



Flexible Manufacturing Line with Multiple Robotic Cells

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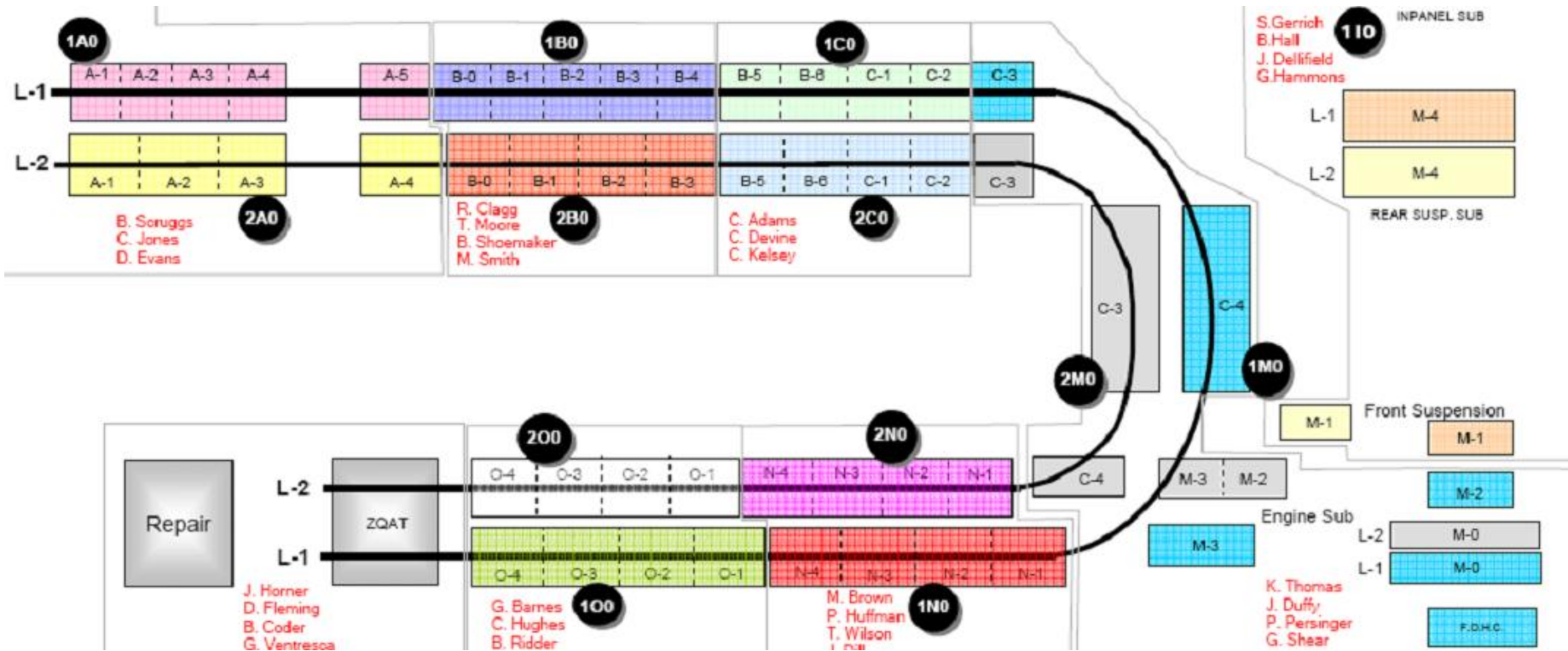
OUTLINE

- Introduction
- Autonomous Robot Teaching
 - POMDP based learning algorithm
- Performance Optimization
 - Single cell optimization
 - Multi cell Optimization
- Discussion





