

Newsletter

Date: 16th of February, 2017 Research Area: Robust and Flexible Automation (RA2)

This newsletter is published prior to each workshop of SFI Manufacturing. The aim is to keep the community up to date with the current research that is being carried out within and related to the SFI. This issue of the newsletter is focused on the research and achievements from the area Robust and Flexible Automation.

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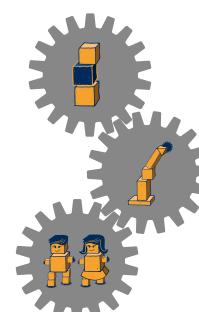
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Centre for Research-based Innovation

The Research Council of Norway

SFI Manufacturing

A cross-disciplinary centre for research based innovations for competitive high value manufacturing in Norway



About the research area

Recent developments within automation technologies increase the ability to rapidly adapt to changing conditions, and open new ways to use automation and robotics in manufacturing systems. Some examples of advances based on novel technology:

- Increased flexibility and robustness
- In-process monitoring and real-time control
- Faster and easier reconfiguration
- · Intuitive and adaptive manufacturing adapted to human needs
- Intuitive programming and tasking
- Trajectory planning with adaption to changing environments
- Sensor systems capable of mapping and analysing changing environments
- Multi-robot and human-robot coordination and cooperation
- Real time sensor based control

The overall objective of the research within the area Robust and Flexible Automation is to further develop and link novel technologies and methodologies within automation to support innovation processes and advanced work systems in the manufacturing industries.

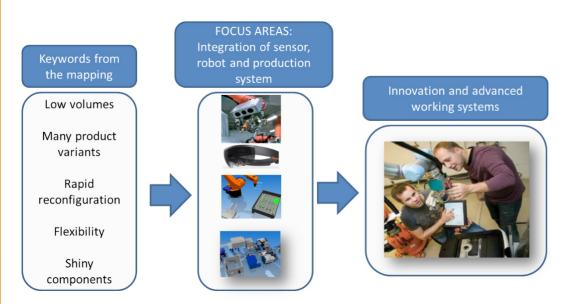


Figure 1. Visualization of the overall objective of the research area.

A number of research challenges will be addressed within the research area. The work is organised into two research and technology development work packages (WP's):

- WP2.1 Extended pick and place robotic handling of non-uniform objects
- WP2.2 Flexible and integrated production systems

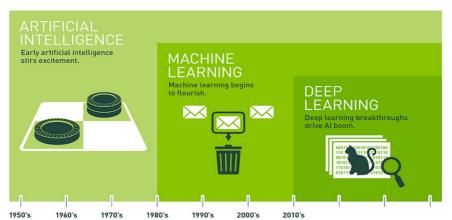
These WP's are also highly interlinked in the way that WP2.2 will focus on using and integrating the specific technologies and methods developed in WP2.1, as well as other technologies on a production system level.

In addition, activities within these WP's also link to the two other research areas within the SFI, i.e. RA1 and RA3. The 2017 work package activities have been chosen based on the mapping of industrial use cases and requirements, technology watch and the preliminary results from 2016.

From the virtual world into the real world

On the 14th of March 2017, the next SFI workshop will take place, with Rolls-Royce as the host organization. Research area Robust and Flexible Automation has the main responsibility for the scientific part of the workshop. In the following sections, we give a brief status report and introduction to the main topics that we plan to present in the workshop, i.e. Grasping using Deep Learning, 3D Point Clouds for object localization and Simulation for developing and virtual testing of production systems.

1) Deep Learning



Automatic grasping using Deep Learning

Pose estimation of objects is one of the key problems for the automatic-grasping task of robotics. Within the research area Robust and Flexible Automation, a long-term goal is to create a generic system for grasping all types of objects in cluttered scenes. As a step towards this goal, we have developed a system for robot grasping using deep learning.

Deep learning algorithms can learn features and tasks directly from images, and can automatically extract high-level, complex abstractions from images. Instead of learning a traditional machine vision system each new object it should handle, a deep learning algorithm can be trained for handling a large spectre of objects and objects in cluttered scenes. These strengths makes deep learning an important tool to achieve a more generic system for grasping.

What is Deep Learning and why using it?

