Robotics Lab - RoboL Vision Lab - CoViL

LearnBiP (2011-2012)



European Clearing House for Open Robotics Development

#### **Grasp Learning in Industrial Bin-Picking**

#### Main Goals:

- Learning in industrial bin-picking
- Evaluation of the use of the dexterous hands SDH-2 in an industrial production context

#### Additional Achievement

Use of simulation for replacing human modelling of grasps



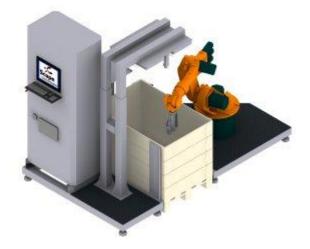




Robotics Lab - RoboL Vision Lab - CoViL

## Learning in industrial bin-picking

- During bin-picking a large amount of experience is generated
  - Cycle time of 25 seconds gives around 100000 experiences a month
- Currently the success rate varies between 50% and 90% depending on the objects and gripper
- Grasp definition requires a lot of manual design
- The large amount of experience is yet completely unused!
  - Aim: Improvement through learning





Robotics Lab - RoboL Vision Lab - CoViL

## **Achievement 1: Improving of** manually chosen grasp preferences



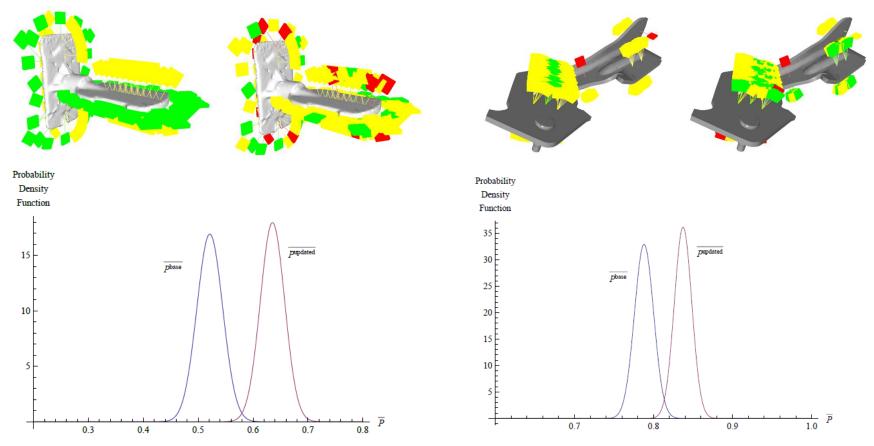


L.-P. Ellekilde, J. A. Jørgensen, D. Kraft, N. Krüger, J. Piater and H. G. Petersen. Applying a Learning Framework for Improving Success Rates in Industrial Bin Picking. IROS 2012.

The Maersk McKinney Moller Institute 10-07-2013

Robotics Lab - RoboL Vision Lab - CoViL

# Achievement 1: Results on improving of manually chosen grasp preferences



L.-P. Ellekilde, J. A. Jørgensen, D. Kraft, N. Krüger, J. Piater and H. G. Petersen. Applying a Learning Framework for Improving Success Rates in Industrial Bin Picking. IROS 2012.



#### Conclusions

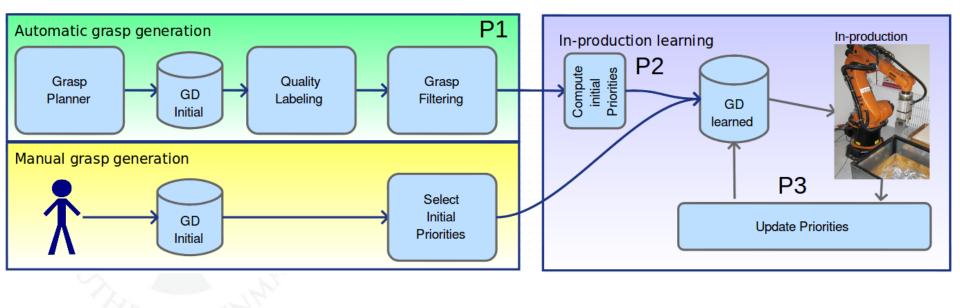
- Learning can improve bin-picking in an industrial context
  - Utilizing vast amount of available experience
  - Reduction of error rate by more than 20% in two set-ups



