# **Trends of Manufacturing Automation in Romania**

Theodor Borangiu, University Professor, PhD National Centre of Research and Training in Robotics, Industrial Informatics and Material Engineering University Politehnica of Bucharest, Dept. of Automation and Applied Informatics E-mail: borangiu@cimr.pub.ro



# Summary

- 1. Generating CNC toolpaths from grey level image processing
- 2. Embedded numerical adaptive control of machine tools
- 3. Robots for construction: from CAD to Real Time Control
- 4. Robot integration in manufacturing: Merged GVR AVI tasks

(Guidance Vision of Robots – Automated Visual Inspection)

5. New paradigms in manufacturing control: the holonic approach



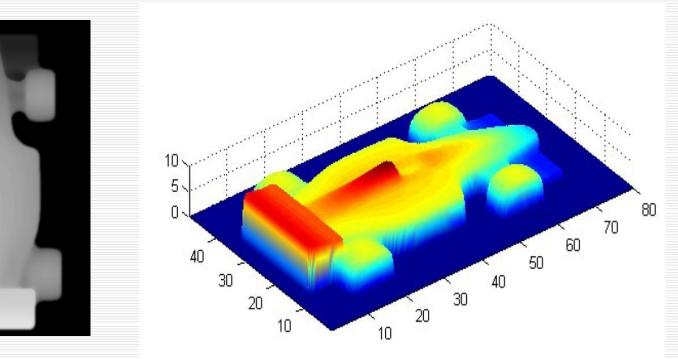
**1. Generating CNC toolpaths from grey level image processing** 

**Problem**: Automatic generation of 3D machining surfaces with tool compensation from grey level image models

- Height Map images
- □ Modelling the machining surface and tool shape
- Performing tool compensation
- Generating roughing toolpaths
- Generating finishing toolpaths
- Error analysis
- **G** Future developments

# 1. Generating CNC toolpaths from grey level image processing

#### Height Map images



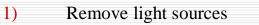
# 1. Generating CNC toolpaths from grey level image processing

### **Obtaining height map images**



#### 3D Model in POV-Ray



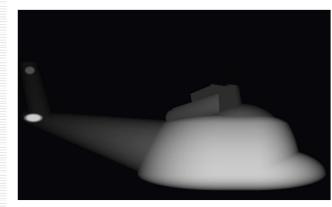


- Remove material textures 2)
- 3) Use an ortographic camera
- Apply a pigment with luminance proportional to the distance from the camera plane: 4)

□ farthest point:

pure black (luminance 0)

**closest point:** pure white (luminance 1)



Height Map Model

# 1. Generating CNC toolpaths from grey level image processing

### **Obtaining height map images**



# The 3D surface of a model to be machined is scanned with a laser range finder device:

- A vertical stripe of laser light is moved across the model object surface, and captured by a video camera. Along each horizontal scan line of the video frame, the *brightest* spot is taken to be the point at which the laser stripe "hits" the surface (detection at sub-pixel resolution).
- The relative positions of the laser and the video camera are used to find the 3D coordinates of the brightest spot by triangulation.
- The x-coordinate of each point in the output depth image is determined by the position of the laser stripe for a particular video frame
- The *y*-coordinate corresponds to a raster line in the video frame,
- The depth value is computed from the brightness peak detected along the raster line in the video frame

1. Generating CNC toolpaths from grey level image processing

#### **2.5D Surface Modelling**

- $\Box$  Pixel graylevel at (i,j) encodes surface height at (x,y)
- □ Pixel-to-millimeter ratio:
  - $\Box x = R i$
  - $\Box$  y = R j

□ Minimum Z of the surface: black pixel

- □ Maximum Z of the surface: white pixel
- Graylevel value:
  - □ 8 bit integer: low precision, low storage space
  - □ 16 bit integer: good precision
  - □ Floating point: best precision, high storage space

# 1. Generating CNC toolpaths from grey level image processing

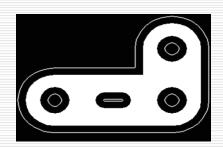
# **2.5D Surface Modelling**

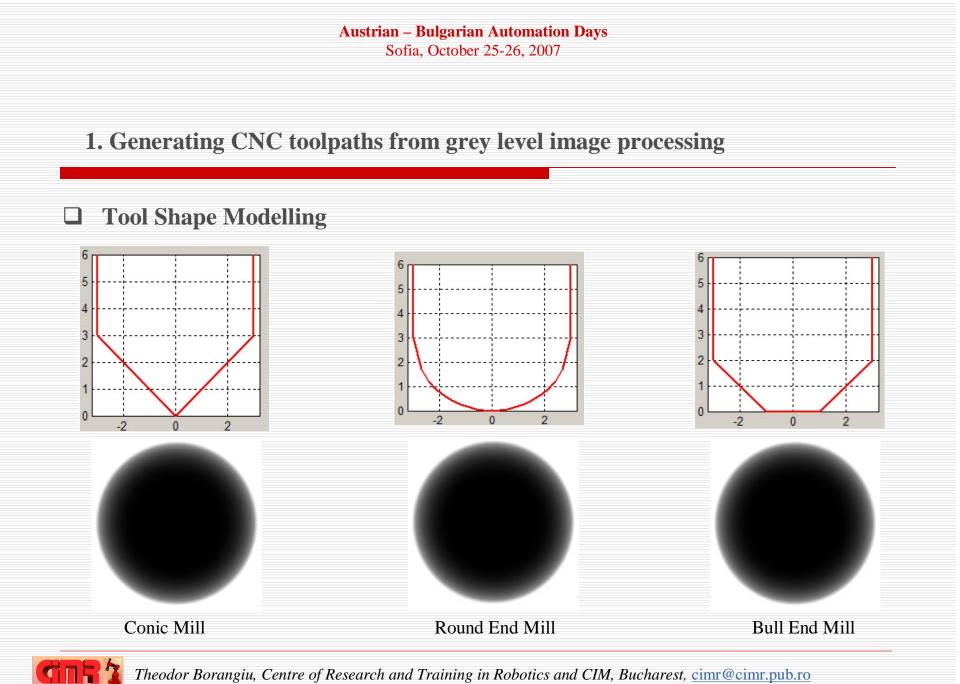
- **Simplest Case: 2 Dimensions**
- □ Offsetting is equivalent to image dilation
- For efficiency, only contour pixels need to be considered
- Tool Path is generated by extracting the contour
- By image erosion, we obtain the machined shape

Part model:

• Tool model:

Tool compensation:



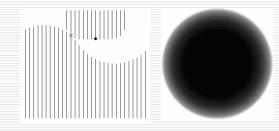


# 1. Generating CNC toolpaths from grey level image processing

### **Tool Compensation**

Objective: Generating gouge-free tool paths
 Idea: For each (x,y) position, find the maximum depth at which the end mill can go down without cutting extra material

Algorithm:Graysale image dilationImage:Surface modelStructural element:Tool model



The principle of discrete cutter compensation based on gradient computing in the 2D grey scale cutter shape image:

**Idea**: at every location in the XY plane (i.e. any image pixel) the depth is computed which should be reached by the milling cutter, in order to be tangent to the surface.

The shape of the milling cutter was modeled as a grey scale image, using the same scale factors as for the surface to be milled.



