

# CODE FOR BETTER \_\_

Hackathon



# Manipulation and Skill Learning and Generalizing

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#### **Learning from Demonstration**

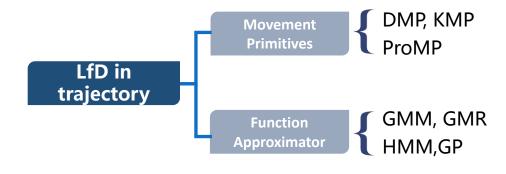
**Definition**: Learning from demonstration (LfD) is the paradigm in which robots acquire new skills by learning to imitate an expert in the context of robotics and automation.

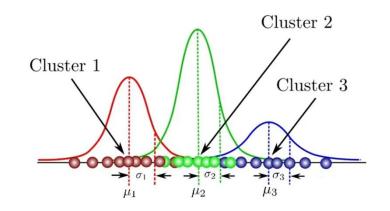
- □ Kinesthetic teaching
- □ Teleoperation
- □ Passive observation
- Active and Interactive Demonstrations















#### **Learning from Demonstration**

- **□** Dynamic Movement Primitive
  - Nonlinear Dynamic System:

$$\tau \begin{bmatrix} \ddot{z} \\ \dot{x} \end{bmatrix} = \begin{bmatrix} \alpha_z (\beta_z (g - z) - \dot{z}) + f(x, g) \\ -\alpha_x x \end{bmatrix}$$

- Forcing Function:

$$f(x,g) = \left(\sum_{i=1}^{N} \psi_i \cdot \frac{\omega_i}{\sum_{i=1}^{N} \psi_i}\right) x(g - z_0)$$

- Basis Function Weight:

$$\omega_i = \frac{s\boldsymbol{\psi}_i \boldsymbol{f}_{target}}{s\boldsymbol{\psi}_i \boldsymbol{s}^T}.$$

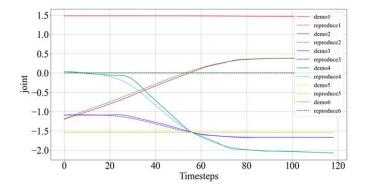
- □ Gaussian Mixture Model and Gaussian Mixture Regression (GMM-GMR)
  - Joint Probability Distribution:

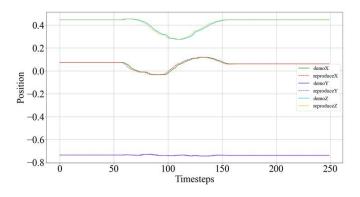
$$P(s,\xi) \sim \sum_{c=1}^{C} \pi_c N\left(\mu_c, \sum_{c}\right)$$

- Gaussian Component:

$$\mu_c = \left[ \begin{matrix} \mu_{s,c} \\ \mu_{\xi,c} \end{matrix} \right]$$

$$\sum_{c} = \begin{bmatrix} \sum_{ss,c} & \sum_{s\xi,c} \\ \sum_{\xi s,c} & \sum_{\xi \xi,c} \end{bmatrix}$$

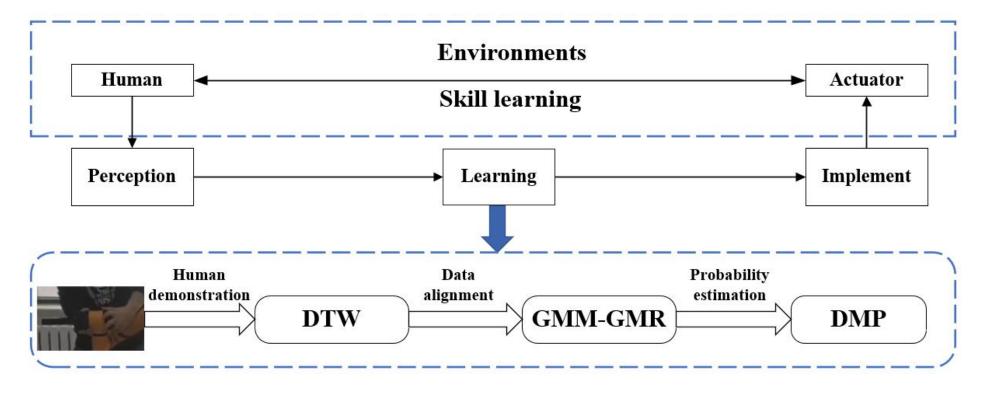








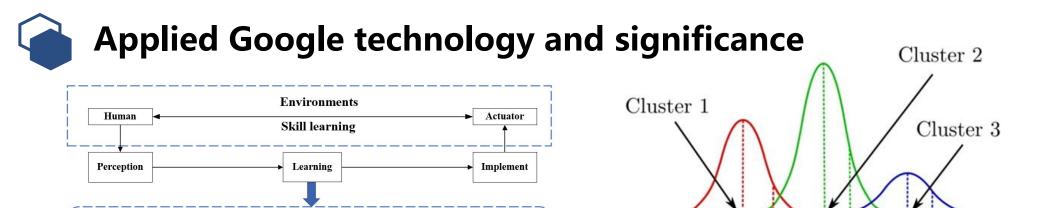
#### **Overview**



**Task I:** To accurately learn motor skills from human demonstrations and experiments data, data alignment is performed on the teaching data through Dynamic Time Warping (DTW).

**Task II:** Then, the data is analyzed using the statistical method Gaussian Mixture Model and Gaussian Mixture Regression (GMM-GMR). Ultimately, Dynamic Movement Primitives (DMPs) learn motor skills from human demonstrations and experiments data.





**DMP** 

estimation

**GMM-GMR** 

The machine learning model uses supervised learning and deep reinforcement learning, and the dynamic motion primitive algorithm uses locally weighted linear regression to learn and generalize the model. GMM-GMR is a set of functions to train a Gaussian Mixture Model (GMM) and retrieve generalized data through Gaussian Mixture Regression (GMR). It allows to encode efficiently any dataset in Gaussian Mixture Model (GMM) through the use of an Expectation-Maximization (EM) iterative learning algorithms. By using this model, Gaussian Mixture Regression (GMR) can then be used to retrieve partial output data by specifying the desired inputs. It then acts as a generalization process that computes conditional probability with respect to partially observed data.

Through the learning algorithm, the complex tasks can be learned and modeled, the learning and generalization of tasks can be realized, and the skills scenes in life, such as opening the door, locking the screw and other tasks, can be handled, so as to improve the cognition, learning and realization ability of skill operation.



# Thanks for your listening!

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