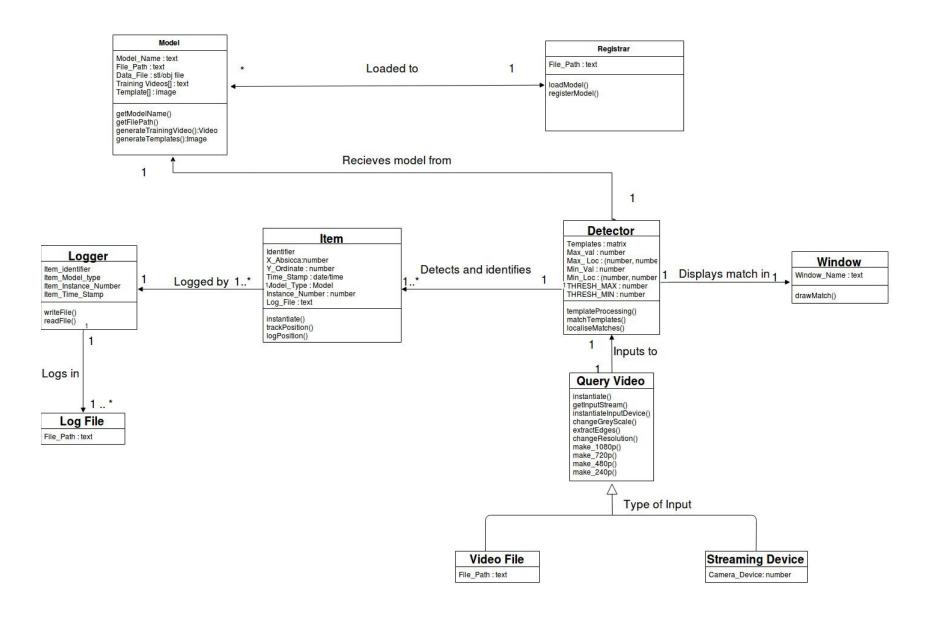
Product Design Team 35

Members

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Object Detector using 3D Mesh Models



Class Number	Class Name	State / Behavior
		Model State
1		 Model_Name: text File_Path: text Data_File: stl/obj file of the model Training_Videos: list of paths to the training videos extracted from the Data_File Templates: Img files
		Model Behavior
	Model	 getModelName() Register the name of the model. getFilePath() Gets the file path from the user. generateTrainingVideo() Generate a set of edge template training videos from predefined orientations of the mesh model. generateTemplates() Get all the edge templates of the model from the Training_Videos, of the mesh model.
		Registrar State
		File_Path : text
2		Registrar Behavior
	Registrar	 loadModel() Load the mesh model into the system. registerModel() Register the mesh model into the database of the system and label it according to the identifier (Model_Name) provided by the Model class.

		Item State
		 Identifier X_Abscissa : number Y_Ordinate : number Time_Stamp : date/time Model_Type : Model Instance_Number : number Log_File : text
		Item Behavior
3	ltem	 instantiate() Constructor for initialize the object detected in the video screen, and labeling it under a Model_Type and giving it an instance number. trackPosition() Modify the X_Abscissa and Y_Ordinate values for the object with respect to a constant time frame. logPosition() Write the position of the object into the Log_File of the Item and mention the time stamp as well.
		Logger State
		 Item_Identifier Item_Model_Type Item_Instance_Number Item_Time_Stamp
		Logger Behavior
4	Logger	writeFile() Get the values returned by logPosition method from the Item class, create appropriate directory structure and log file if not already present and write into the file. readFile() Read the position and time stamp values for a particular

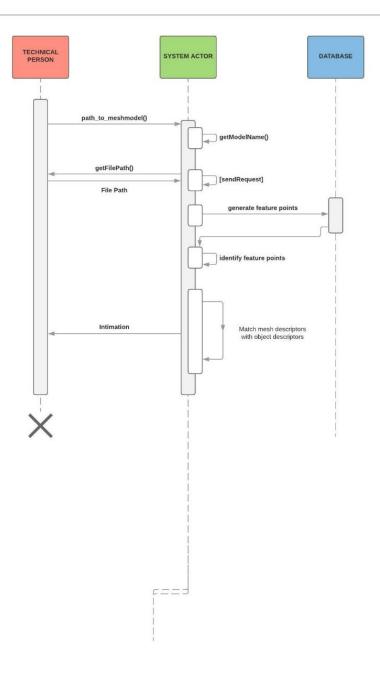
		item of a model type and instance number.
		QueryVideo State
		File_Path : textCamera_Device : number
		QueryVideo Behavior
5	QueryVideo	 instantiate() Constructor for the input video stream. Initialize the File_Path of the video stream and Camera_Device index. getInputStream() Get the query input stream from the Video_Path or the Camera_Device. instantiateInputDevice()

