

# Lab 12

# Learning Picking

Song Chaoyang

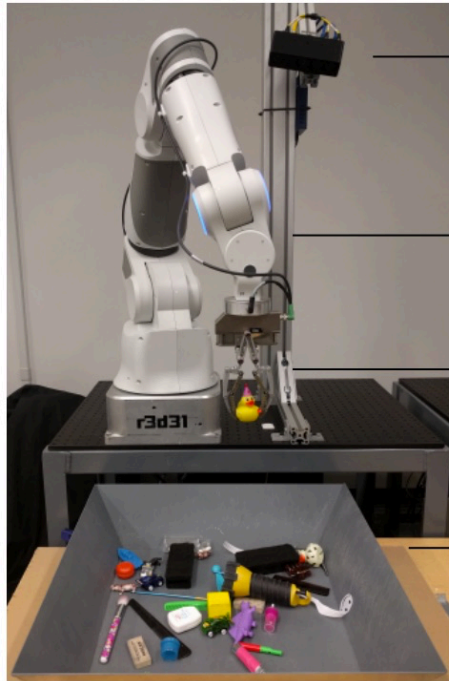
Assistant Professor

Department of Mechanical and Energy Engineering

[songcy@sustech.edu.cn](mailto:songcy@sustech.edu.cn)

# Common Setups of Vision-based Robotic Grasping

*From academic research to industrial applications*

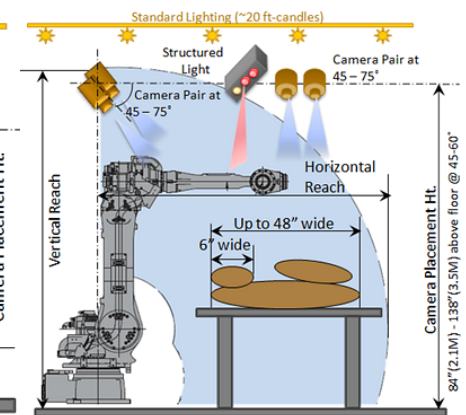
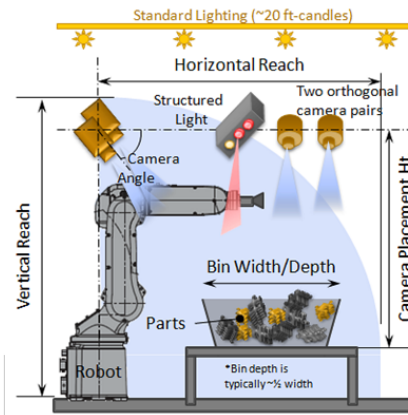
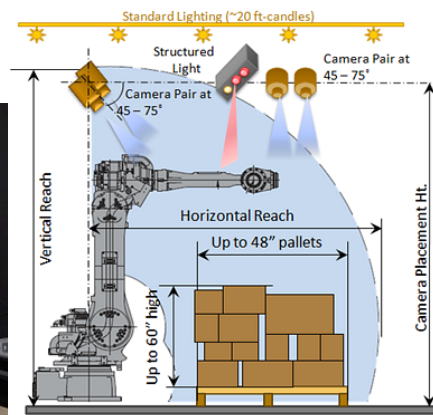
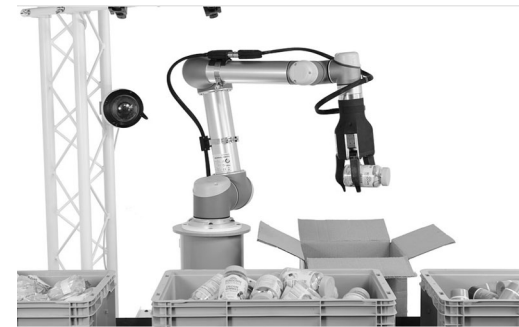
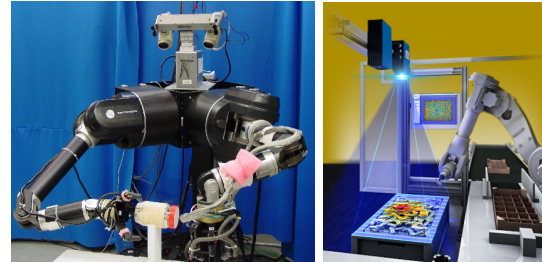