Robotics Lab - RoboL Vision Lab - CoViL

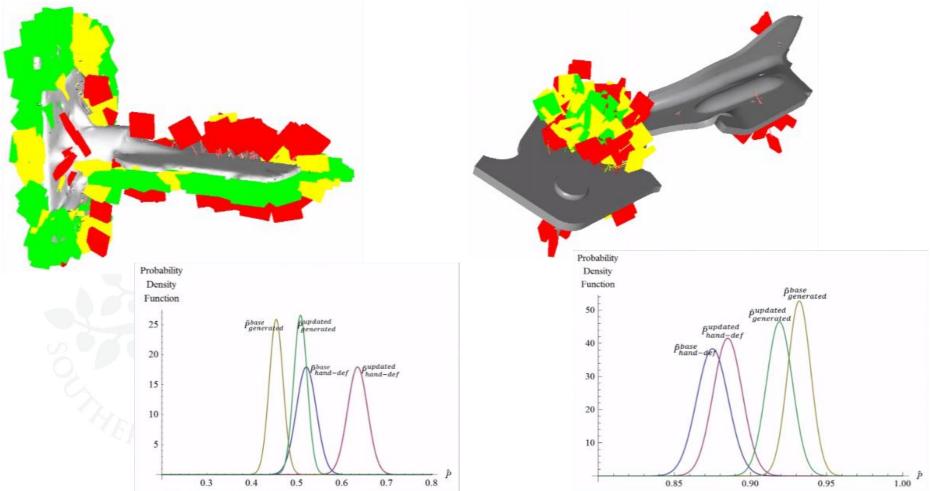
# Achievement 2: Replacing Manual design through simulation

- Problem: Potential grasps are designed manually
- Replace by process by simulation

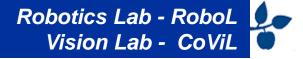


Robotics Lab - RoboL Vision Lab - CoViL

## Achievement 2: Replacing Manual design through simulation



10-07-2013



### Conclusions

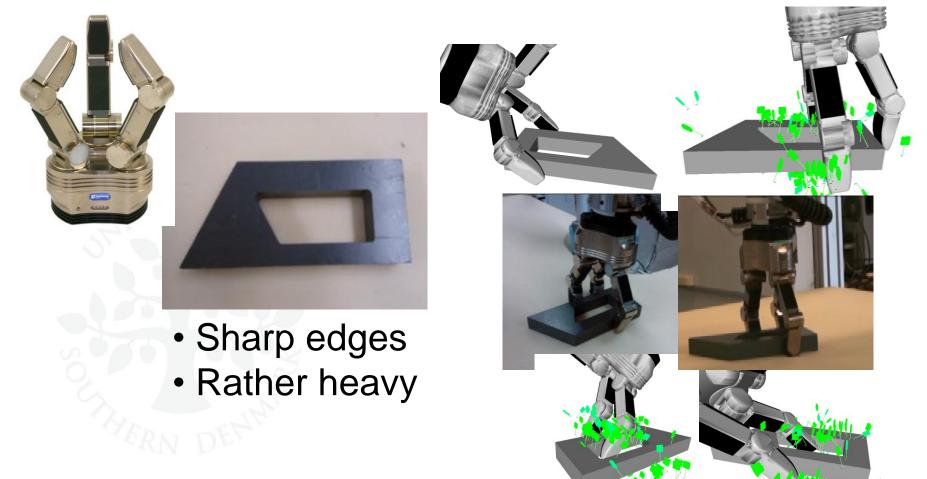
#### Learning can improve bin-picking in an industrial context

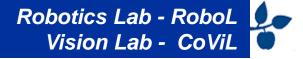
- Utilizing vast amount of available experience
- Reduction of error rate by more than 20% in two set-ups
- Dynamic simulation can substitute manual intervention in grasp definition while keeping similar performance