# 1. Generating CNC toolpaths from grey level image processing

### **Tool Compensation Results**

#### **Advantages**

- Gouge-free tool paths for many tool shapes
- □ Immediate generation of basic roughing and finishing tool paths
- □ Simple implementation, no need for complex 3D geometry computations

### **Disadvantages**

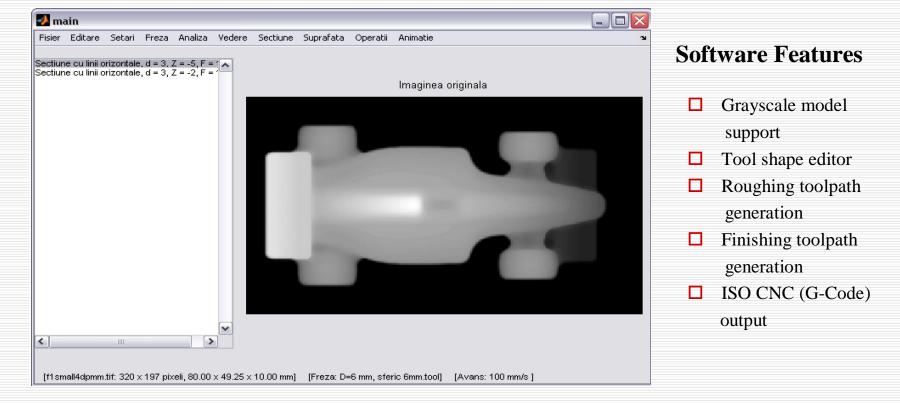
- □ High computation time
- □ High storage space for the image model
- □ Compromise between precision and speed!

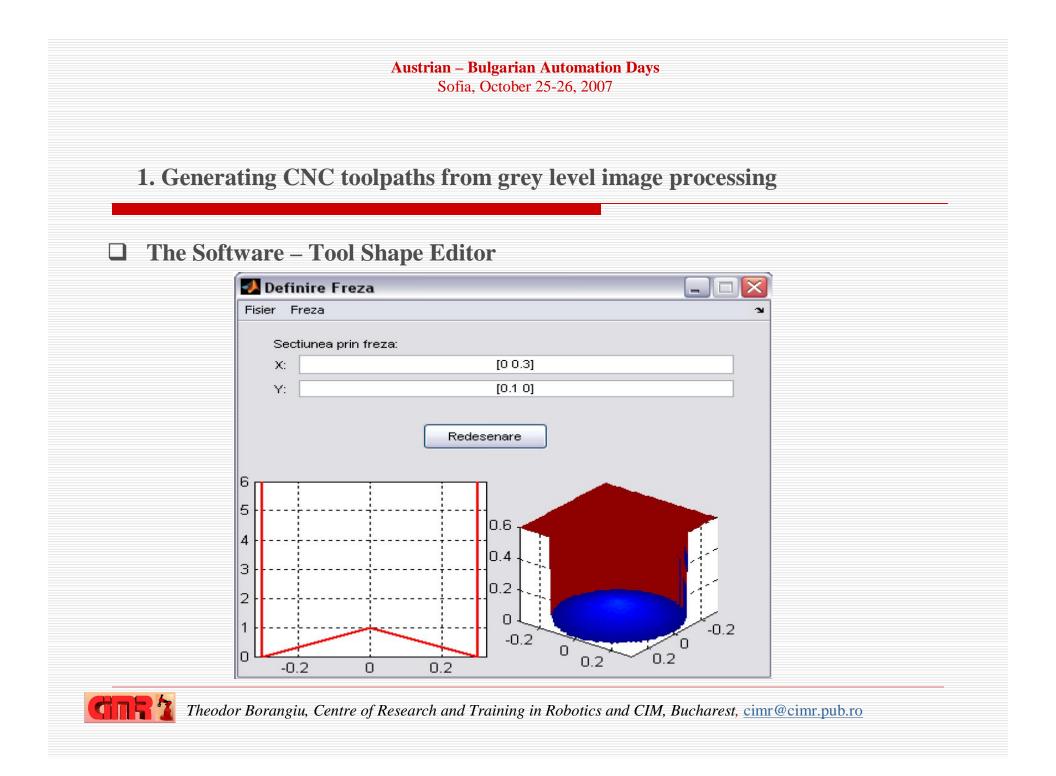
### **Increasing Speed**

- □ It is not always necessary to compute the whole surface
- □ The algorithm can be parallelized

### 1. Generating CNC toolpaths from grey level image processing

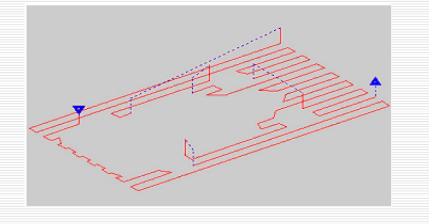
### **The Software – GUI Design**

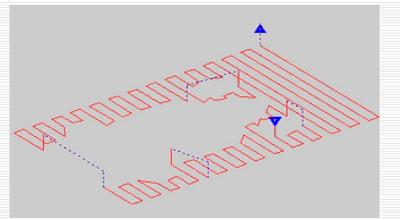




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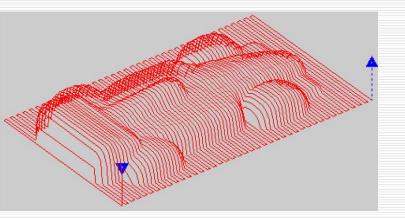
- Generating Roughing Toolpaths
- Each roughing stage is performed at constant Z level
- At a given Z level, selecting the region where the cutter should clean up is an image thresholding operation
- □ For flat endmill cutters we use 2D offset compensation

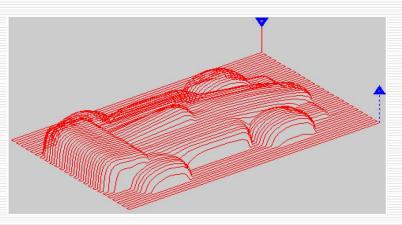




# 1. Generating CNC toolpaths from grey level image processing

- Generating Finishing Toolpaths 1<sup>st</sup> meth
- □ In XY plane, the tool moves parallel with one axis or direction
- □ The tool moves on the "safe surface"
- □ There is no need to compute the whole "safe surface"