monocular RGB camera

7 DoF robotic manipulator

2-finger gripper

object bin



# Challenges in Setting Up A Robotic Cell for Learning

## *Shareability and Reproducibility*

- **Hardware: still expensive**
  - Arm: AUBO (~90K), UR5/Franka (~180K), UR10 (~400K)
  - Camera: Photoneo (~120K)
  - Gripper: Robotiq 3F (~250K)/2F (~50K), OnRobot (~50K)
  - Cheaper Alternatives:
    - Open Arms (~?), Realsense (~1.5K), Suction Cup (~1K)
- **Software: steep learning curve & expensive**
  - ROS, manufacturer specific, RoboDK, ...
- **Integration: time consuming to setup right (~H+S)**
  - Cost of data per sample
    - Total Cost (H+S+I) / Cycle Time (~5-10sec): not negligible, yet

# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - $90K+1K+1.5K+5K +5K < ¥100K$  per cell
- Yearly operation time per robot:  $\sim 3840$  h/year (in motion)
  - 2 shifts of high use or three shifts of moderate use with moderate demands (a portion of the robot's motions with full load, a portion with tool only)
- Cost per robot cell on an hourly basis:  $\sim ¥ 26$  cell/hour
  - Assume 10 sec per grasp
    - 1 hour usage =  $60\text{min/h} * 60\text{sec/min} / 10\text{sec/grasp} = 360$  grasp/h
- Cost per cell per grasp:  $\sim ¥ 0.07/\text{grasp}$ 
  - Disregard electricity and maintenance, and assume every grasp counts
- To generate a dataset of 100K samples
  - 1 robot cell:  $100K \text{ sample/dataset} * ¥ 0.07/\text{grasp} = ¥ 7K/\text{dataset}$
  - Cost of Time/Cell?:  $100K \text{ sample/dataset} / 360 \text{ grasp/h} \sim 277\text{h} \sim 1\text{mth}$  ☹️

# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - $90K+1K+1.5K+5K +5K < ¥100K$  per cell

What if

- A different robot arm is needed?
- A different end-effector is needed?
- A different camera system is needed?
  - A new sensor is needed?

For example:

UR5 (\*2) + OnRobot RG2 (\*50) + OnRobot FT (+50K) + Photoneo M (\*80) + Man Hour (\*2)

- $90K*2+1K*50+(50K)+1.5K*80+5K +5K*2 \sim 415K (>100K*4)$

What if it is the case of an industrial arm?

Which part could possibly be remain unchanged?

# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - 90K+1K+1.5K+5K +5K < ¥100K per cell
- Yearly operation time per robot: ~3840 h/year (in motion)
  - 2 shifts of high use or three shifts of moderate use with moderate demands (a portion of the robot's motions with full load, a portion with tool only)

## Daily Checks:

- Clean sensors and optics within the robot cell and visually check component parts for damage before daily operation
- Check Mechanical Unit for grease/oil leakage or exudation

## First 1-month (320 hours) Checks:

- Ventilation of the robot controller' cooling fans – keep these clean for efficient air flow

## First 3-month (960 hours) Checks:

- Check Mechanical Unit Cables – free from severe kinks, pinch points, cuts or tears in the wire insulation, check for secure terminations
- Retighten external main mounting bolts
- Clean chips and debris from Mechanical Unit
- Check the end effector or end of arm tool and its' mounting bolts or screws, tighten appropriately
- Check robot connection cables (T.P. Cable, EE Cable, Vision Cables) – free from severe kinks, pinch points, cuts or tears in the wire insulation, check for

secure terminations

## 1-year (3,840 hours) Checks:

- Grease balancer housing (if so equipped)

## 1.5-year (5,760 hours) Checks:

- Replace batteries in the mechanical unit (these are backup batteries to maintain the robot's factory alignments – don't let these batteries die or you may be in for some unpleasant work to re-zero your robot)

## 3-year (11,520 hours) Checks:

- Replace the grease of each axis. This is not an idle warning – use only the grease recommended by your robot's manufacturer – the robot controls are sensitive and grease can affect robot motion and operation.