Trim and Final Assembly Line



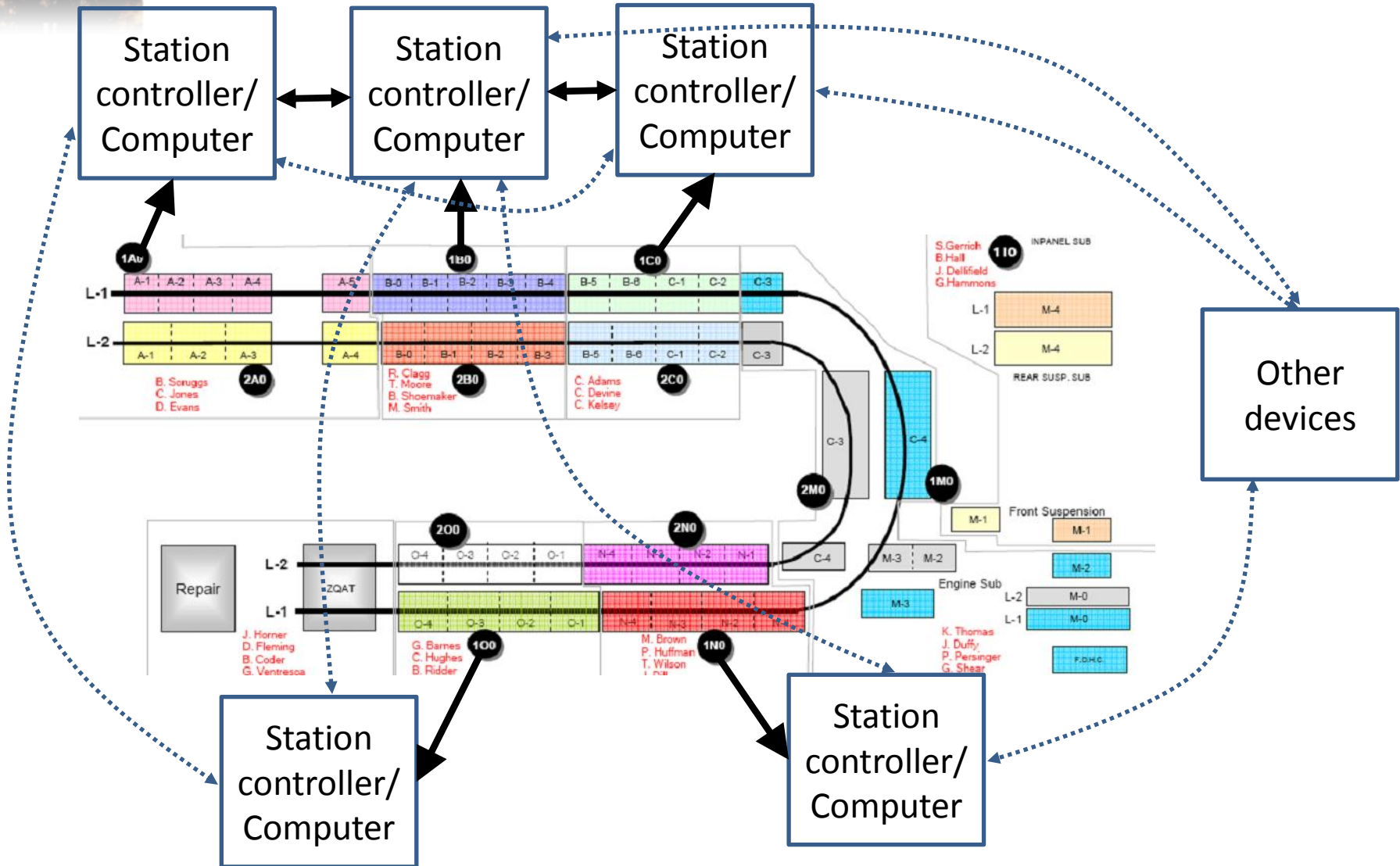


Problems in Production Line

- When one work station fails, the production line has to be stopped
 - Low efficiency
 - High cost
- For complex manufacturing processes, is it possible to optimize the system performance?
 - Single work station optimization?
 - Multi work station optimization?