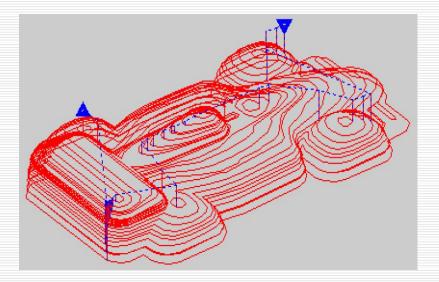
Theodor Borangi

1. Generating CNC toolpaths from grey level image processing

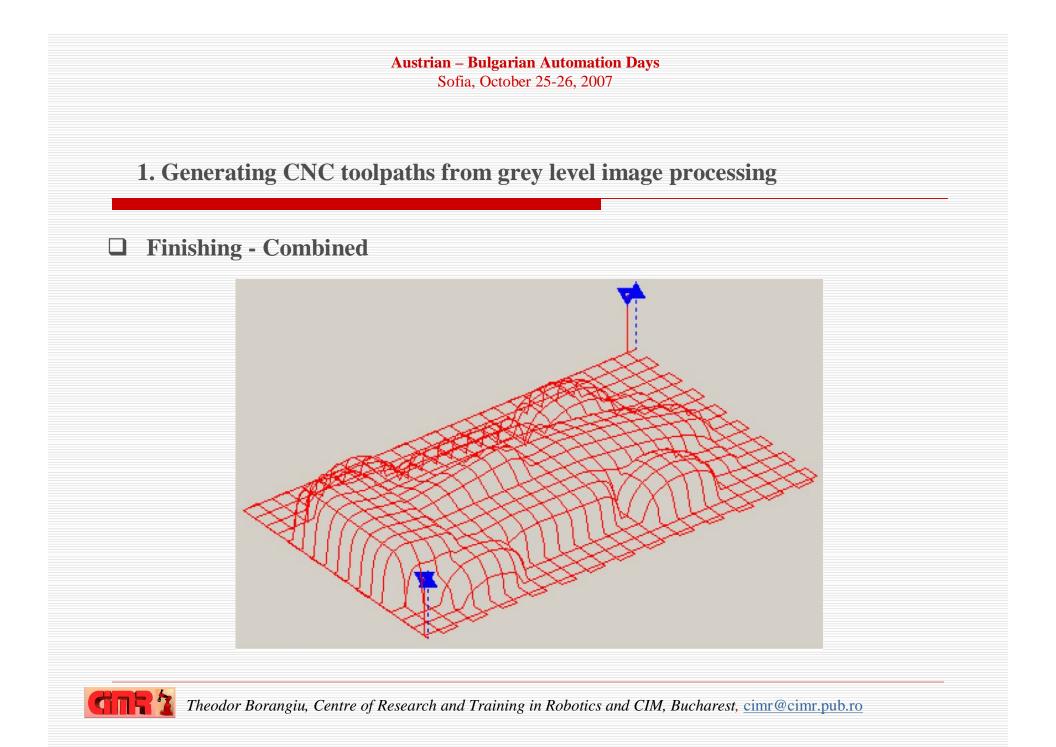
**Generating Finishing Toolpaths – 2<sup>nd</sup> method** 

□ Tool paths are at constant Z levels

- Because of the tool shape, we cannot use 2D compensation any more
- □ The whole surface needs to be computed!



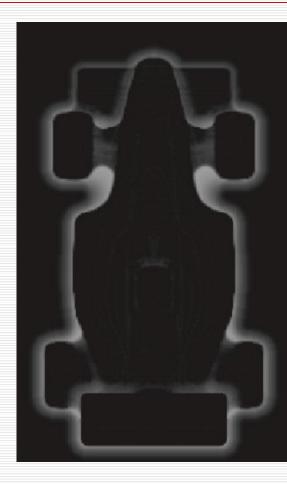




# 1. Generating CNC toolpaths from grey level image processing

**Error Analysis 1** 

- A tool can be too big to machine the fine details
- At first, we can use a bigger tool to machine surfaces without details, and then a smaller tool to machine only the small details
- We can simulate one cutting operation and see what could not be machined



1. Generating CNC toolpaths from grey level image processing

**Error Analysis 2** 

- Model the toolpath like a grayscale image
- □ Erode the image using the tool model
- Compare the eroded image with the original model





# 1. Generating CNC toolpaths from grey level image processing

□ ISO CNC Output	M03
	G0 X80 Y7.75
Toolpaths are made of linear segments and circular arcs	G1 Z-40 F100
	G1 X65.5
	G0 Z0
	G0 X71.75 Y10.75
	G1 Z-40 F100
<ul> <li>Succesive segments may be approximated with circular arcs</li> <li>Toolpath optimization: reduce the time for moving the head without cutting</li> </ul>	G1 X80
	G1 Y13.75
	G1 X73
	G1 Y16.75
	G1 X80
	G1 Y19.75
	G1 X73.25
	G0 Z0
	G0 X0 Y0
	M05

Theodor Borangiu, Centre of Research and Training in Robotics and CIM, Bucharest, cimr@cimr.pub.ro

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# **1. Generating CNC toolpaths from grey level image processing**

### **Gample Workpiece**



1. Generating CNC toolpaths from grey level image processing

**Future Developments – Collision Detection** 

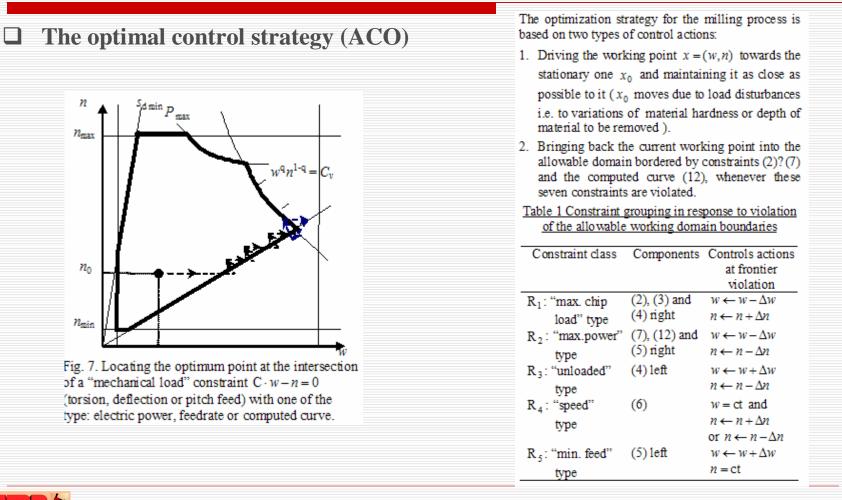
- □ The whole tool shape can be modelled, including the tool holder, to check if a tool path will cause a collision with the workpiece
- □ At every moment the amount of material can be computed; this is useful to check if the maximum allowed cutting depth of a tool is not exceeded.
- Vary feedrate and speed at roughing: save time, increase throughput

#### 2. Embedded numerical-adaptive control of machine tools

#### **1** The inequality-constrained Optimal Control of machining (ACO)

- **Quality Function**: the *cutting productivity*, for each one of the 2D closed *XY* roughing paths approximating one locus of spatial points of same grey level (points of depth  $Z_i, Z_i = Z_{i-1} \Delta_j$  in the part model image). The first paths to be machined = loci of nearest points relative to the range sensor (grey level value  $Z_0$ ), with cutting depth  $\Delta_0$ . If the furthest image points relative to the sensor have a grey level  $Z_j$ , then, by piecewise estimating the grey level gradient in the point depths range  $Z_0...Z_f$ , the number of *C* cutting passes is computed to plan the 21/2 approximation of 3D machining with cutting depths  $\Delta_j$ , such that  $\sum_j c_j \cdot \Delta_j = Z_f Z_0$ , and  $\sum_j c_j = C$ , and  $c_j$  roughing paths  $p_i$  have the same cutting depth  $\Delta_j$ .
- Computation of **machining passes** considers: *form accuracy* (piecewise change of the cutting depth  $\Delta_j$  on grey level gradient basis) and *material characteristics* (type and hardness of material to be removed impose upper limits on  $\Delta_j$ ). **Result**: each roughing path  $p_i$  should be realized at constant cutting depth  $\Delta_j$ , feedrate  $w_j = F$  and spindle speed  $n_j = S$  to protect the tool and the machine's kinematical chain.
- ACO approach: maintains  $\Delta_j$  constant at roughing path level, but *varies* in real time *w*, *n* along trajectory  $p_i$  computed from image points of same grey level optimize a cost function.

