

# Progress Report

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## 1 Specific Research Goals

- Pose Estimation: Implement and improve.
- NBV-Grasping.
- Pose estimation survey.
- Universal pose estimation.

## 2 To Do

- Catch up on my reading list.
- Pose Estimation:
  - Implement QEKF.
  - Implement key point feature extraction: ORB, SIFT, SURF.
  - Survey: I need start working on this.
  - Survey implementation: Classical, 2-stage, and end-to-end methods. PnP, QuEst, PVNet and else.
- NBV-Grasping:
  - Update URDF and Xacro files for UR5e to include sensor, sensor mount (with offset), and the gripper. - On-going.
  - Add movement constraints for tables and scenes.
  - Write two IK functions for gripper and sensor, one for each. It should plug-in with MoveIt configurator.
  - Research and implement point-cloud data to training TensorFlow models.
  - Learn and implement GraspIt package.
- MSI Fellowship: On pause.
- Look into methods of generating uncertainty data for pose estimation.

### 3 Reading List

- Vision-based robotic grasping from object localization, object pose estimation to grasp estimation for parallel grippers - a review [1] - On-going.
- Leveraging feature uncertainty in the pnp problem [2].
- Normalized objects [3].
- Berk Calli's YCB [4].
- NASA papers [5].
- Roadmap [6].

### 4 Progress

The following items are listed in the order of priority:

- UTARI: This week I worked on setting up OpenCV git submodules and building it from source. I had to repeat the process multiple times and made sure to learn it and document it. Moreover, I am almost done debugging catkin-make process that build and bind ROS packages for this project. So far, I have been able to debug everything myself. This is a major achievement as it aligns with my robotic goals. I have gone over angular and linear EKF modules with Cody and currently I am trying to compile it and run the ROS packages to able to test it on data. Next, I will finish debugging Catkin-make process. It currently build up to 96 percent and is missing a dependency. Moreover, I need to reach out Asif for the latest QEKF code, I asked him last night and he was making some improvements. In deriving the H and F matrices there are so dimension reduction that I find intriguing.
- In [7], the authors introduce a novel framework as well as a Python library for representing learned high dimensional deep neural model using strong inductive basis. They train Graph Neural Network (GNN) to develop sparse latent representation of learned features. This is done by deploying genetic algorithm to test and evaluate variations of short closed-form symbolic expression describing each learned feature. This forces the deep neural network to represent high dimensional data in

a much lower dimension representation while maintaining accuracy. This is very similar to how we describe physical phenomena using math, we tailor math to match physics. The authors were able to extract Newtonian and Hamiltonian dynamics equations in their experiments. Newtonian dynamics describe forces acting on objects which is summarized best by  $F = ma$ . Hamiltonian dynamics describe a system in terms of total energy with the constraint that total energy is always conserved. Hamiltonian is best described by  $H = E_k + E_p$ .

- NASA MSI Fellowship: Need to read more NASA papers.
- PyTorch Tutorials: Transfer learning.
- NBV Grasping Project: I did not get to this project at all this week. Starting this week, I am dedicating Tuesdays this project. Please let me know what you think of this. I think I am most productive when I am on strict schedule.
- PE Survey: Starting this week, I will dedicate Mondays to write this paper. Please let me know if you think it is a good idea.
- SD Team: No update.
- EE Autonobots: Cody has ordered 4 TurtleBot-3 robots and student members are to pick a date to meet up and if I can I will go help and mentor them.

## 5 Immediate Plans - Summer 2021:

The following items are listed in the order of priority:

- UTARI: Dr. Gans' pose and velocity estimation paper.
- NBV-Grasping:
- Pose estimation: Survey paper.

## 6 Intermediate Goals - Fall 2021:

- Pose estimation: I must be finished with implementation, perhaps make some improvements, and should be working on a paper for ICRA or CVPR.

- Scene understanding and active learning: After pose estimation, I want to expand my research into scene understanding and active learning in the context of advanced manufacturing.
- ARIAC: Once I am up to speed, I will do the ARIAC workshops/tutorials and will talk to Jerry about possible contributions.

## References

- [1] G. Du, K. Wang, S. Lian, and K. Zhao, “Vision-based robotic grasping from object localization, object pose estimation to grasp estimation for parallel grippers: a review,” *Artificial Intelligence Review*, pp. 1–58, 2020.
- [2] L. Ferraz Colomina, X. Binefa, and F. Moreno-Noguer, “Leveraging feature uncertainty in the pnp problem,” in *Proceedings of the BMVC 2014 British Machine Vision Conference*, pp. 1–13, 2014.
- [3] H. Wang, S. Sridhar, J. Huang, J. Valentin, S. Song, and L. J. Guibas, “Normalized object coordinate space for category-level 6d object pose and size estimation,” in *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition (CVPR)*, June 2019.
- [4] B. Calli, A. Singh, A. Walsman, S. Srinivasa, P. Abbeel, and A. M. Dollar, “The ycb object and model set: Towards common benchmarks for manipulation research,” in *2015 international conference on advanced robotics (ICAR)*, pp. 510–517, IEEE, 2015.
- [5] NASA, “Nasa technical reports server (ntrs).” <https://ntrs.nasa.gov/>, 2020. (Accessed on 05/07/2021).
- [6] USA-CRA, “roadmap-2020.pdf.” <https://cra.org/ccc/wp-content/uploads/sites/2/2020/10/roadmap-2020.pdf>, 2020. (Accessed on 04/30/2021).
- [7] M. Cranmer, A. Sanchez-Gonzalez, P. Battaglia, R. Xu, K. Cranmer, D. Spergel, and S. Ho, “Discovering symbolic models from deep learning with inductive biases,” *arXiv preprint arXiv:2006.11287*, 2020.