Deep learning is based on a hierarchical learning architecture and is motivated by artificial intelligence emulating the deep, layered learning process of the primary sensorial areas of the human brain, which automatically extracts features and abstractions from the underlying data. Deep learning models have demonstrated strong power in learning hierarchical features, which greatly facilitates computer vision tasks like object detection and recognition.

Deep Learning for detecting robotic grasps

Traditionally, object localization and grasping have been solved by a fixed setup with an exact position of the object. Alternatively, vision based grasping can been applied. The need for a fixed setup is then removed. However, the objects have to be known to the vision system. In this work, we present a vision-based robotic grasping system, which not only can recognize different objects, but also estimate their poses by using a deep learning model, finally grasp them and move to a predefined destination. We apply a 3D convolutional neural network (3DCNN) to detect potential grasps and estimate their pose.

Figure 2. Development of Artificial intelligent methods.

We have built a database comprised of five types of objects with different poses and illuminations for experimental performance evaluation. The experimental results demonstrate that the vision-based robotic system can grasp objects successfully regardless of different poses, illuminations and scenes.

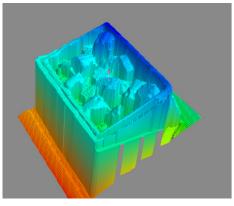


Figure 3. 3D image of a bin with steel parts. The red arrow shows the gripping point and angle.



Figure 4. By using a differential flow sensor it is possible to distinguish a successful from an unsuccessful grasp.

Using VR for training

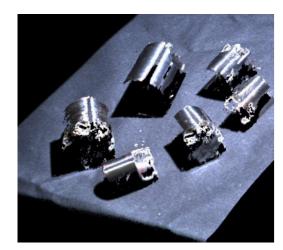
Deep learning models need huge amount of data in order to generate a good model. Last year, Google made an experiment on grasping by learning. Between six and 14 robots were at any time picking up objects, and over the course of two months the robots had picked objects 800 000 times. We have developed a simulation tool that produces simulated data to train our model. This will give an improved model, reduce the total number of grasps needed, and save time.

Further work

Even though the algorithm can learn how to grasp from simulated data, a physical experiment will be needed to learn from real world data. To improve the grasping model, we will automate the process of learning by using a robotic manipulator which will be left to continuously try to grasp objects based on the already developed algorithm. By sensing a successful grasp in the experimental setup, the algorithm will continue to learn how to make the best possible grasps.

2) 3D vision

Machine vision is the use and processing of images or point clouds by computers to execute tasks such as verification and localization of objects. In SFI Manufacturing we are focussing on 3D vision with point clouds. Point clouds represents the external surface of objects and can be described as listed point coordinates x, y and z, visualised with or without colours.

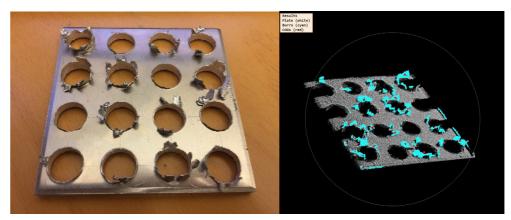


Recent developments in the sensor-field enable us to generate point clouds fast and with high accuracy, which makes it useful for industrial inspection and quality control. 3D vision, compared to 2D vision, is robust to variation in light conditions, simplify object localization and makes inspection of more unconstrained scenes possible. 3D vision in SFI Manufacturing

SFI manufacturing has two approaches to 3D vision. We work on developing new algorithms to locate objects in point clouds, using Deep Learning, and we use these algorithms as well as existing methods in available industrial vision systems to solve generic industrial issues, like bin-picking

In 2016 and the beginning of 2017, we have been evaluating the open source Point Cloud Library (PCL) and the commercial systems Scorpion and Halcon, in order to compare them with the methods developed in SFI Manufacturing, as well as use them directly for object localization and inline inspection in industrial applications.

The innovation project NAP¹, with help from SFI Manufacturing, has successfully demonstrated the use of point cloud in combination with laser triangulation to localize large parts with highly reflective material, see figure 6 and 7.



Further work

In 2017, the plan is to get a better understanding of state-of-the-art methods, and use this knowledge as input and reference for further development of new algorithms, as well as apply it to new industrial applications. We will especially focus on reducing the time necessary to setup an object localization system based on 3D vision.