## 4-year (15,360 hours) Checks:

- Replace the Mechanical Unit cables
- Replace the lithium battery in the CPU

**Robots don't ask for benefits, but still need maintenance**

<https://motioncontrolsrobotics.com/how-often-do-robots-need-preventative-maintenance/>

# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - 90K+1K+1.5K+5K +5K < ¥100K per cell
- Yearly operation time per robot: ~3840 h/year (in motion)
  - 2 shifts of high use or three shifts of moderate use with moderate demands (a portion of the robot's motions with full load, a portion with tool only)
- Cost per robot cell on an hourly basis: ~ ¥ 26 cell/hour
  - Assume 10 sec per grasp
    - 1 hour usage = 60min/h \* 60sec/min / 10sec/grasp = 360 grasp/h

Cost per robot cell on an hourly basis: ~ ¥ 26(\*4) cell/hour

Cycle time is task-specific, and may require advanced control

- Advanced motion planning (additional cost, like mujin)
- Waiting for other machines to finish workflow (convey belt, CNC machines, ...)
  - In short, most robotic cells waits longer to execute each grasp

# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - 90K+1K+1.5K+5K +5K < ¥100K per cell
- Yearly operation time per robot: ~3840 h/year (in motion)
  - 2 shifts of high use or three shifts of moderate use with moderate demands (a portion of the robot's motions with full load, a portion with tool only)
- Cost per robot cell on an hourly basis: ~ ¥ 26 cell/hour
  - Assume 10 sec per grasp
    - 1 hour usage = 60min/h \* 60sec/min / 10sec/grasp = 360 grasp/h
- Cost per cell per grasp: ~ ¥ 0.07/grasp
  - Disregard electricity and maintenance, and assume every grasp counts

Cost per cell per grasp: ~ ¥ 0.07/grasp

- \*4?
- Or \*4\*2?
- Or \*4\*?



# Cost Breakdown of Collecting 100k Samples

*A relatively cheap version in an ideal setup*

- Robot System (when everything is well-planned):
  - AUBO i5 + Suction Cup + Realsense + Mechanical + Man Hour
  - 90K+1K+1.5K+5K +5K < ¥100K per cell
- Yearly operation time per robot: ~3840 h/year (in motion)
  - 2 shifts of high use or three shifts of moderate use with moderate demands (a portion of the robot's motions with full load, a portion with tool only)
- Cost per robot cell on an hourly basis: ~ ¥ 26 cell/hour
  - Assume 10 sec per grasp
    - 1 hour usage = 60min/h \* 60sec/min / 10sec/grasp = 360 grasp/h
- Cost per cell per grasp: ~ ¥ 0.07/grasp
  - Disregard electricity and maintenance, and assume every grasp counts

Probably cost more than ¥ 7K/dataset of 100 K samples per robot cell,  
And takes longer than 1mth in practice ... ☹️☹️☹️☹️

- To generate a dataset of 100K samples
  - 1 robot cell: 100K sample/dataset \* ¥ 0.07/grasp = ¥ 7K/dataset
  - Cost of Time/Cell?: 100K sample/dataset / 360 grasp/h ~ 277h ~ 1mth ☹️

# The Arcade Claw Machine Game

Structured grasping controlled by human player for a reward.

- Insert coins (cost per grasp), countdown starts (~30 sec)
- All planning done by human player
  - Eyes as vision sensor, observe from multiple angles to calibrate
  - Brain as the processing unit, weighing between probability of grasp against value of the prize
  - Hand on joystick to plan claw motion in x-y plane
  - Hand on button to execute the grasp
- Pick execution by the machine, dropping claw in z-axis
- Return the claw to drop coordinate, claws open and toys will drop if grasp success

Scene Segmentation

- What we are going to interact with ...

Object Recognition

- Representation & Classification

Pose Estimation

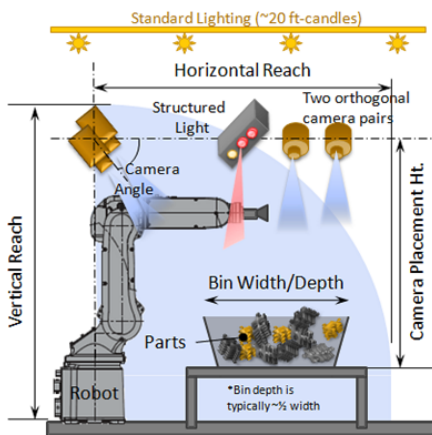
- Object & Picker

Pick Planning

- Picker & Arm

Pick Execution

- MoveIt & PickIt



# Advantages of the ACM

## Accessibility and Flexibility towards Shareability and Reproducibility

	Arcade Claw	Robotic Cell
Accessibility	Everyone	Professionals
Arm	Cartesian (3D)	Serial (6D)
Gripper	3 Finger (loose)	End-effectors
Vision	Human Eye	Camera
Control	Joystick	Controller
Decision	Human	Algorithms
Object	Almost anything	