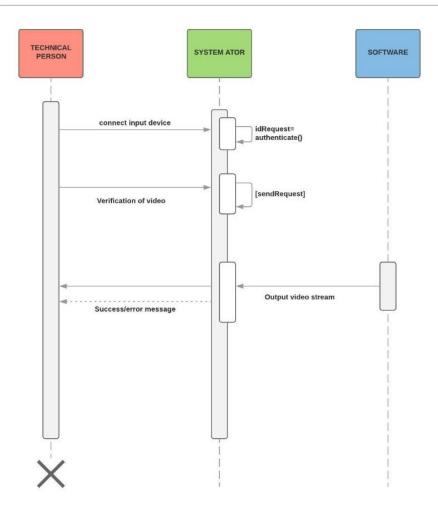
		Detector State
		 Templates: matrix Max_Val: number Max_Loc: (number,number) Min_Val: number Min_Loc: (number,number) THRESH_MAX: number THRESH_MIN: number tH: number tW: number
		Detector Behavior
6	Detector	templateProcessing() Retrieve templates from the corresponding directory. Convert image to grayscale. Apply Canny edge detector on each template. Display the result.
		matchTemplates()
		localiseMatches() Calculate bounding box on the region of interest specified by matchTemplates()
		drawMatch() Draw bounding box based on the coordinates specified by localiseMatches()
		Window State
7	Window	 Name: text Dimension: (number, number) Match_Coordinates: (number, number)

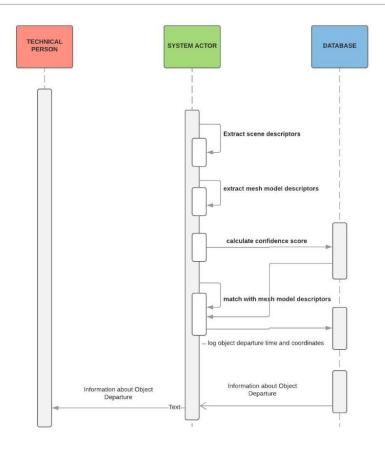
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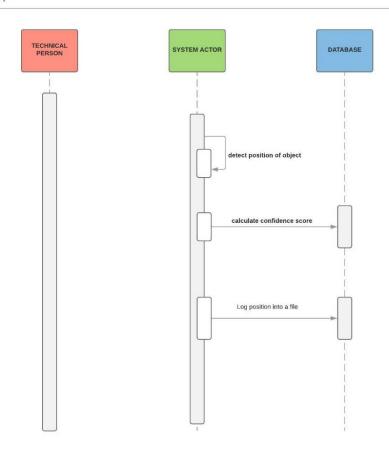
Sequence Diagram(s)

UC1









Design Rationale

First Design

Used a feature based keypoint detector and extract feature descriptors using SIFT/SURF/ORB. This information is extracted for the 3D Mesh Model and for each frame of the video stream. An appropriate matching algorithm is used to match the keypoints of the frame and the most appropriate model identified out of the set of mesh models.

<u>Advantage</u>

- Fast, as the algorithms used were highly optimised for OpenCV.
- Scaled well for multiple objects
- Invariant to rotation, lighting, blur, scale changes of the objects.

Disadvantage

- Time taken for the whole process was slower than what was required.
- Detection and extraction required the objects to have distinct textures and the mesh models does not have a texture. (Major drawback)

The main issue with the first design was that it required the objects to have texture. But mesh models do not have any texture associated with it. Hence, we had to opt for a different approach.

Second Design

After doing appropriate research, we chose multi-scale template matching to proceed with our project. This method used used the projection of the 3D mesh model onto a 2D plane at a particular orientation as a template. This template was used to match the models seen in the frame with the 3D models we had.

Advantage

- Scale invariant
- Did not require the 3D models to have textures.
- Moderate tolerance to blur and lighting effects.

Disadvantage

- Method does not scale well for multiple objects
- As multiple objects are introduced, the software becomes slow.
- Algorithm is inherently rotation invariant.

Since our project required our final design to track objects in any orientation, we had to make our software invariant to rotation effects. For this, we projected the 3D mesh model onto a 2D plane in different orientations. Each of these projections formed a set for that particular model. The set of projections were used to identify and track the respective model in each frame of the video. Since this method is computationally expensive, the software becomes slow on including multiple objects to track. Hence, we will be using a design which is optimised for multi-object tracking.