Proposed Solution



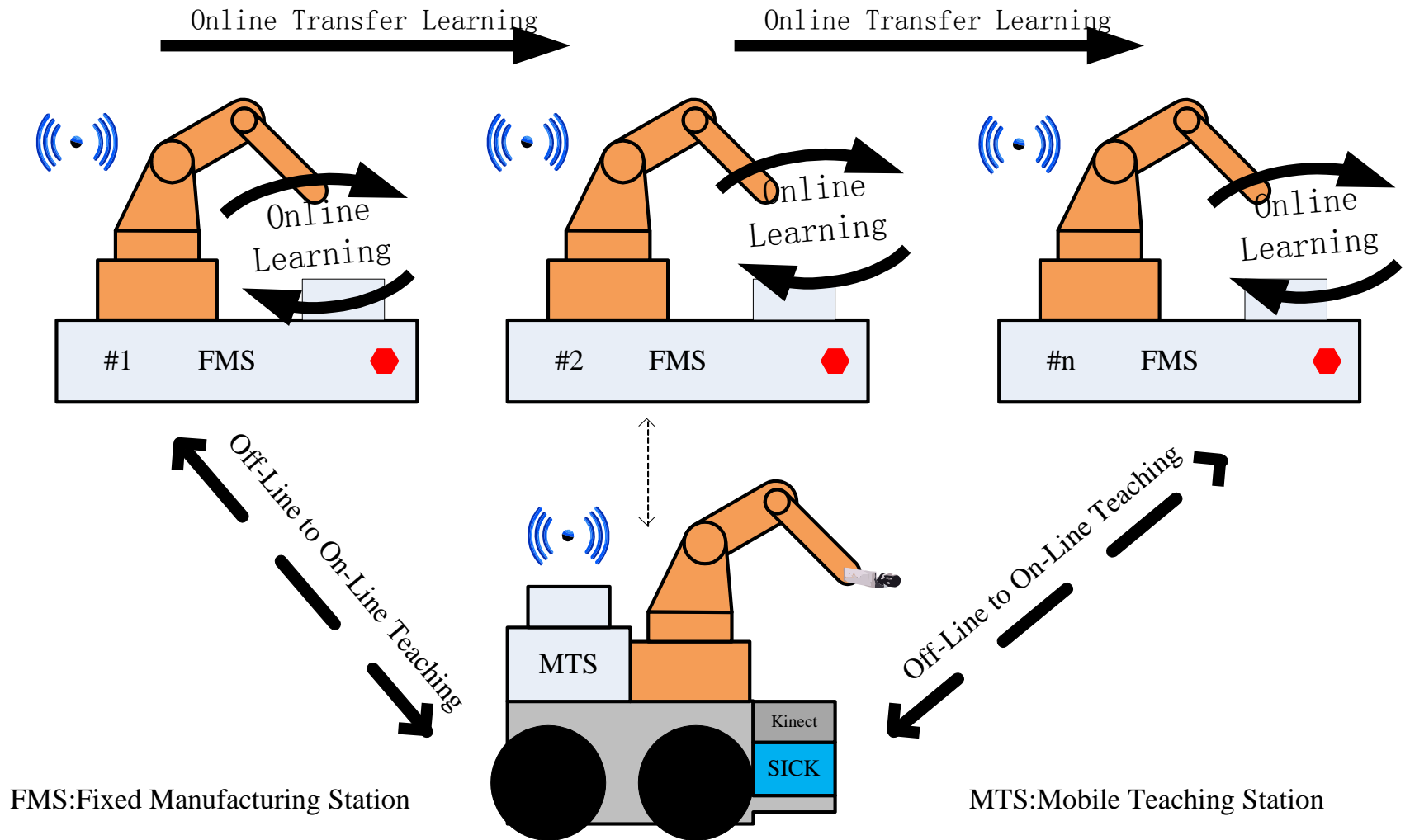


Problems

- Big part location errors
 - Can happen after a new batch is launched
 - Parts could come from different suppliers
- Typically manual teaching methods are used to adjust the parameters
 - Production line has to be stopped
- Installing additional sensors generates new problems
 - System upgrading cost
 - Maintenance issues
 - Sensor failure