3) Simulation

Simulation in manufacturing control consists of developing a model and a virtual scene that will form the basis for developing and virtual testing of control systems. Simulation has the potential to reduce development time for new processes drastically. Many commercial products are available and used in the industry, but the potential remains huge.

Simulation in SFI Manufacturing

Continuing on the work of SFI Norman, SFI Manufacturing has been working on verification of control applications using simulation models. In SFI Norman, a demonstration was made using 3D Create from Visual Components. The integration was done using a software stack with a proprietary protocol. The raising of the OPC UA protocol as the main Industry 4.0 protocol offers new possibilities, by reducing the work needed to connect the control stack to the simulation system and the real shop floor.

The OPC UA protocol itself is a standard available into many products such a PLCs, Vision systems and MES systems, but the realization of its potential requires a standardisation of its profiles. SFI Manufacturing, together with the project NAP, has been focusing on increasing knowledge on OPC UA and improving its software ecosystem and device profiles.

Figure 5. Colour point cloud of shiny metal tool parts from Teeness.

Figure 6 and 7. Detection of burrs using 3D sensor.

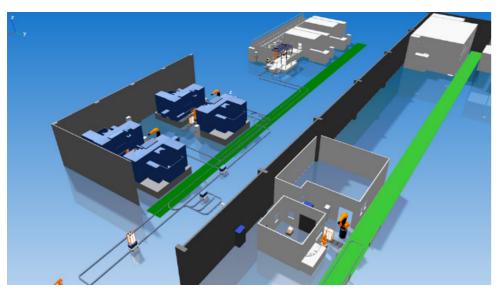


Figure 8. Shop floor simulation using Visual Components 3D Create.

Together with the Innovation project AutoFlex², SFI Manufacturing uses a method of near simultaneous execution of a physical environment and a simulated environment. To visualize the physical and simulated environment, augmented reality (AR) is used. Figure 9 and 10 show the environment before and after AR is used for visualization.

A part of the result from the innovation project Autoflex, is a method for effective and intuitive programming of robots and robot assembly applications, through restructuring and adding required data to existing CAD-models. The structure of these CAD-models has been developed further in SFI Manufacturing, to simplify extraction of required data. The CAD-models is input to a simulated environment, where changes can be simulated and verified to be automation friendly.

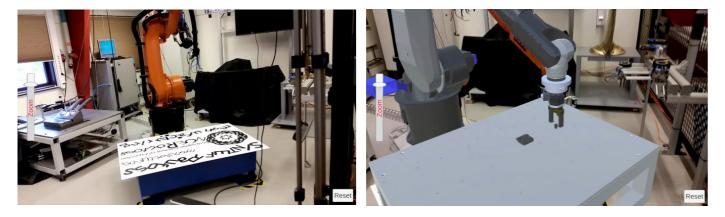


Figure 9 and 10. The environment before and after AR is used for visualization.

Further work

We will continue with working on visualisation to decrease development time for new manufacturing processes and modification of existing processes. Furthermore, we will be working on using and pushing interoperability standards and develop new methods for effective and intuitive programming.

New projects and project proposals

In the last grant from the Research Counsil's programme for User-driven Research based Innovation (BIA), a total of 42 innovation projects and 5 researcher-driven projects were invited to negotiations. The projects described below are all relevant for the research topics within the area Robust and Flexible Automation, because they partly address the same research topics, and they have one or several industrial and/or research partners from the SFI. In addition, we have also added a brief description of an EU-proposal that has been submitted, with SFI partners and relevance.

Innovation project DAMP – Fast Development of new Automated Manufacturing Processes through digital integration and testing

Rolls-Royce



Main objective: To achieve a digitally integrated manufacturing development process capable of delivering proven new and automated production processes in the shortest possible time, at a minimum cost, and as automated as possible.

Expected results: To greatly reduce the time and cost involved in production process development, starting with a digital product model, and ending with a verified and proven automated production process. The outcome will include automatic programming of virtual production processes from digital product models: automatic programming of production processes on manufacturing equipment, and automatic data capture and analysis for physical product and production process verification.

Innovation project FREM – Fleksibel Robotisert Elektromekanisk Montasje





MEKTRON

Main objective: To enable automatic assembly of electromechanical products in small batches by breaking research in human-robot cooperation and other key technological areas.