### 2. Embedded numerical-adaptive control of machine tools



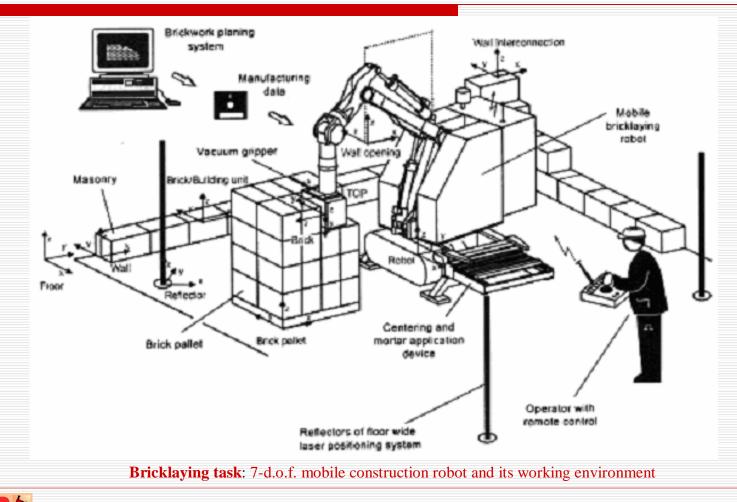
# 3. Robots for Construction – from CAD to RT Control

### Control trends in the building industry. R&D objectives:

- **Requirements in construction industry**: quality, standardization and cost reduction: *replication of construction tasks* (bricklaying, windows placing, applying mortar and finishing operations painting, polishing) and *increasing work productivity*.
- **Rationalization efforts** in the construction industry: attempt to *create information systems* progressively automating the building processes e.g. bricklaying
- New programming and control solutions for robot systems for the building industry: implementing *CAD/CAM solutions* generating, managing and using at run time in robot programs production data extracted from civil engineering projects.
- **Research Objectives**: Design and implement an open and modular control solution for construction robots based on embedded systems integrated in a Service Oriented Architecture: (i) *AGV functionality* to access locations in the building site; (ii) *Arm dexterity* for construction tasks; (iii) *Open motion control HW & SW in an embedded architecture*; (iv) *KBTS* mapping technology data (materials, operations) to programs.

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# 3. Robots for Construction – from CAD to RT Control



# 3. Robots for Construction – from CAD to RT Control

# **Functional architecture:**

•The mechanical device: a 5-d.o.f. *cylindrical arm*, partially closed-chain kinematic structure carried by a *mobile wheeled platform*.

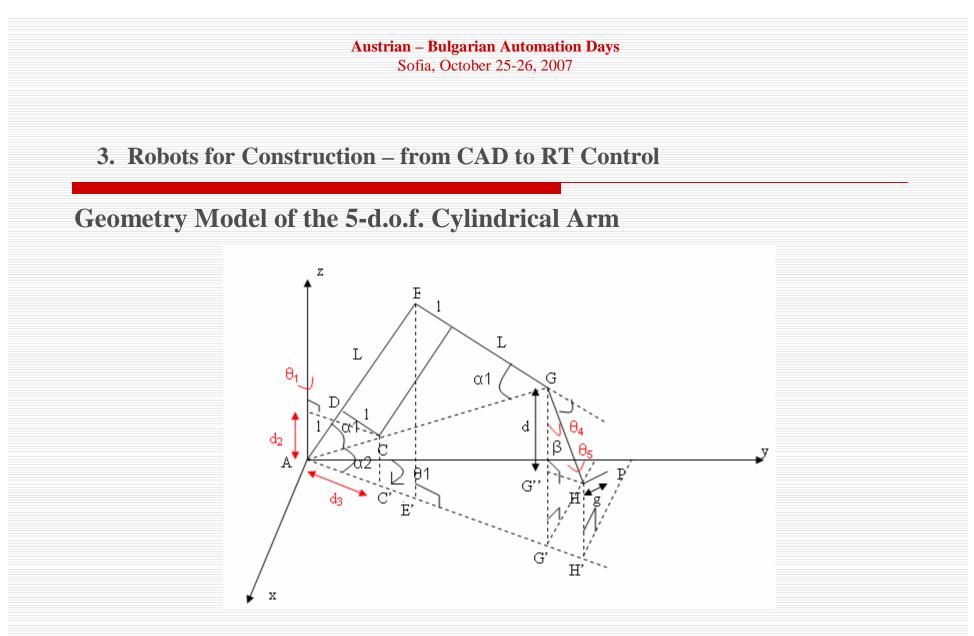
•The resulting 7-d.o.f. mobile construction robot is able to move on *quasihorizontal prepared surfaces* (floors), and generate a *workspace of 3.5 m height*.

•The mobile robot is a *free-ranging* (non-guided) *wheeled vehicle*, capable to avoid obstacles (e.g. brick pallets) in a structured environment; its arm performs coordinated movements either in the *Cartesian space* or in the *5-dimension joint space* automatically at program execution or under manual control.

•A computer-based operator console (wireless laptop) is used both as:

teach pendant for robot point learning and

robot terminal for execution of monitor commands, program editing, debugging, and execution start-up and monitoring.



The closed-chain kinematic structure of the 5-d.o.f. cylindrical robot manipulator

3. Robots for Construction – from CAD to RT Control

# **Generating Production Data for the Bricklaying Robot**

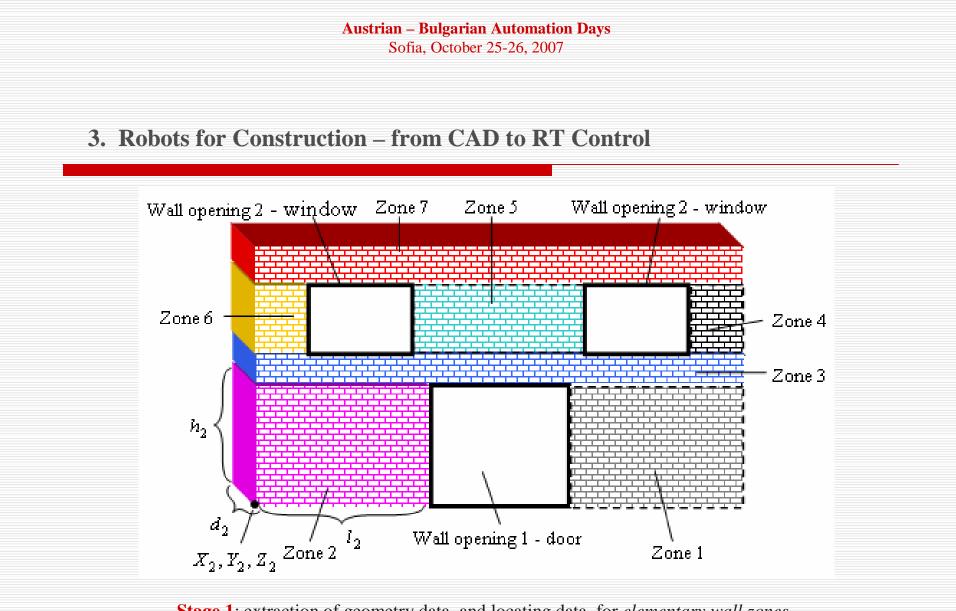
The process of *generating production data* for the bricklaying robot is done in 3 stages:

1. Extraction of geometry data from the architecture project and construction specifications (either from AUTOCAD files or manually input). 3D locations and dimensions of the walls, h-stockades (bulwark), wall openings, a.o, represent the input data. The output data computed in this stage refers to:

(i) the *dimensions* and coordinates of each *elementary masonry zone* relative to a unique world frame, and

(ii) the specification of materials necessary for the construction of elementary zones.