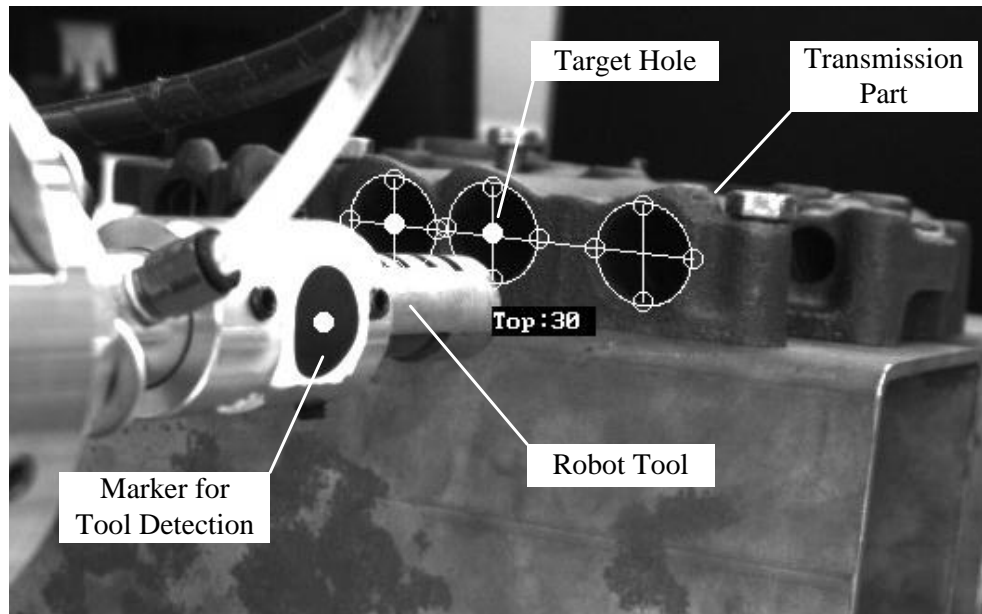
Failure Correction





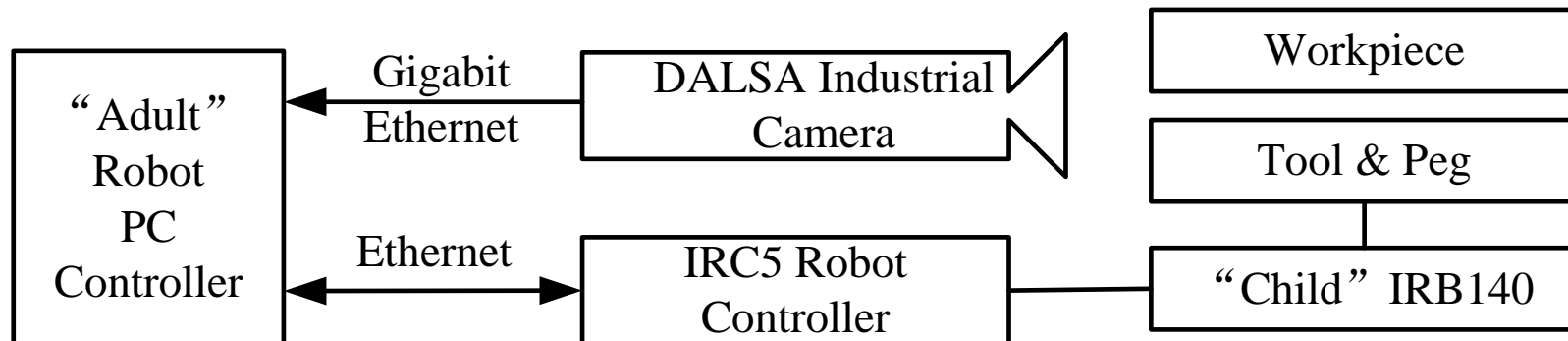
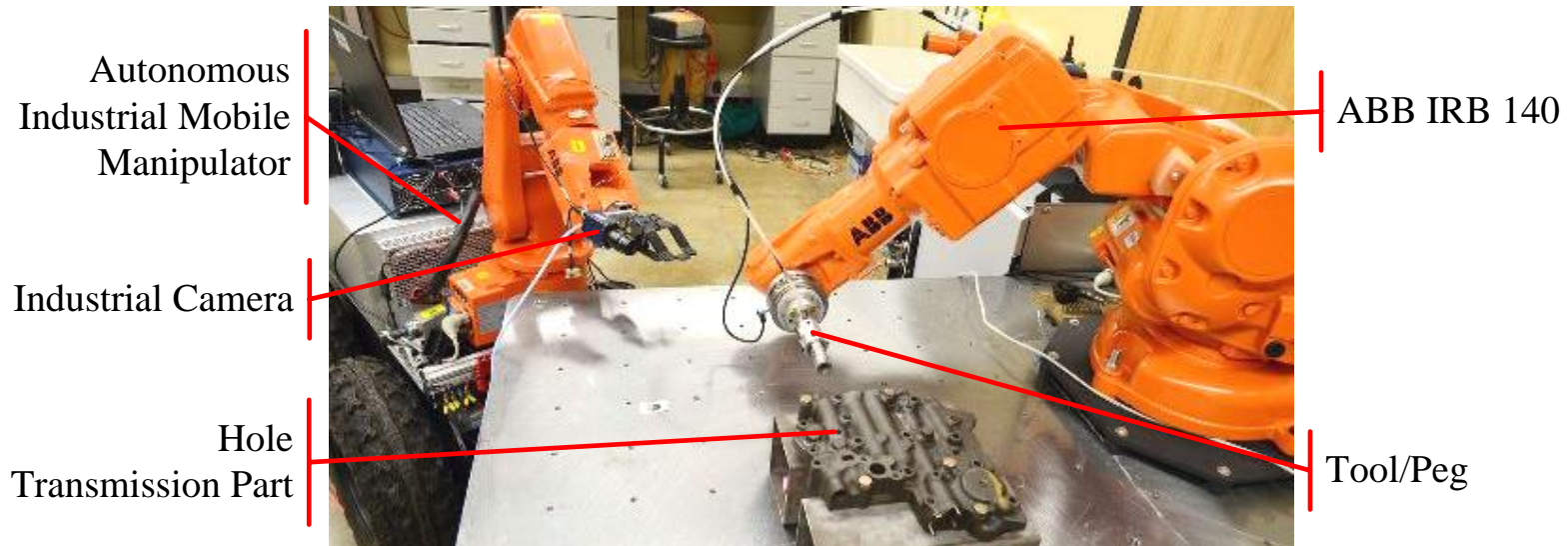
Autonomous Robot Teaching

- Reducing the operational cost
- Sensor accuracy problem
- Calibration is difficult
- Noise





Autonomous Robot Teaching

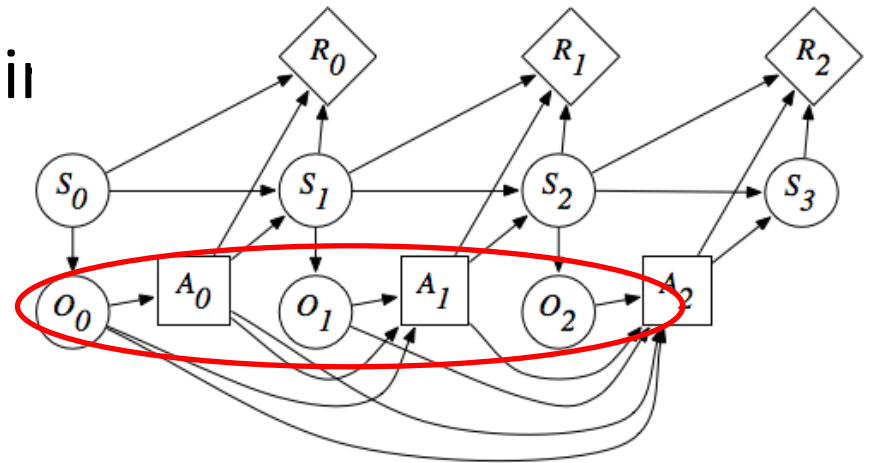




POMDP

- Partial Observable Markov Decision Process

- State Transition is Uncertain
- Observation is uncertain
- State is partial observable



- Why POMDP?

- Using POMDP to estimate the underlying errors through executing actions and receiving observations