### • Object Relocation

- Handling object from one location to another
- Pick & Place | Machine Tending | Packing and Palletizing

### • Material Releasing

- Releasing material from the robot to the target location
- Gluing | Dispensing | Welding | Screwdriving

### • Material Removal

- Removing material from the target object using the robot
- Polishing | Grinding | Deburring

### • Information Gathering

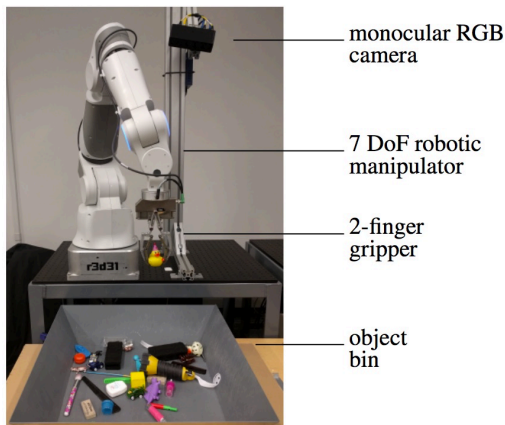
- Collecting information using sensors attached to the robot
- Quality Inspection

### Robot Movement

- Fixed
- Patterned
- Continuous
- Changing
- Random

### Critical Components

- End-Effector
- Vision System
- I/O Interfacing
- Conveyor Tracking
- Force-Torque Sensor
- External Jig
- Protective Suit



# DeepClaw Setup

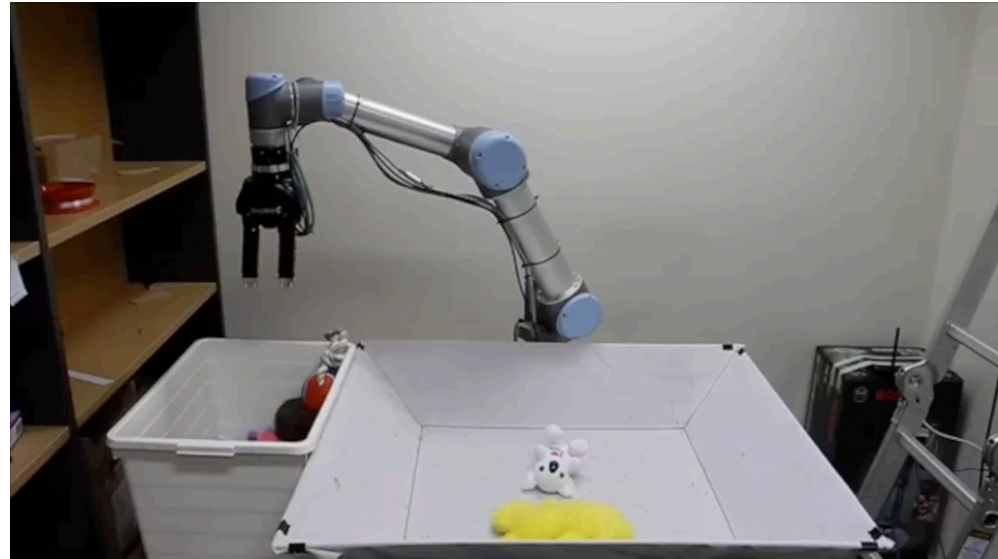
## Task Decomposition

- Start
  - Preliminary Preparation
  - Safety Check
  - From Idle Position ( $x_{idle}$ ,  $y_{idle}$ ,  $z_{idle}=0$ ,  $\theta_{idle}$ , open)
- Operation
  - Blind Grips (~5k in 1 week)
  - Blind Learning with all Blind Grips attempted in 1 week
  - Daily Grips (~1k in 1 day)
  - Daily Learning with all Daily Grips attempted in 1 day
  - Repeat Daily Grips Daily Learning for 8 weeks
- End
  - Back to Idle Position ( $x_{idle}$ ,  $y_{idle}$ ,  $z_{idle}=0$ ,  $\theta_{idle}$ , open)
  - Learning and Grasping Result Summary