**2. Partition of the global masonry in wall segments** (relatively to a single floor of the building); this consists in joining several *elementary masonry zones*, , which may be completely included within the dexterous space of the cylindrical robot arm, with respect to a robotic bricklaying task associated to a *wall segment k*.



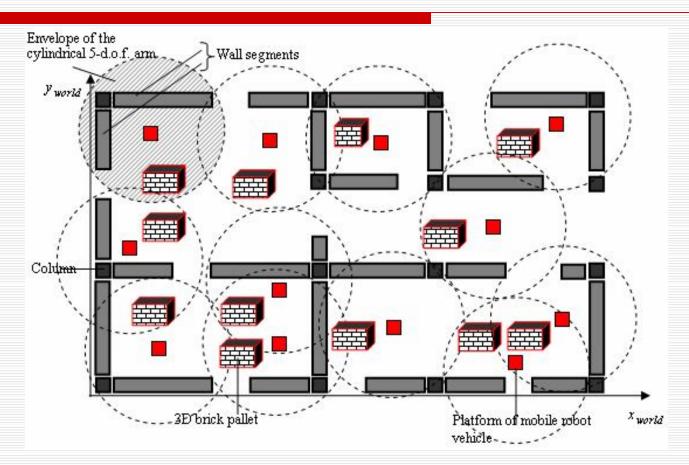
<u>Stage 1</u>: extraction of geometry data and locating data for *elementary wall zones* (from architecture plans and construction specs.)

# **3.** Robots for Construction – from CAD to RT Control

- **3**. **Determining the material requirements** number and types of bricks, the specification of *brick pallets* location, size, form, organization, and the *data for application programs* robot points, parameters of ,,pick-and-place" routines, execution order for wall segments, *planning of intermediate mobile robot locations* for navigating to successive wall segment locations or exiting the working area.
- The **typical robot program sequence** allowing execution of the task associated to a current wall segment eventually provides access to more than one brick pallets.
- I. *Manual/automatic guidance of the robot vehicle* to a world location from which access is granted both to the 3D brick pallet and to the ensemble of wall segments included in one elementary masonry zone.
- II. *Auto locating of the mobile robot platform* using the laser scanner and range finder device mounted on the robot platform; the sensor scans over 360 degrees in a planar movement and, from the distances measured to three fixed reflectors (of known locations) determines its own location *X*, *Y* in the world frame.
- III. Performing the bricklaying task ("pick and place" loop, 3D stack access, ...)

**GIR** 

# 3. Robots for Construction – from CAD to RT Control

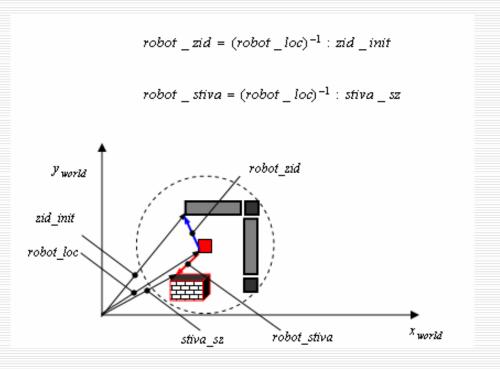


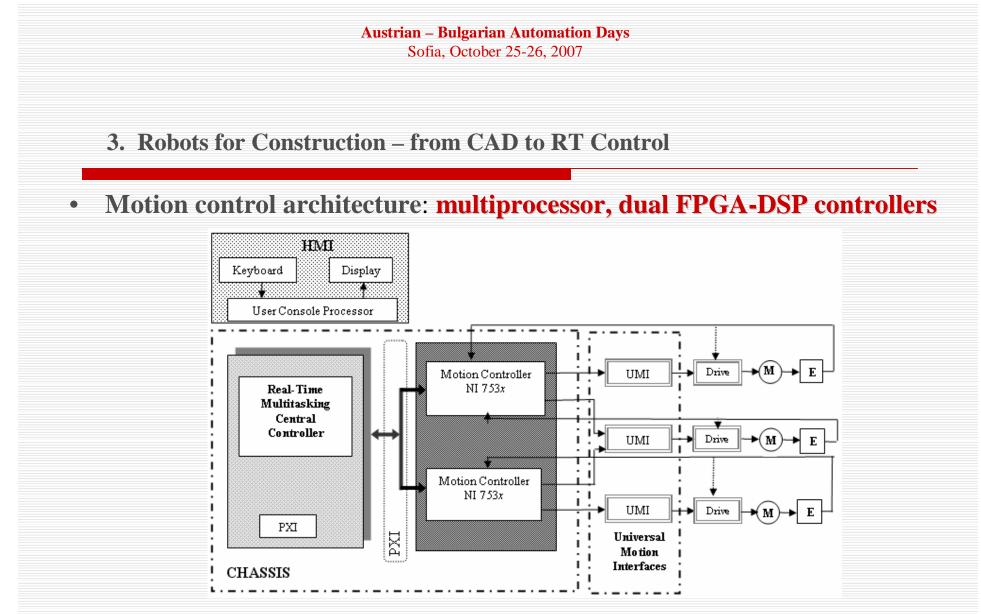
Partitioning a global masonry task in 12 bricklaying subtasks each one associated to a wall segment

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### 3. Robots for Construction – from CAD to RT Control

*Computing the relative transformations* to access respectively the base of the brick pallet and the base of the current wall segment:





12-axis HW implementing solution for robot motion control: 5-robot arm; 3-AGV; 4-platform inclination

## 3. Robots for Construction – from CAD to RT Control

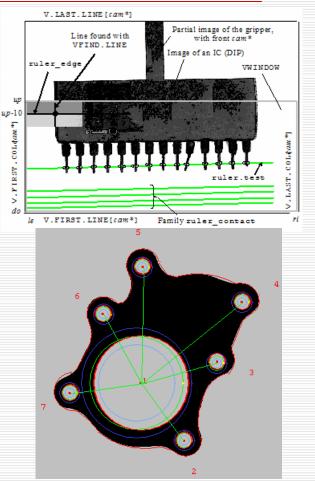
### Modelling & embedded control of robot vehicle & arm motion



RT embedded NI PXI Controller (National Instruments)