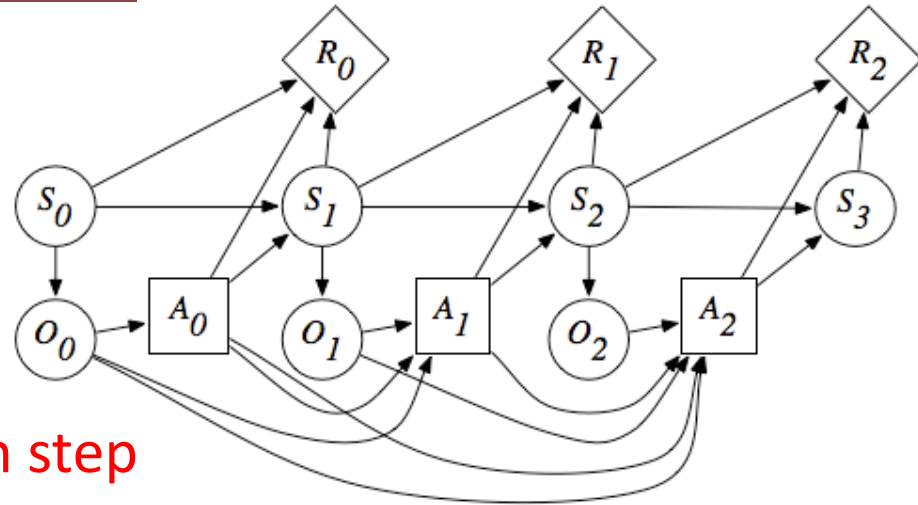


POMDP Method

- **Belief State**

$$\sum_s b(s) = 1$$

- Real state is unknown
- How to make decision?
- Assign a belief b over state S
- Update the belief state in each step



$$b'(s' | b, a, o) = \frac{O(o | s') \sum_s b(s) T(s' | s, a)}{\sum_{s'} O(o | s') \sum_s b(s) T(s' | s, a)}$$

- **Defining Value Functions**

$$V^*(b) = J^{\pi^*}(b)$$

$$J^{\pi}(b_0) = \sum_{k=0}^{\infty} \gamma^k r(b_k, a_k) = \sum_{k=0}^{\infty} \gamma^k E[R(s_k, a_k) | b_0, \pi]$$

$$= \max_{a \in A} \left[r(b, a) + \gamma \sum_{o \in O} z(o | b, a) V^*(\tau(b, a, o)) \right]$$

- **Approximating Value Functions**

- **Alpha Vector**

$$V(b) = \max_{\alpha \in \Gamma} V^{\alpha}(b) \quad V^{\alpha}(b) = \langle \alpha \cdot b \rangle = \sum_i \alpha_i b_i$$



