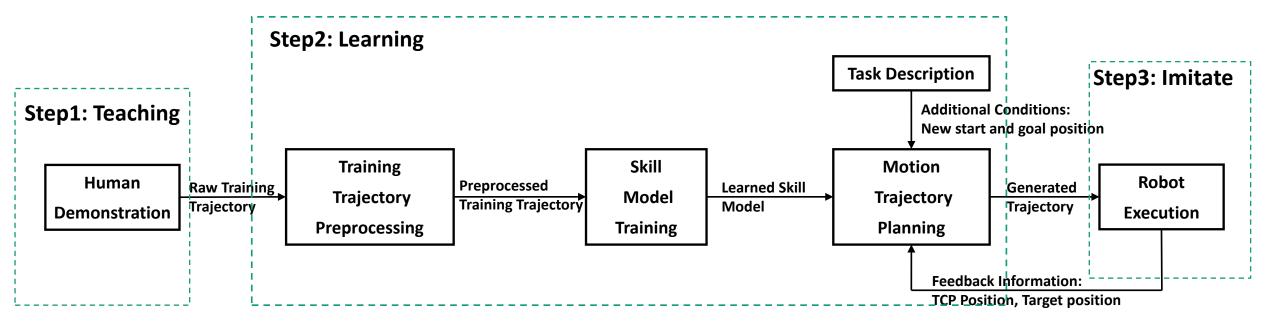
WORKFLOW OF LEARNING FROM DEMONSTRATION



System Overview of Learning from Demonstration



Workflow of Learning from Demonstration

Step 1 Teaching Process

- Human teacher demonstrate a motion under a motion capturing system (Openpose + ZED)
- The demonstrated motion trajectory is recoreded as the raw training trajectory

Step 2 Learning Process

- Preprocess the raw training trajectory with Kalman filter
- Find the keypoints of the training trajectory and divide the entire training trajectory into segments
- Use DMP to get the model training trajectory
- Set the new start and end position of the motion to generate a new motion trajectory

Step 3 Robot imitate human motion in simulation

UR10 execute the generated motion trajectory in ROS Gazebo



Step 1: Teaching Process

■ Software Setup

	Information:
Operating System	Windows 10
Dependences	OpenCV 4.1.1 CUDA 10.0 with cudnn 7.4.2 ZED SDK 3.1.2 Openpose C++ API

■ Usage

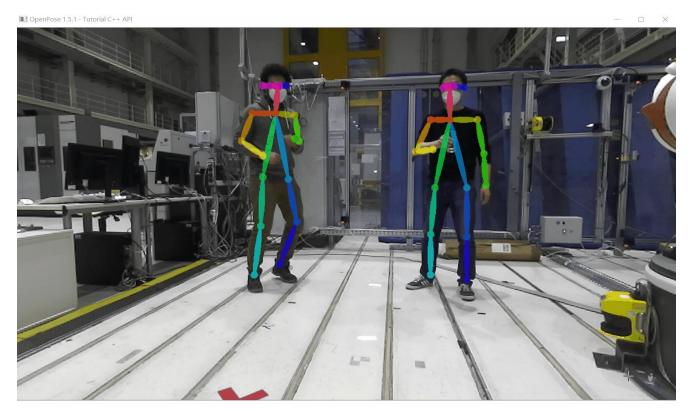
- 1. Input the name of the TXT.file used to save the demostated motion
- 2. Press 's' for start saving
- 3. Press 'q' for quite and save the record data into the TXT.file

■ Folder in Gitlab Project:



Step 1 Teaching Process

Example of observational teaching





An example of saved motion

Input: Human demonstration

Teaching Module

Output: Recorded motion trajectory



Step 2: Learning Process

■ Software Setup

	Information:
Operating System	Windows 10, Linux (Ubuntu 16.04)
Python version	3.6
dependences	numpy
	scipy
	matplotlib
	PyYAML

■ Usage

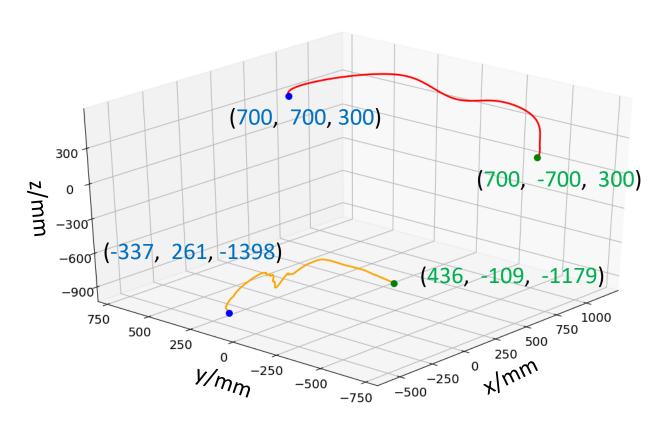
- Set parameters in ,config_param1.yaml'
- 2. Set the name of the training trajectory
- 3. Set the new start and end positions
- 4. Set the name of the generated trajectory
- 5. Run the python script 'DMP_Keypoints.py'

■ Folder in Gitlab Project:



Step 2 Learning Process

An example of generated motion trajectory (change start and goal position)



I traj gen 21.txt - 记事本 文件(F) 编辑(E) 格式(O) 查看(V) 帮助(H) 0.701000; 0.699247; 0.299124 0.702897; 0.697753; 0.297468 0.705590; 0.695530; 0.295127 0.708989; 0.692599; 0.292192 0.713007; 0.688980; 0.288759 0.717565 ; 0.684702 ; 0.284926 0.722590 ; 0.679798 ; 0.280805 0.728018; 0.674306; 0.276525 0.733790; 0.668275; 0.272241 0.739857; 0.661762; 0.268141 0.746177; 0.654834; 0.264444 0.752714; 0.647562; 0.261387 0.759437; 0.640018; 0.259201

Input: Record motion trajectory

Learning Module

Output: generated motion trajectory



Step 3: Robot imitate human motion in simulation

■ Software Setup

	Information:
Operating System	Linux (Ubuntu 16.04)
ROS version	Kinetic 1.12.14
Simulation	Gazebo
dependences	moveit universal_robot

■ Folder in Gitlab Project:

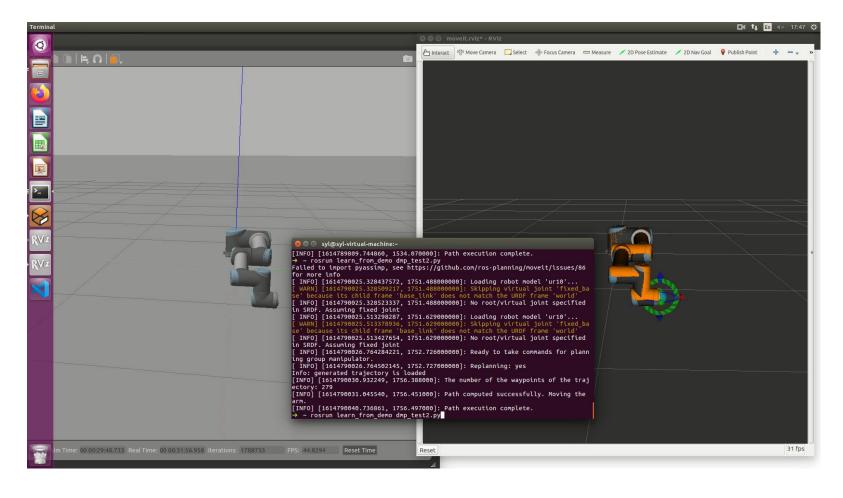
□ DMP_Python
■ OpenPose_ZED
learn_from_demo

Usage

- 1. Start the simulation environment Gazebo and needed ros node: Roslaunch learn from demo dmp ur10.launch
- Load the generated motion trajectory:
 Put the generated motion trajectory (txt.file) into folder
 open python script ,robot_imitate.py' and input the file name of the generated motion trajectory
- 3. UR10 Robot execute the generated motion trajectory rosrun learn from demo robot imitate.py



Example of Robot Execution



Input: generated motion trajectory

Robot Imitate



Structure of DMP_Python (Code for model training)

- Structure of folder DMP_Python
 - config param1.yaml # yaml file to set parameters
 - DMP_Standard.py # main function 1: for testing standard DMP (without keypoints)
 - DMP.py # functions for Dynamic Movement Primitive
 - plot result3D.py # Plot results
 - DMP_Keypoints.py # main function 2: for testing DMP with keypoints)
 - Training_Traj_Preprocessor.py # functions for training trajectory preprocessing
 - Model_Training.py # functions for model training
 - Skill Model Optimizer.py # functions for adjusting the learned motion model
 - Trajectory_Generator.py # functions for generating trajectory by using adjusted learned model
 - plot function.py
 - Load_Save_Data.py # Load training Data
 - kalman filter.py # Kalman Filter
 - Obstacle_avoid.py # functions for testing obstacle avoidance
 - variance_calculation.py # Calculating the variance between training and generated motion trajectory



Parameters for model training (Set in the config_param1.yaml)

Parameter	Information (Comments)
nb_Samples	Number of demonstrations
nb_Data	Length of training data after resampling (interpolated training data)
KP	Stiffness gain (Spring Damped System)
alpha	Decay factor (Phase Variable)
dt	Duration of time step
nbStates	Number of activation function (i.e. number of States in the GMM)
threshold_keypoint	threshold used for select keypoints: remove the keypoints locate too close
threshold_goal	threshold for judging if the end position is close enough to the goal position
data_handwrite_path	path and name of handwriting training trajectory
data_handover_path	path and name of handover training trajectory (from OpenPose+ZED)
traj_gen_path	path and name for saving the generated motion trajectory

- The new start and end position of generated motion trajectory can be set in the **DMP_standard.py**
- Load handwriting or handover data can select in the main function in DMP_standard.py and DMP_Keypoints.py
- Dimension (x,y,z) of Keypoints found can be set in the function **keypoint_finder** in script **Training_Traj_Preprocessor.py** (line 219)
- When using the real data (get from openpose) use Kalman filter before model training (DMP_Keypoints.py line 290)