# 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks



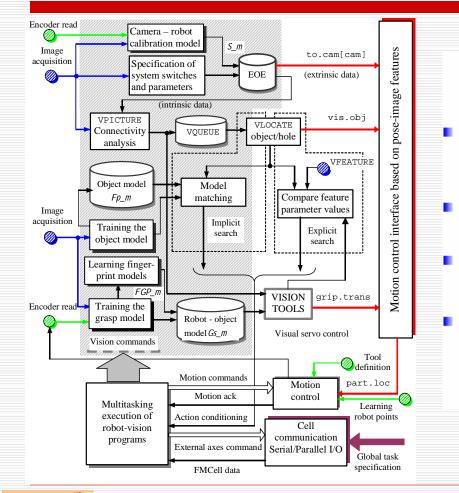
- **Need for implementing machine vision systems in robotics and manufacturing:** AI techniques are applied to create the best vision environment and adapt processing to lighting variations and part flow characteristics.
- How did the distinction between Guidance Vision in Robotics (GVR) and Automatic Visual Inspection (AVI) become blurred?  $\rightarrow$  More and more inspection tasks require manipulation, and more and more component assembling / material processing tasks require quality inspection.
- Why is intelligence needed for industrial vision systems? Two technologies that have hitherto been almost disparate: *Image Processing* (IP) and *Artificial Intelligence* (AI) are currently being integrated. Of special interest are the tasks of *inspecting and manipulating industrial artefacts*.
- Important vision applications in industry: automotive, electronics, semiconductors, robotics, fabricated metal, printing, food/beverage, and pharmaceutical/medical.

## 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks

The primary role of AI techniques in industrial vision systems lies in:

- Inspecting objects that are very complex. (Car engine blocks, complicated moulding /castings, populated PCBs, car body panels.)
- 2. Inspecting assemblies of objects. (Air dryer, automobile carburettor.)
- Inspecting objects which are very variable in form. (Processed food items: chocolates, cakes, loaves, pizza, etc.)
- Inspecting non-rigid objects and those which are composed of articulated levers. (Cable harnesses, leather and fabrics.)
- Inspecting objects that are made in very small batches. (It was estimated that 70% of manufactured goods are made in batches of 50 or fewer items.)
- Aiding in the design process for both GVR and AVI systems. (Knowledge-based systems are finding their way into such tasks as choosing the camera, lens, and lighting arrangement.)

# 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks



Guidance vision for robot part grasping (RVAVI function sets):

Scene functions: accessing and use the *camera*robot calibration model for visually blob locating and measuring in robot world

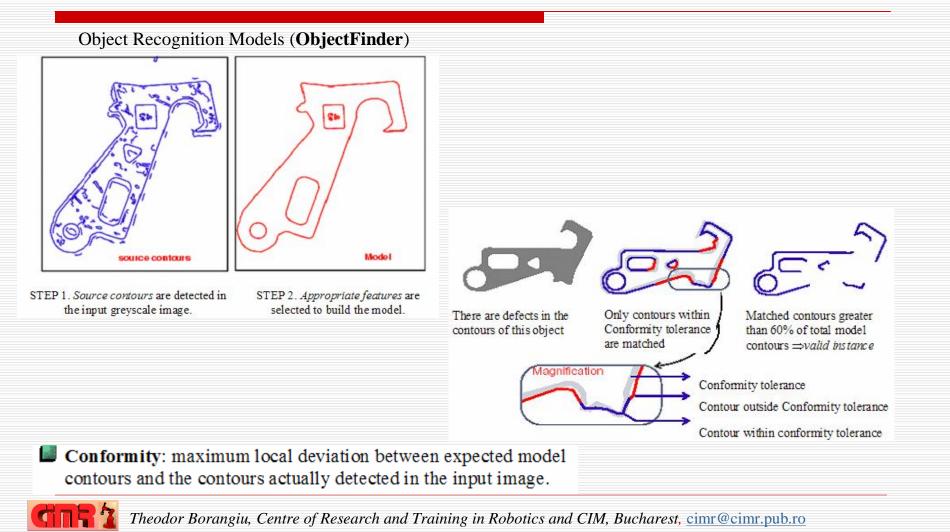
Scene-object functions: learn, plan, and use object models for recognition

Robot-object functions: learn and apply object

grasp models for object classes and grasping styles

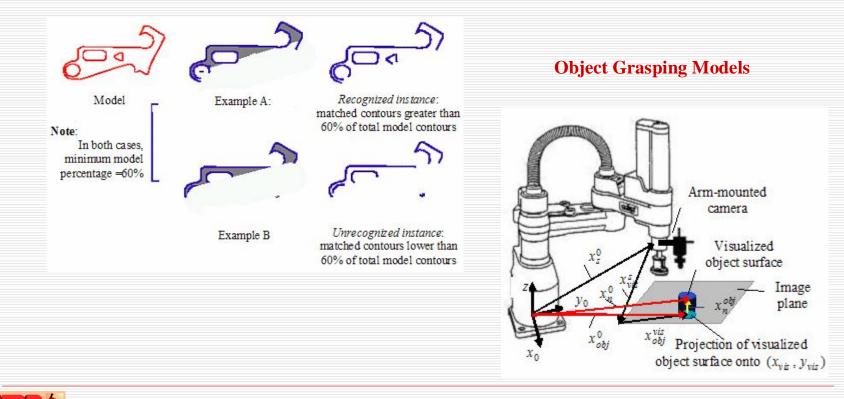
**Robot-scene functions**: define, integrate, and check at run time *fingerprint models* against gripper-object related poses and scene occupancy for *collision-free* access to objects

# 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks

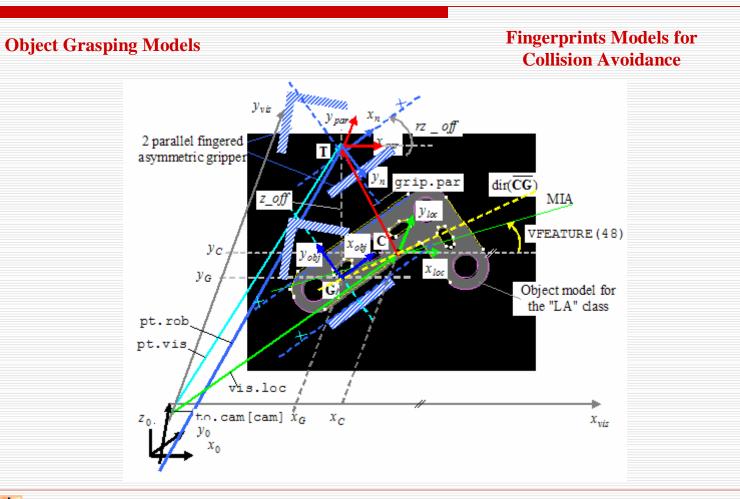


## 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks

Verify%: minimum percentage of model contours that need to be matched in the input image in order to consider the instance as valid. A higher value ⇒ faster recognition & higher rejection rate.