## Start Stage

- System Setup
  - World => Pedestal => Arm => FT Sensor => Gripper
  - World => Desk => Tray => Objects (to be picked)
  - World => Desk => Bin => Objects (to be placed)
  - World => Camera Kinect (main rgd-d camera for training, focusing on the tray)
  - World => Camera LifeCam (auxiliary rgd camera for training, focusing on the tray)
  - World => Camera Canon (record the whole experiment, focusing on the whole robot operation)

AncoraSIR.com

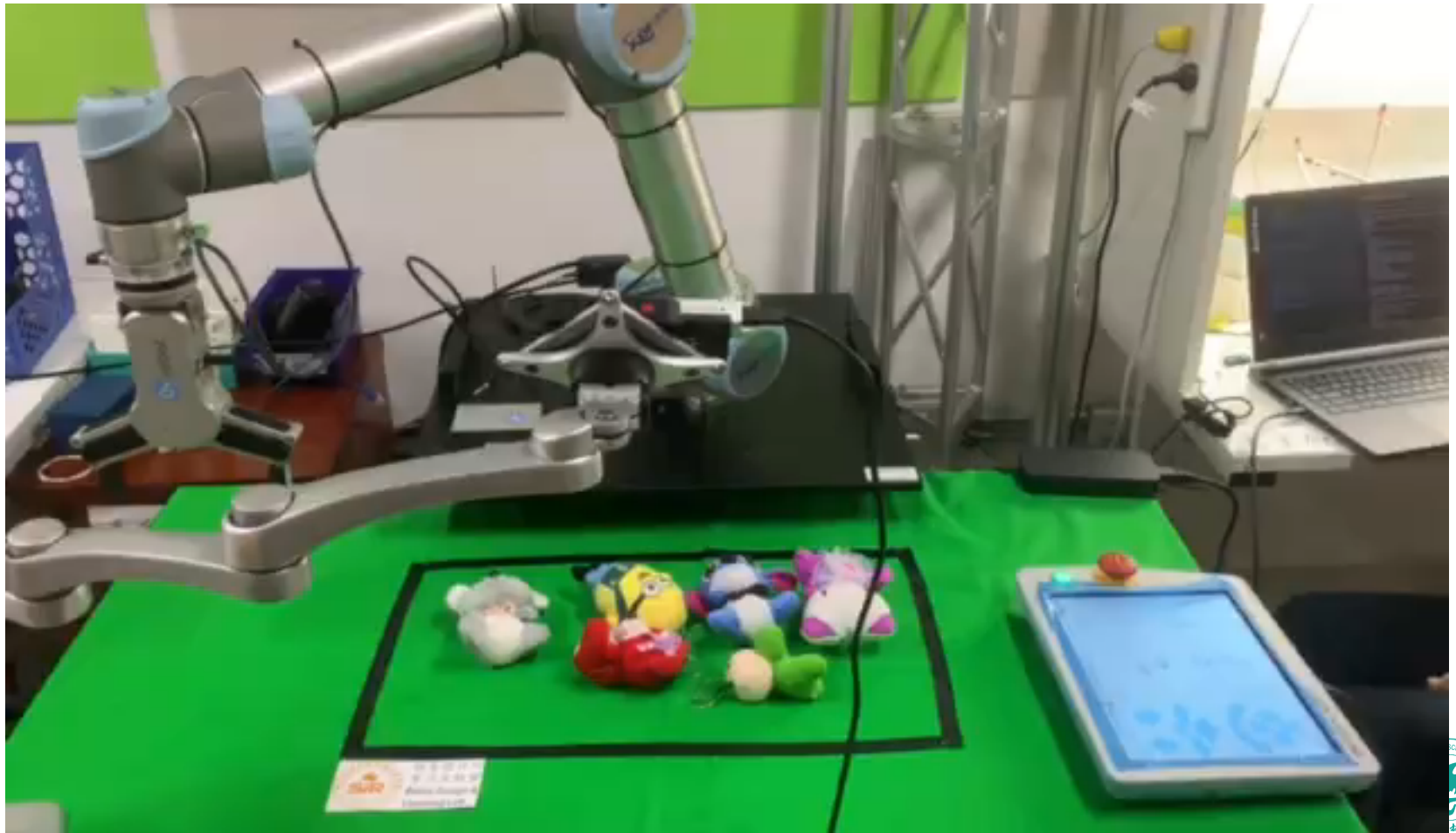


Hardware

- Robot System (RS)
  - Arm: UR5 from Universal Robot
  - FT Sensor: FT300 from Robotiq
  - Gripper: Adaptive 3 Finger from Robotiq
  - Camera 1: Kinect for X-box One from Microsoft
  - Camera 2: LifeCam from Microsoft
  - Camera 3: Canon EOS M3
- Learning Computer (LC)
  - OS: Ubuntu Trusty 14.04.5
  - CPU: Intel Core i7 6800K Hex-Core 3.4GHz
  - GPU: Single 12GB NVIDIA TITAN X (Pascal)
  - RAM: 32GB Corsair Dominator Platinum 3000MHz (2 X 16GB)
  - SSD: 1TB Samsung 850 Pro Series

# Project 4

<https://github.com/ancorasir/BionicDL-CobotLearning-Project4.git>



# Hardware

---

- Robot System
  - UR5/...
  - Realsense
  - OnRobot RG6 gripper/...
- Learning Computer
  - Training: 4 x NVIDIA 1080Ti
  - Inference: can be run on cpu with lower speed
  - Tensorflow-gpu 1.12 + cuda 9.0 + cudnn 7.5
- Toys suitable for the gripper

