacobian(const P3D &p, RB *rb, Matrix &dpdq)

Task:

• Complete estimate_linear_jacobian functions that computes a Jacobian matrix of position/vector by FD.

Details:

• We use central difference with a perturbation h 0.0001.

Ex.2-2 Inverse Kinematics - IK Solver (baseline)

Now, it's time to implement a IK solver. Use **Gauss-Newton** methods we learned in the previous lecture.

We solve four independent IK problems (one for each leg.) Let's say **q**desired,i is a solution for ith feet.

equation: ik for ith foot

We can just solve each IK problem one by one and sum up the solutions to get a full desired generalized coordinates quesired.

equation: full desired generalized coordinate

Code:

- Files:
 - src/libs/kinematics/IK_solver.h
- Functions:
 - void solve(int nSteps = 10)

Task:

• Implement an IK solver based on the Gauss-Newton method.

Details:

- Use Jacobian matrix computed by gcrr.estimate_linear_jacobian(p, rb, dpdq) we implemented for Ex. 2-1.
- I left some useful hints on the source code. Please read it carefully.
- When you overwrite eigen matrix variable values in a single statement, you should use eval() function to prevent corruption. (note. this is caused by *aliasing*. See this for more details.) For example,

```
// overwrite Matrix J in a single statement
// use eval() function!
J = J.block(0,6,3,q.size() - 6).eval();
```

Let's see how the robot moves. Run locomotion app and press **Play** button (or just tap **SPACE** key of your keyboard). Do you see the robot trotting in place? Then you are on the right track!

figure: trotting in place

Ex.3 Trajectory Planning (baseline)

Now, let's give some velocity command. Press **ARROW** keys of your keyboard. You can increase/decrease target forward speed with up/down key, and increase/decrease target turning speed with left/right key. You can also change the target speed in the main menu.

figure: main menu

Oops! The base of the robot is not moving at all! Well, a robot trotting in place is already adorable enough, but this is not what we really want. We want to make the robot follow our input command.

Let's see what happens here. Although I'm giving 0.3 m/s forward speed command, the target trajectories (red for base, white for feet) are not updated accordingly. With a correct implementation, the trajectory should look like this:

figure: base trajectory

Code:

- Files:
 - src/libs/simAndControl/locomotion/LocomotionPlannerHelpers.h
- Class:
 - bFrameReferenceMotionPlan
- Function:
 - void generate(const bFrameState& startingbFrameState)

Task:

 Your task is completing generate function so that it updates future position and orientation according to our forward, sideways, turning speed commands.

Details:

• We plan the target base trajectory by numerical integration: we integrate forward, sideways and turning speed to get future position and orientation.

Once you finish this step, you can now control the robot with your keyboard.

By the way, planning the feet trajectories is a bit more tricky. I already implemented a feet trajectory planning strategy in our code base so that once you complete generate function, the feet trajectory is also updated by user commands. I will not explain more details today, but if you are interested,

please read the paper, Marc H. Raibert et al., Experiments in Balance with a 3D One-Legged Hopping Machine, 1984. Although this is a very simple and long-standing strategy, almost every state-of-the-art legged robot still uses this simple heuristic. (Sidenote. Marc Raibert, who was the group leader of Leg Laboratory, MIT, later founded Boston Dynamics Company in 1992.)

From now on, we will improve our kinematic walking controller.

Ex.4 Analytic Jacobian (advanced)

Okay, we compute Jacobian matrix with FD. But in general, FD is not only inaccurate but also very slow in terms of computation speed. Can we compute Jacobian matrix analytically? The answer is yes. With a small extra effort, we can derive analytic Jacobian matrix, and implement this into our code.

Code:

- Files:
 - src/libs/simAndControl/robot/GeneralizedCoordinatesRobotRepresentation.cpp
- Functions:
 - void compute_dpdq(const P3D &p, RB *rb, Matrix &dpdq)

Test: Compile and run src/test-a2/test.cpp. Test 4 should pass. Note that passing this test does not necessarily mean you get a full point Your implementation should give correct results for every case. We will auto-grade your implementation with bunch of test cases after the deadline.

Ex.5 Uneven Terrain (advanced)

Our robot can go anywhere in the *flat-earth world*. But, you know, our world is not flat at all. Now, we will make our robot walk on an bumpy terrain. Imagine you have a height map which gives you a height of the ground of given (x, z) coordinates (note that we use y-up axis convention i.e. y is a height of the ground.) To make the robot walk on this terrain, the easiest way is adding offset to y coordinates of each target positions.

figure: terrain

You have a full freedom to choose the terrain map you want to use: you can just create a bumpy terrain by adding some spheres in the scene as I've done here. Or you can download a landscape mesh file in .obj format (there's an example terrain obj file in data folder). Please figure out the best strategy to implement this by your own.

IMPORTANT: For Ex.5, create a new git branch named ex5 and push your code there while your implementation of Ex.1-4 still remains in main branch. If your Ex.5 implementation breaks Ex.1-4, you may not get full points from Ex.1-4.

Task:

- Make the robot walk on an uneven terrain (5%)
- Visualize your terrain in locomotion app (5%).
- Please record a video ~ 20 secs demonstrating your implementation. The robot in the video should walk on an uneven terrain, and your video should be visible enough to see how well your controller performs the task. Upload it to YouTube or any other video hosting website and add its link to on the top of this README.md. If I cannot play or access your video until the deadline, you won't get full points.
- Please adjust your camera zoom and perspective so that your robot is well
 visible in the video. You may not get full point if your robot is not clearly
 visible.

Hint:

- We want to give y-offset to IK targets. But remember, offset value for each target could be different.
- Do not change double groundHeight = 0; in SimpleLocomotionTrajectoryPlanner. This will merely give the same offset to every target trajectory. We want to give different offset for each individual foot and the base.

Final Note

Congratulations! You can now control the legged robot! Hooray!

Can we use our kinematic walking controller to a real robot? Well... unfortunately it's not very easy. In fact, working with a real robot is a completely different story because we need to take into account **Dynamics** of the robot. But, you know what? We have implemented fundamental building blocks of legged locomotion control. We can extend this idea to control a real robot someday!

By the way, can you make a guess why using kinematic controller for a real legged robot doesn't really work well in practice? If you are interested in, please leave your ideas on the GitHub issue.