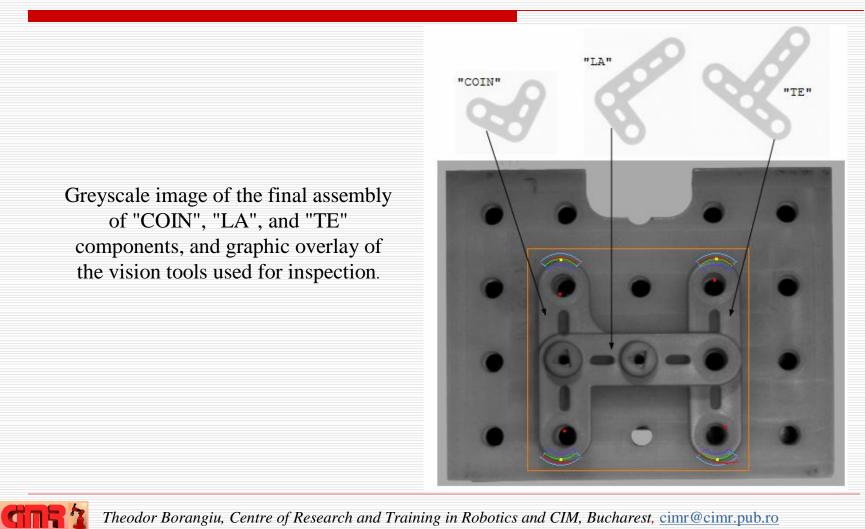


## 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks



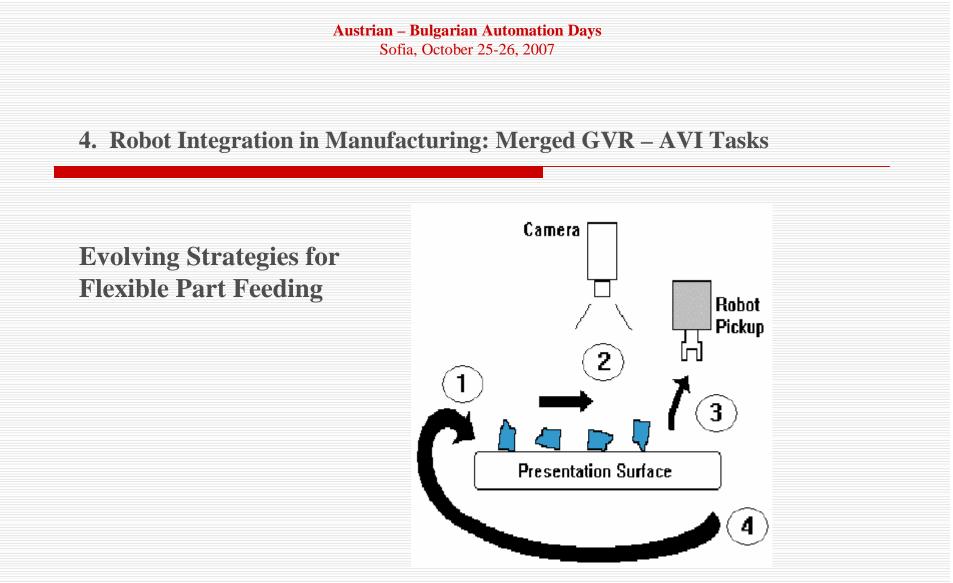
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## 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks



## 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks

eVisionFactory - FordCellCA010 - | X File Made Login Help **Anchor Feature Detection** eV and Measurements Pattern35 Source Pattern36 C Camer Pattern37 @ File Pattern38 Pattern39 Camera Pattern40 Pattern41 Cursor Pattern42 Screenshot of the vision C mm @ pix Pattern43 -Pattern44 × 167 Pattern45 system user interface MathE Y 470 Meth7 FICOCO during part training, WriteObject WebWare3 B RightPinPresence L 52 Flun Step4 showing a cylinder head ViewPort4 Crossbar PMRightPin Update Image MathB and the features used at Math9 Save Clear WriteObject WebWarer run time by the 3D part Settings Deviation Limits 1/D locating kernel to Calibration Object hand calculate the object's 3D Object Ref Height (mm); 17 pose. Clear Messages Onein Imn Deviations T: 0.00, 0.00, 0.00, P: 0.00, 0.00, 0.00 Origin - 37.10, 88355, 393.01, XAnis: -386.7, -110.40, 389.23, YAnis: 61.89, 87.80, 391.72 T: -150.42, -102.83, 587.78, P: 0.20, 1.18, -0.03 Deviations T: 0.00, 0.00, P: 0.00, 0.00, 0.00 XAvia (mm) Devietions 1: 0.00, 0.00 Load image

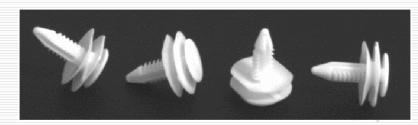


Basic flexible feeding concept: 1 – supply parts; 1 – locate desired parts; 3 – pick qualified parts; 4 – recycle parts that are not picked.

CITE 1

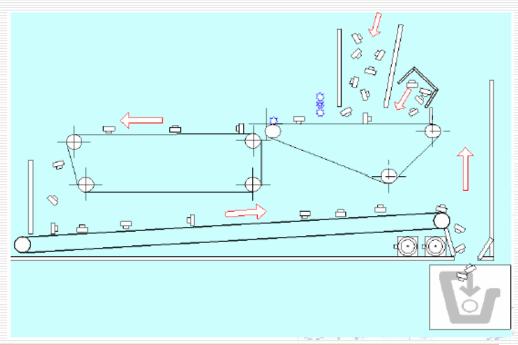
# 4. Robot Integration in Manufacturing: Merged GVR – AVI Tasks

Examples of stable states for some type of part on a conveyor belt.

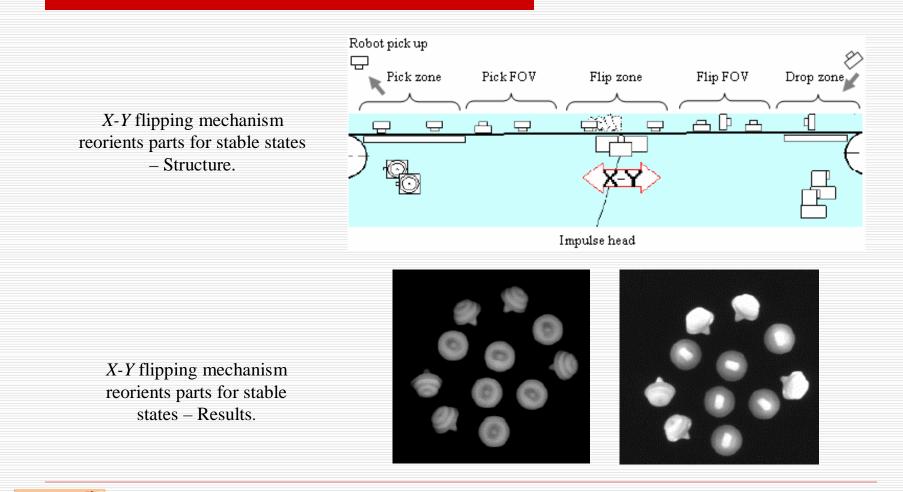


The solution for flexible feeding with continuous mode and part reorientation.