Expected results:

- Focus on flexible solutions for secure and efficient human-robot cooperation
- 3D machine vision for automatic localization of components from CAD models
- Robotic force based insertion at high precision component assembly
- Communication and data models for information flow between production and design processes •
 - Methods for quick and efficient programming of a large number of assembly operations
 - Generic usable rules for design and development that takes into account the latest assembly technologies and thus facilitates positioning and assembly with cooperating robots

SINTEF Raufoss

SINTEF Raufoss

Innovation project SmartChain – Digitaliserte, automatiserte og integrerte verdikjeder



Main objective: To develop methods and solutions that ensure efficient supply chains with a high degree of technology supported production, management and control.

Expected results:

- Strategic concept for efficient supply chains with high degree of automation and digitalization
- Method for automated production of volume products in conjunction with the production of premium products and technical prototypes in a laboratory environment
- Methodology for integrated development of production-friendly products
- Integrated system for real-time monitoring and control, which effectively manages material flow and unexpected events in the supply chain

Knowledge-building project for Industry CPS – The Cyber Physical System Plant Perspective



Main objective: To develop, utilize, implement and evaluate enabling technologies for Norwegian industries of the future in a combined physical and virtual industrial model built on the principles of Industry 4.0 including digitalization and interconnectivity.

Expected results:

- Framework for Norwegian approach for digital manufacturing industry
- Roadmap of new digital (Industry 4.0) technologies
- Demonstration and evaluation of new technologies for CPS where decision support through simulation capacities improves overall plant efficiency
- Methods on how advanced automated and human potentials and solutions in the plant manufacturing system can be increased, based on a new digital plant platform
- Use of cloud resources providing functionalities and computational capacities

Submitted EU-proposal Optimus – OPTimized In-line Measurement and control for manUfacturing Systems



Main objective: To develop, implement and demonstrate a system-level architecture for metrology and control to improve the first pass yield of micron accurate products from 80-90% to 99%.

Expected results: The development and demonstration of three non-destructive inspection methods, a holistic process control, and system level architectures, which will be implemented and demonstrated in three pilot production lines.

PhD and postdoc progress reports

PhD's and postdocs are essential resources within the SFI. Linn, Signe and Mathias are connected to the research area Robust and Flexible Automation.

Linn Danielsen Evjemo

My name is Linn Danielsen Evjemo and I started my PhD on the 1th of December, 2016. I have my master's degree from the department of engineering cybernetics at NTNU. In my master work I attempted to perform real-time telemanipulation of a humanoid robot using hand-held motion trackers. In my PhD I will focus on large-scale, robotized additive manufacturing using industrial robots and cold metal transfer welding. I will try to see if it is possible to combine the large workspace of an industrial robot arm with the flexibility and relative affordability of traditional additive manufacturing methods.

Signe Moe

My name is Signe Moe and I started my postdoc on the 1th of January, 2017. I have a master and PhD from the department of engineering cybernetics at NTNU. In my master, I worked on path following of marine vehicles in the presence of unknown ocean currents. I finished my PhD in November 2016 on the topic of guidance and control of marine vehicles and set-based control of robotic systems. I am now planning on extending set-based theory to industrial needs. I will partly collaborate with Linn Danielsen Evjemo on 3D-printing using robots, and will also focus on applications such as collision avoidance and orientation control.

Mathias Hauan Aarbo

I am Mathias and I started my PhD in the fall of 2015. My PhD focuses on robotic assembly and sensor fusion. I come from the department of engineering cybernetics and work mainly with sensor fusion and robotics. My master thesis was on sensor fusion of delayed displacement measurements. The Bayesian formulation of how to handle that delay was my main topic. In my PhD I look at assembly with articulated robots under uncertainty. With enclosed robots we can know exactly where everything is placed and can make plans that will work in these environments. We as humans operate in an environment of uncertainty, objects are larger or smaller than we expected, but we compensate. And that is what we are trying to make the robots do in a systematic fashion.

The probabilistic approach to uncertainties sees things as likelihood. It is likely that the piece is there, and based on how likely it is, the robot can form a better understanding of the object. Of particular interest is the screw-in process, where a small uncertainty can cause large issues. Estimation of the angle of entry, the location of the hole, and the relative orientation of the screw combined with robust control strategies will allow users to do faster prototyping of robot tasks and simplify automation.









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