

Robotic Arm : Grasping task: Documentation

File: demo.py

This file comprises of code which performs the data collection by running grasping simulations on the loaded object urdf.

Function name: move_arm_and_rotate :

This function would set the robotic arm at a random angle before initiating to approach the object

Function name: approach_object:

This function initiates the robotic arm to move towards the object

Function name: return_object_to_original_robot_pose:

This function allows the robotic arm to come back to its original position after attempting to grasp the underlying object.

Function name: reset_object:

This function resets the object to grasp for successive simulations

The above functions description can be known by referring
https://www.etedal.net/2020/04/pybullet-panda_2.html

Function name: get_grasp_img:

This function facilitates in fetching the snapshots at various instances during the grasp.

Function name: `grip_object`:

This function executes the robotic arm fingers to inch towards grasping the object.

During this function's execution, we ought to record

-> Pre-grasp robot coordinates

-> Post grasp robot coordinates

-> Robotic arm left finger's point cloud info (the information on contact points of robot's left finger)

-> Robotic arm right finger's point cloud info (the information on contact points of robot's right finger)

Each of these functions involve using inbuilt pybullet functions which can be referred from : http://dirkmittler.homeip.net/blend4web_ce/uranium/bullet/docs/pybullet_quickstartguide.pdf

Pre/Post grasp robot coordinates:

During capturing of robotic coordinates, we try to capture all the joint level information.

The robot under simulation is having 11 joints (numbered from 0 to 10)

For each of the joint, we fetch the joint information by calling the `getjointstate(robot_urdf, joint_number)` function.

The get joint state function returns the below information:

`getJointState` output

<code>jointPosition</code>	float	The position value of this joint.
<code>jointVelocity</code>	float	The velocity value of this joint.
<code>jointReactionForces</code>	list of 6 floats	There are the joint reaction forces, if a torque sensor is enabled for this joint. Without torque sensor, it is [0,0,0,0,0,0].
<code>appliedJointMotorTorque</code>	float	This is the motor torque applied during the last stepSimulation

We would need to fetch all these information and store it individually and thus

If `joint_info = getjointstate(robot_urdf, joint_number)`

Then:

`Joint_info[0]` -> joint position

`Joint_info[1]` -> joint velocity

`Joint_info[2]` -> joint reaction forces

Joint_info[3] -> motor torque

In addition to the joint level information, we capture the link level information of the link between the two joint fingers (joint fingers are joint 9 and joint 10, and the link between them is link 11)

We use the `getlinkstate(robot_urdf,link_number)` function which returns information as shown below:

`getLinkState` return values

linkWorldPosition	vec3, list of 3 floats	Cartesian position of center of mass
linkWorldOrientation	vec4, list of 4 floats	Cartesian orientation of center of mass, in quaternion [x,y,z,w]
localInertialFramePosition	vec3, list of 3 floats	local position offset of inertial frame (center of mass) to URDF link frame



localInertialFrameOrientation	vec4, list of 4 floats	local orientation (quaternion [x,y,z,w]) offset of the inertial frame to the URDF link frame.
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Out of these only the world position is our interested value, and thus we fetch `getlinkstate(robot_urdf,joint_number)[0]`

All these details totally in combination sum up to a robotic coordinate information, and we thus capture these values during the pre-grasp and post-grasp points.

Robotic arm point cloud info [left/right]

To capture the point cloud (the points of contact of the left and right gripper of the robot while trying to grasp an object), we leverage the `getcontactpoints` function which returns values as shown:

`getContactPoints` will return a list of contact points. Each contact point has the following fields:

contactFlag	int	reserved
bodyUniqueldA	int	body unique id of body A
bodyUniqueldB	int	body unique id of body B
linkIndexA	int	link index of body A, -1 for base
linkIndexB	int	link index of body B, -1 for base
positionOnA	vec3, list of 3 floats	contact position on A, in Cartesian world coordinates
positionOnB	vec3, list of 3 floats	contact position on B, in Cartesian world coordinates
contactNormalOnB	vec3, list of 3 floats	contact normal on B, pointing towards A
contactDistance	float	contact distance, positive for separation, negative for penetration
normalForce	float	normal force applied during the last 'stepSimulation'

If `contact_information = getcontactpoints(object_urdf,robot_urdf)`

Out of these the information of interest include:

positionOnA : obtained by `contact_information[5]`

positionOnB : obtained by `contact_information[6]`

Contact Normal: obtained by `contact_information[7]`

Contact distance obtained by `contact_information[8]`

Normal force : obtained by `contact_information[9]`

We try to see if the information belongs to joint 9 or joint 10 through the “linkIndexB” parameter (`contact_information[4]`)

Additionally we ought to ensure that the contact information is an actual information. To filter the actual information out, we ensure the contact distance (`contact_information[8]`) is in the range of $1e-3$.

To run the code:

```
python -u demo.py --object <object_name> --num_sim <no.of simulations to run> --folder_path  
<path where the json results need to be stored>
```

Eg:

```
python -u demo.py --object soap --num_sim 5 --folder_path grasp_soap
```

To view the json in a more readable format, if soap_data.json is the json output, run this command to turn the json into a readable json file

```
cat output_data/soap_data.json | python -mjson.tool > output_data/aligned_soap_data.json
```

The aligned output is stored in aligned_soap_data.json