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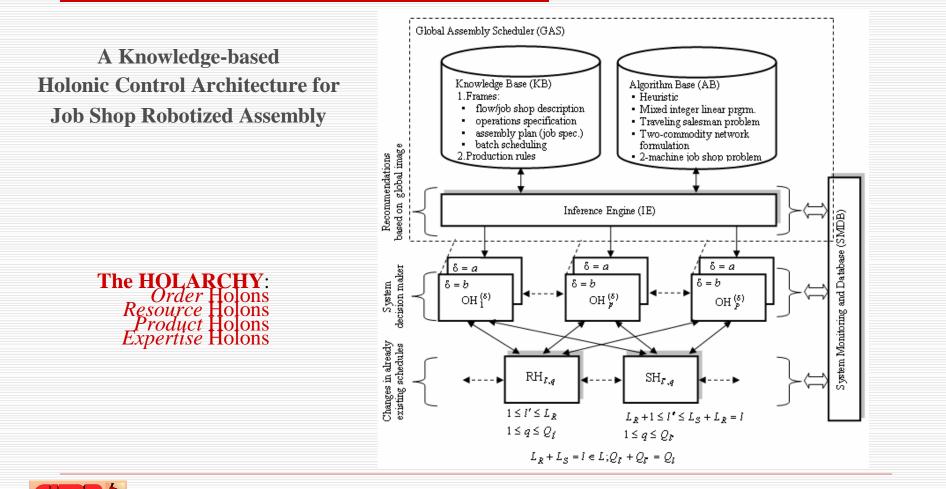


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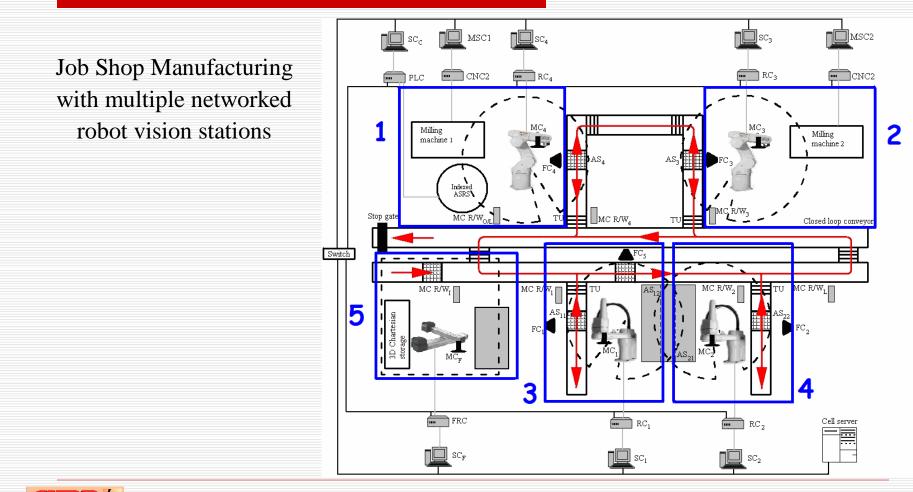


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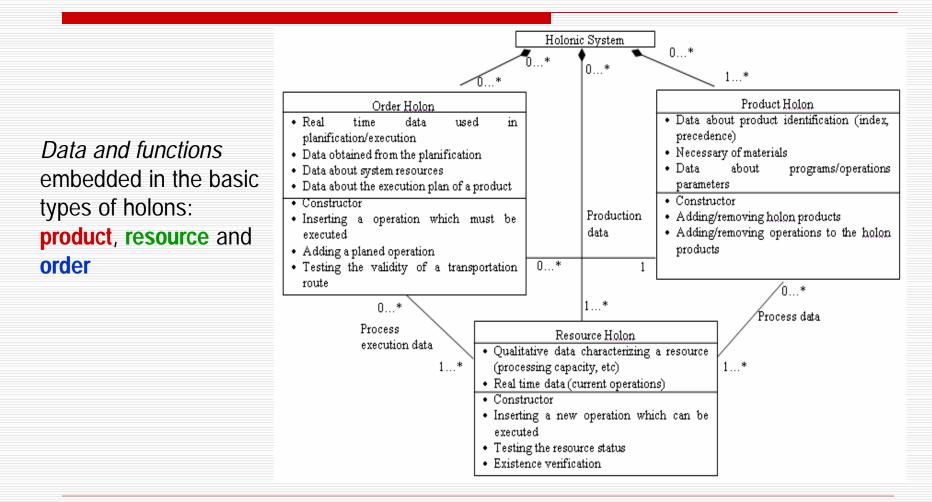
## 5. New Paradigms in Manufacturing Control: the Holonic Approach



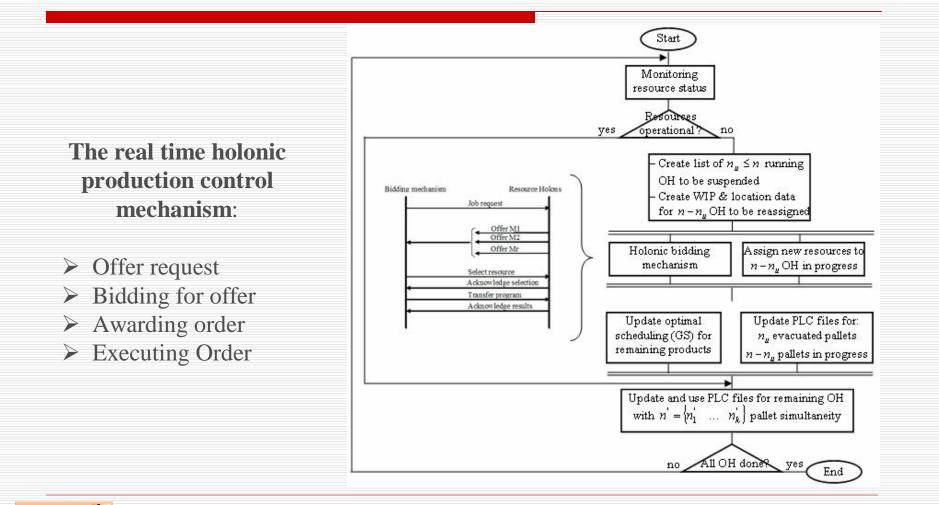
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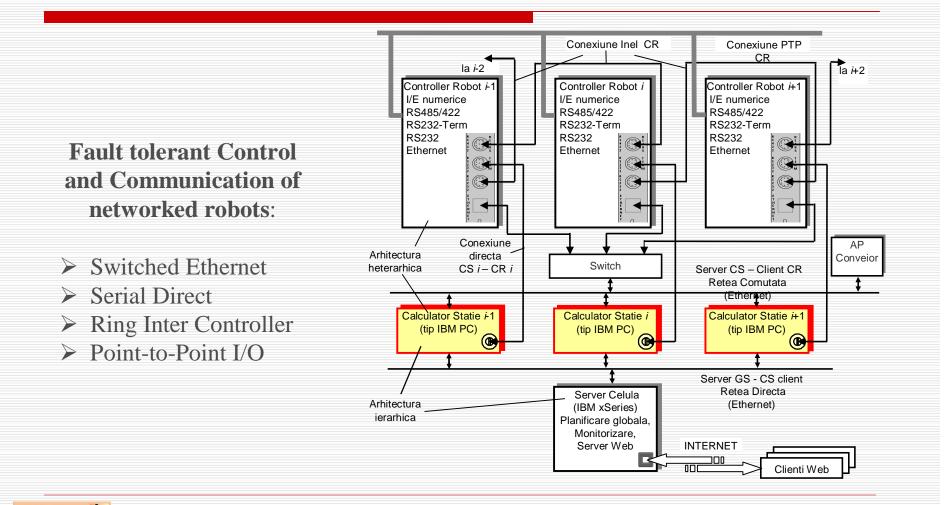
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# 5. New Paradigms in Manufacturing Control: the Holonic Approach



The End

# Thank you !