Robotics Lab - RoboL Vision Lab - CoViL

# Achievement 3: Show potential of use of dexterous hands in bin-picking



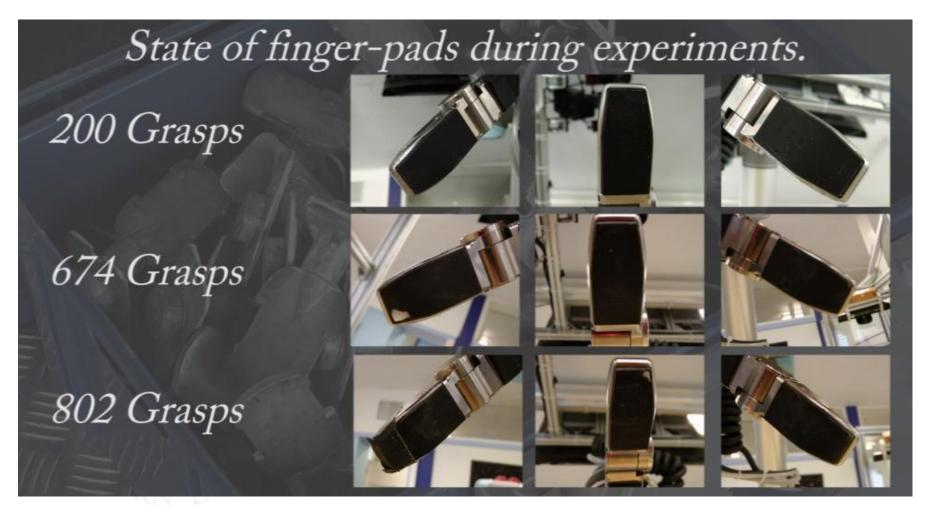


### Conclusions

- Learning can improve bin-picking in an industrial context
  - Utilizing vast amount of available experience
  - Reduction of error rate by more than 20% in two set-ups
- Dynamic simulation can substitute manual intervention in grasp definition while keeping similar performance
- Dexterous grippers have a large potential for industrial binpicking

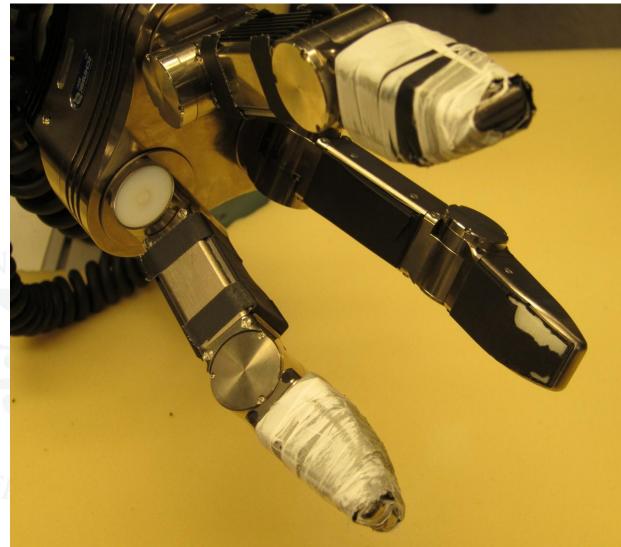
Robotics Lab - RoboL Vision Lab - CoViL

#### But!





#### And after more than 5000 grasps



Robotics Lab - RoboL Vision Lab - CoViL

## Conclusions



European Clearing House for Open Robotics Development

- Learning can improve bin-picking in an industrial context
  - Utilizing vast amount of available experience
  - Reduction of error rate by more than 20% in two set-ups
- Dynamic simulation can substitute manual intervention in grasp definition while keeping similar performance
- Dexterous grippers have a large potential for industrial binpicking
- But the SDH-2 Schunk hand is not yet ready for use in an industrial context

Robotics Lab - RoboL Vision Lab - CoViL