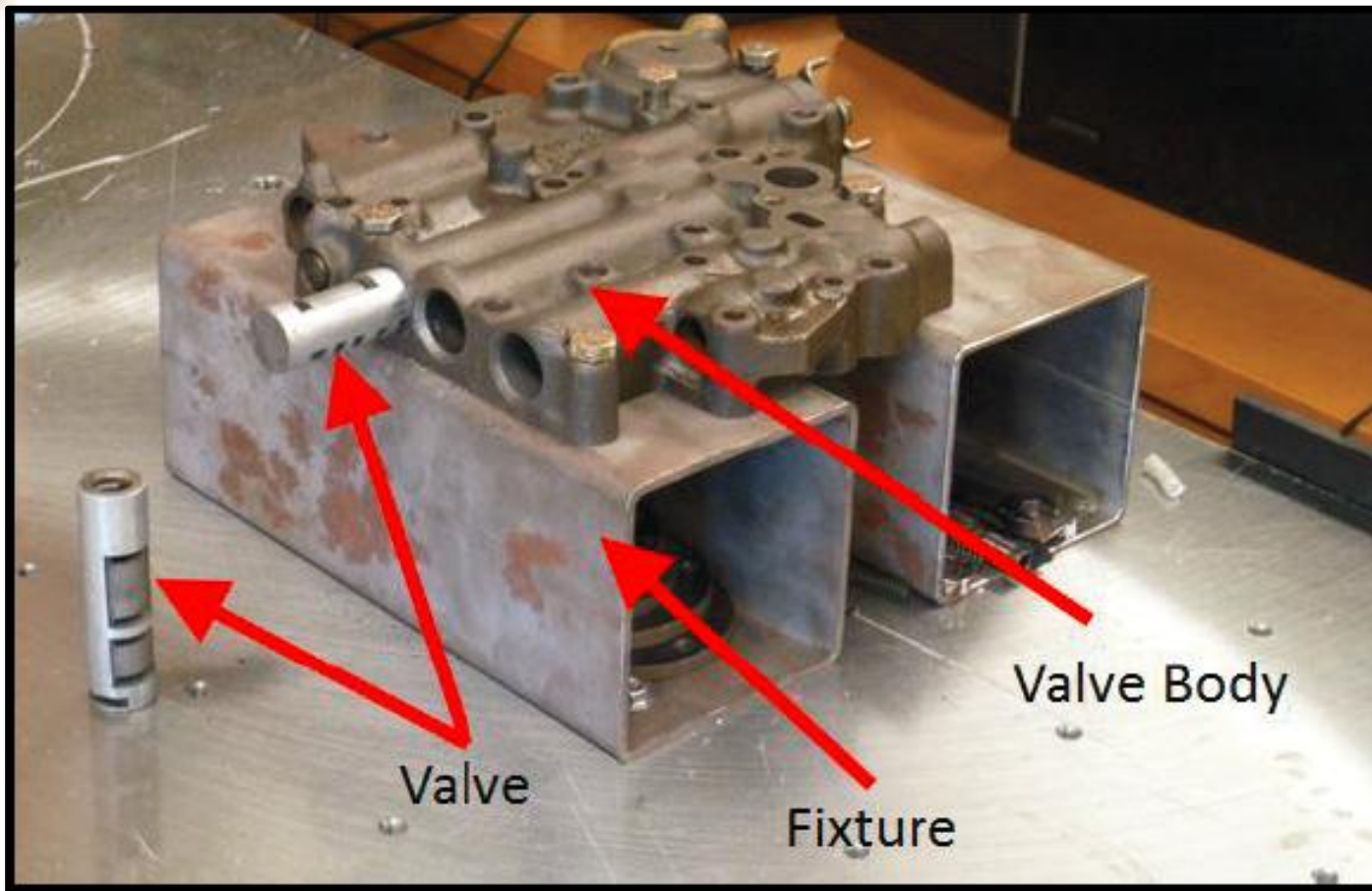
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 - Multi work station optimization
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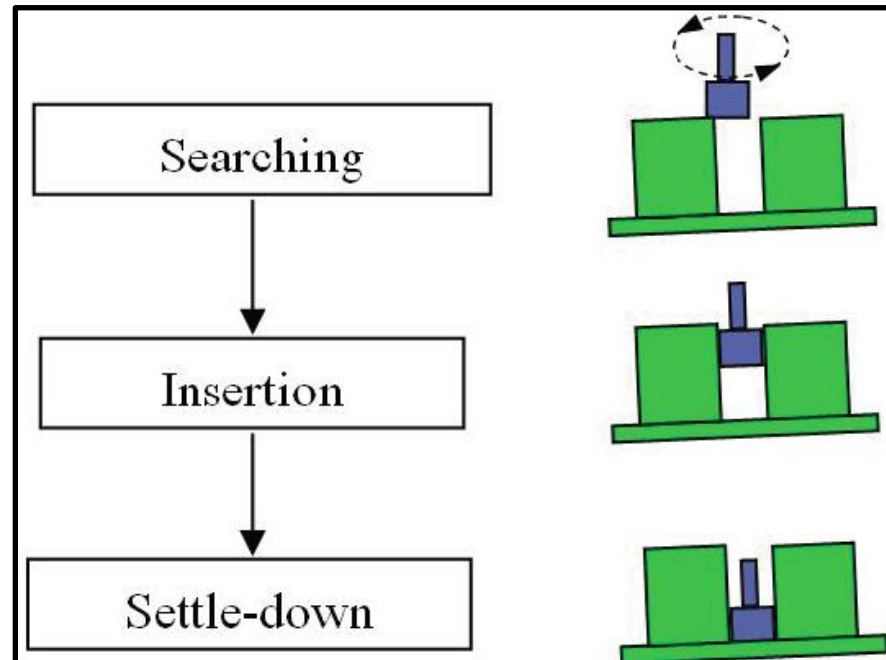
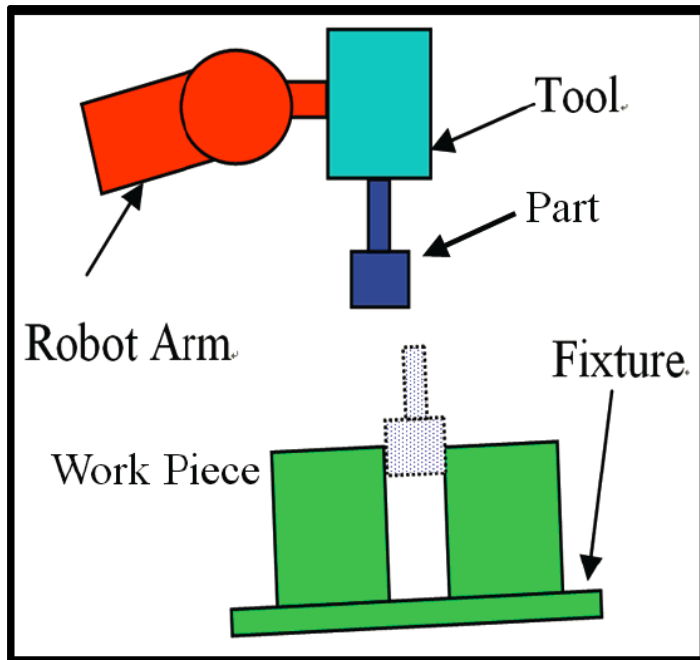