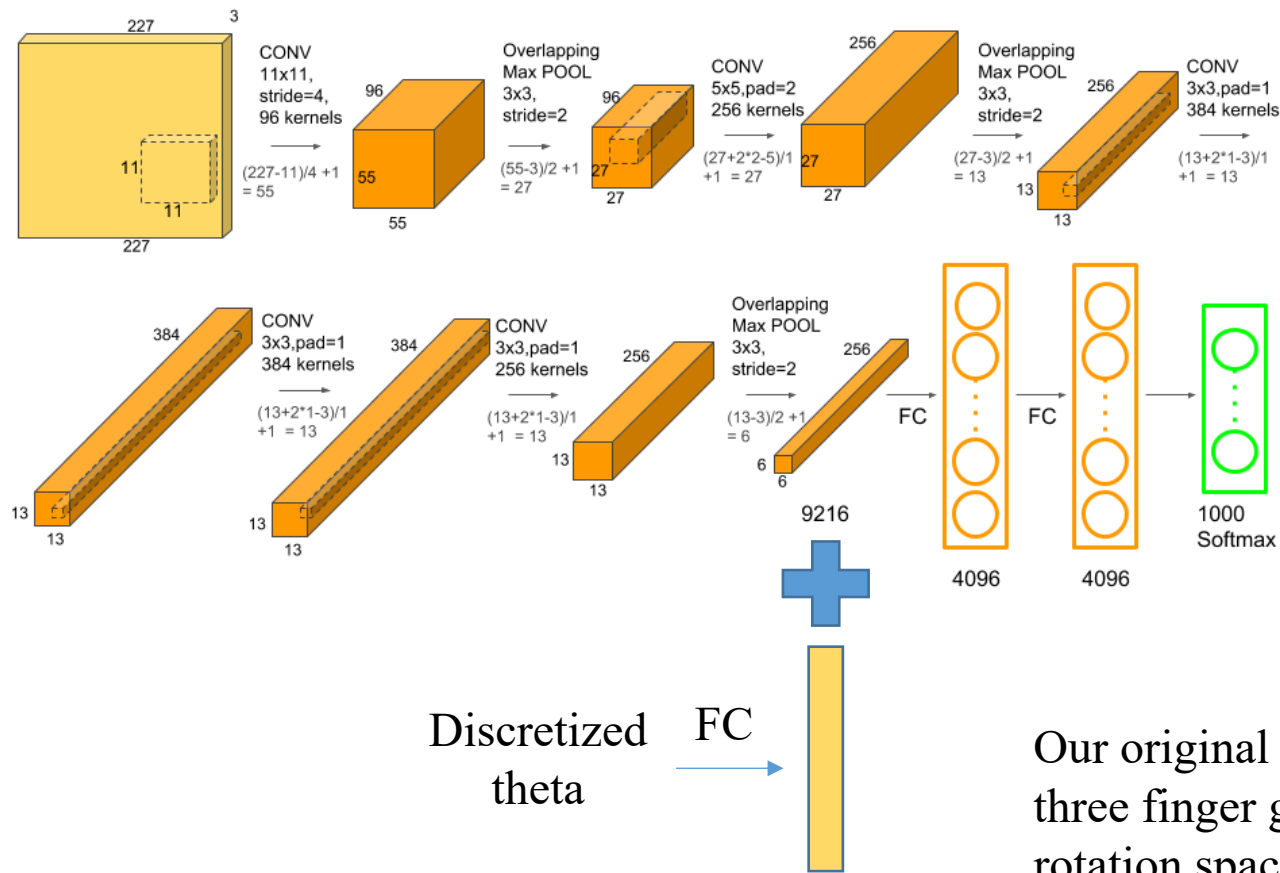
# The blind grasp dataset

- Grasp pose ( $x$ ,  $y$ ,  $\theta$ ) is randomly generated.
  - Image: for each grasp is cropped such that the grasp point is at the center of the cropped image.
  - Rotation angle of gripper is recorded



# GraspNet

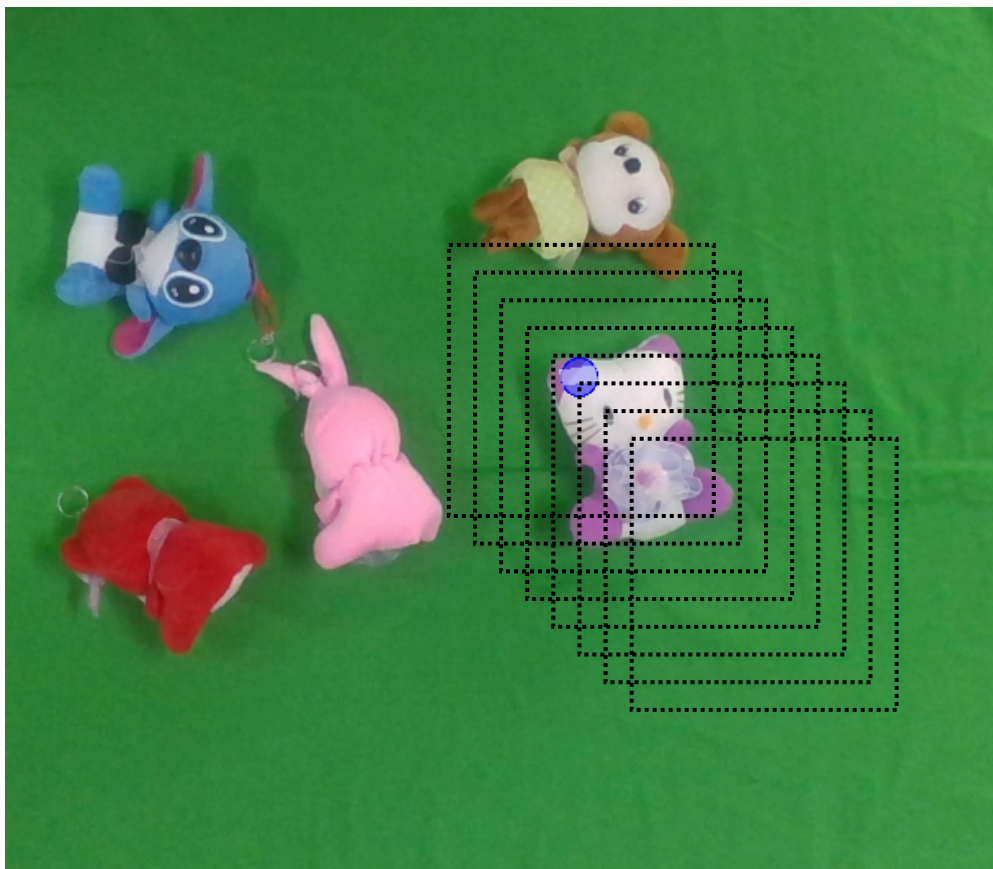
## Fine tuning modified Alexnet



Our original version used a three finger gripper. So the rotation space  $0 \sim 2\pi$  is divided into 18 bins.



# Predictor



- You may need to crop the original image from realsense if the view area is too big.
- Transform from  $u, v, \theta$  to robot pose:
  - $(u, v)$  to  $(x, y)$  though perspective transformation.
  - $\theta$  to yaw. The  $\theta$  predicted by the model is

negative



0



positive



# Thank you!

Prof. Song Chaoyang

- Dr. Wan Fang ([sophie.fwan@hotmail.com](mailto:sophie.fwan@hotmail.com))