Example: Single Stage Assembly





Assembly Process



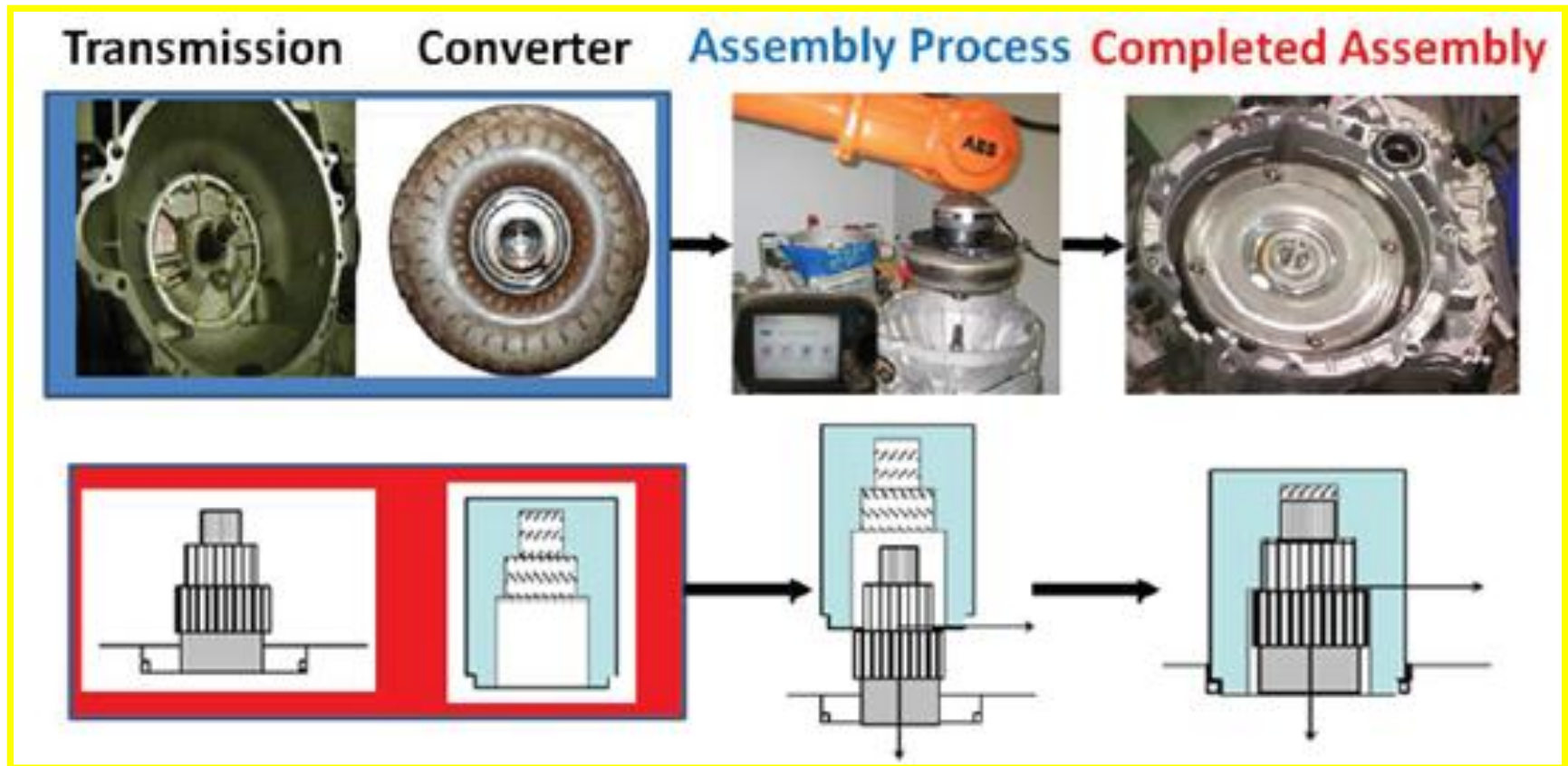
Parameters:

search force, search radius, search speed, insertion force

How to tune the Assembly Parameters?



Example: Multi-Stage Assembly





Problems

- Design of Experiment (DOE)
 - Genetic Algorithm (GA)
 - Genetic Algorithm (GA)+ANN
- } • Model Free
- Offline
- Manufacturing line has to be stopped
- Cannot deal with variations, including part location errors, part geometry errors and environmental errors.

Low Efficiency
Low Accuracy



Single Work Station Optimization

- First Time Through (FTT) rate

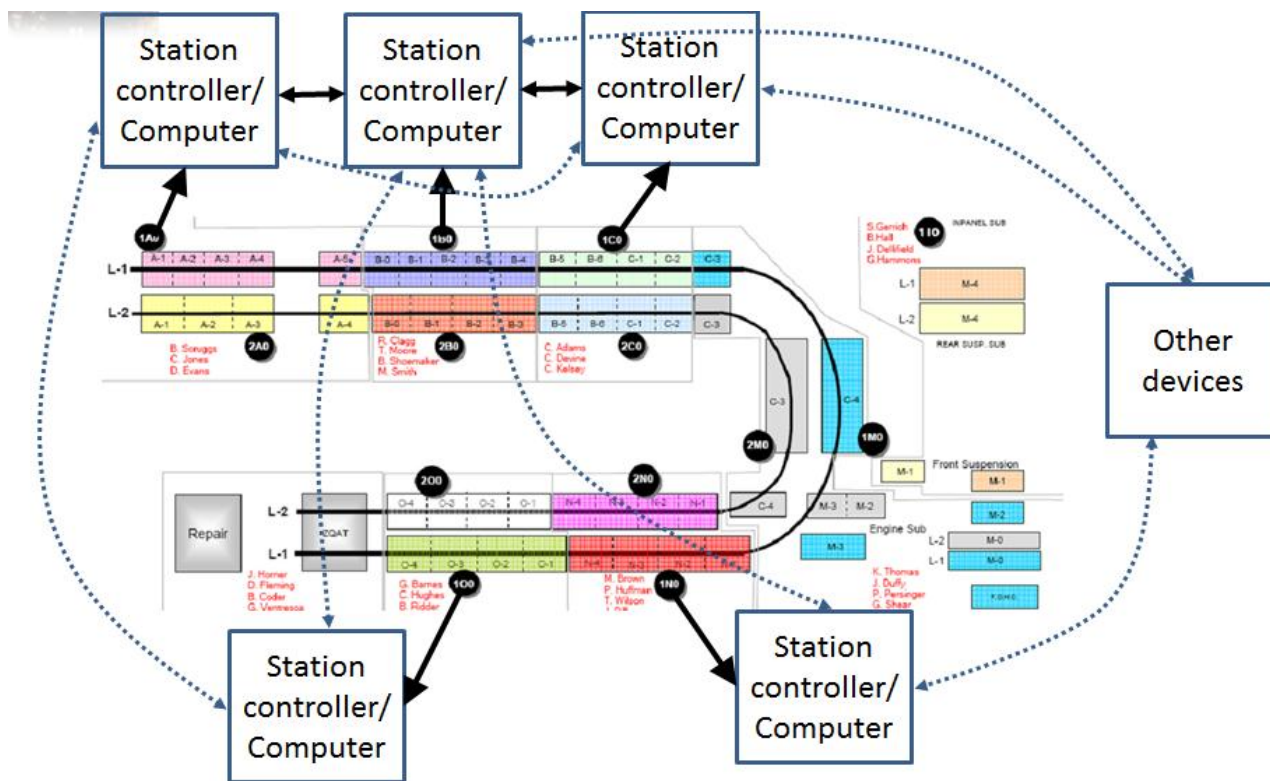
$$\bar{r}_f(\mathbf{x}, \mathbf{T}_c) = \frac{1}{n} \sum_{i=1}^n \rho(t_i, \mathbf{T}_c); \quad \rho(t_i, \mathbf{T}_c) = \begin{cases} 1 & t_i \leq T_c \\ 0 & \textit{otherwise} \end{cases}$$

- How to balance FTT rate and cycle time?
 - FTT rate is calculated statistically.
 - Cycle time is recorded each assembly
 - Modeling?
 - Optimization methods?
 - FTT prediction? Cycle time estimation?



Multi Work Station Optimization

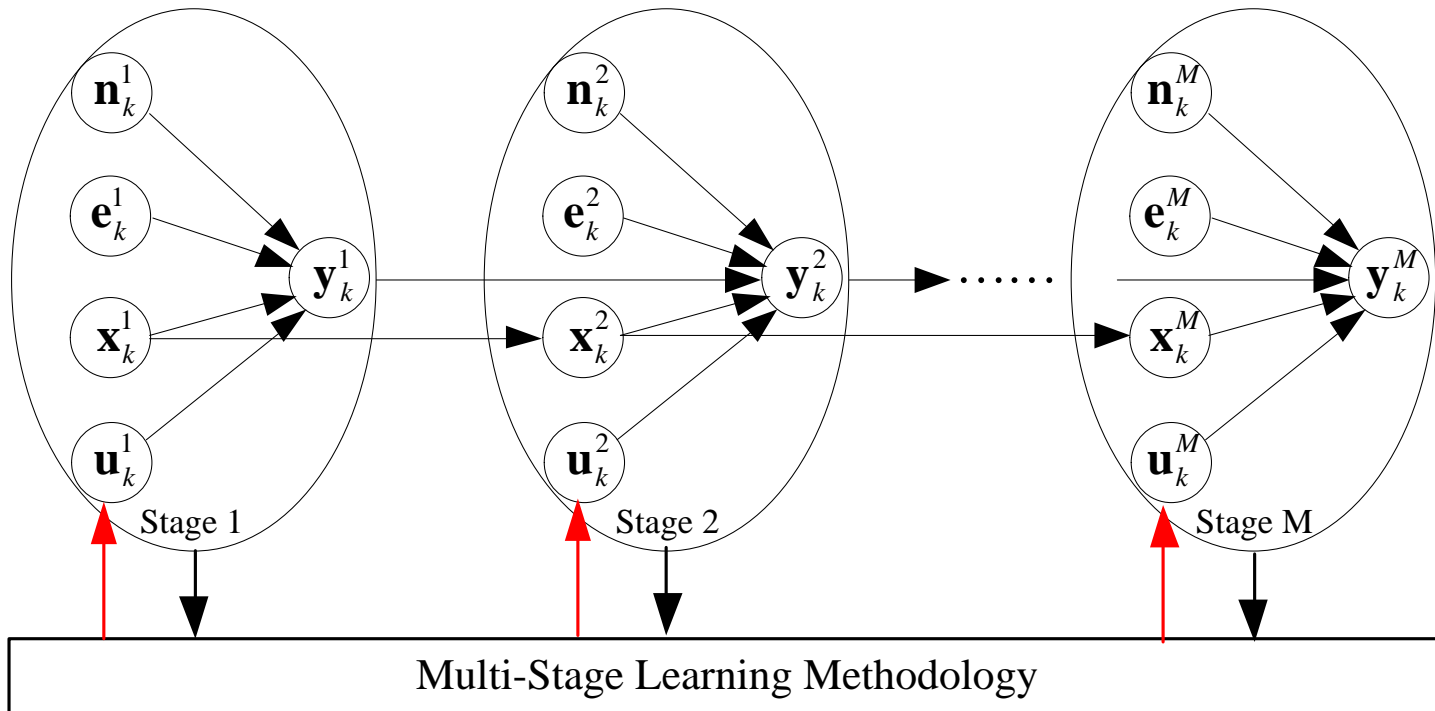
- FTT rate of the whole system
- Cycle time of the production line





Performance Optimization

- Performance optimization
 - Knowledge transfer
 - What can we transfer? How to transfer?





Discussions

- **Comments**
 - Robot teaching
 - System optimization
- **Interesting topics?**
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