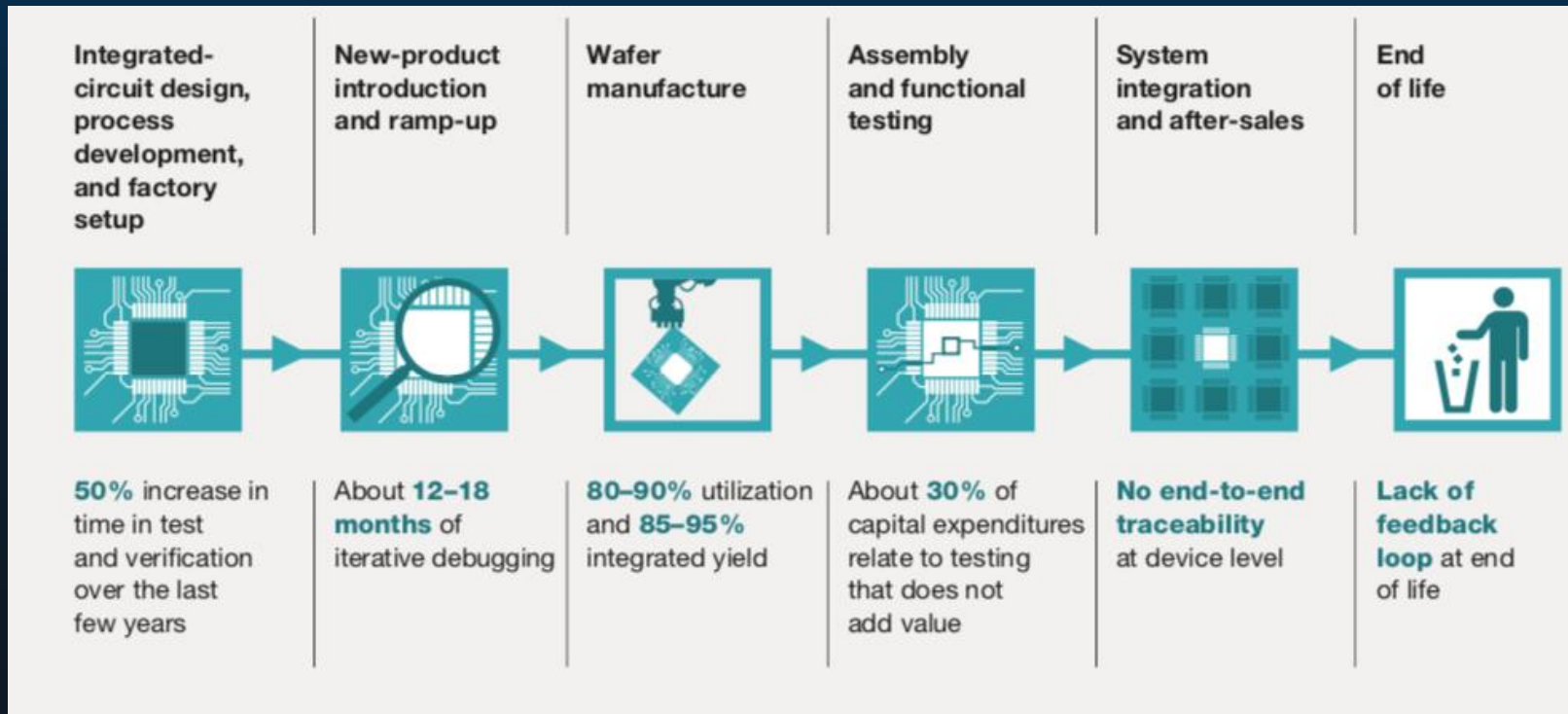




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**Applying Machine Learning to Improve Production &  
Yield for Semiconductor Fabrication**

# Increased Complexity in Fabs Needs New Approaches

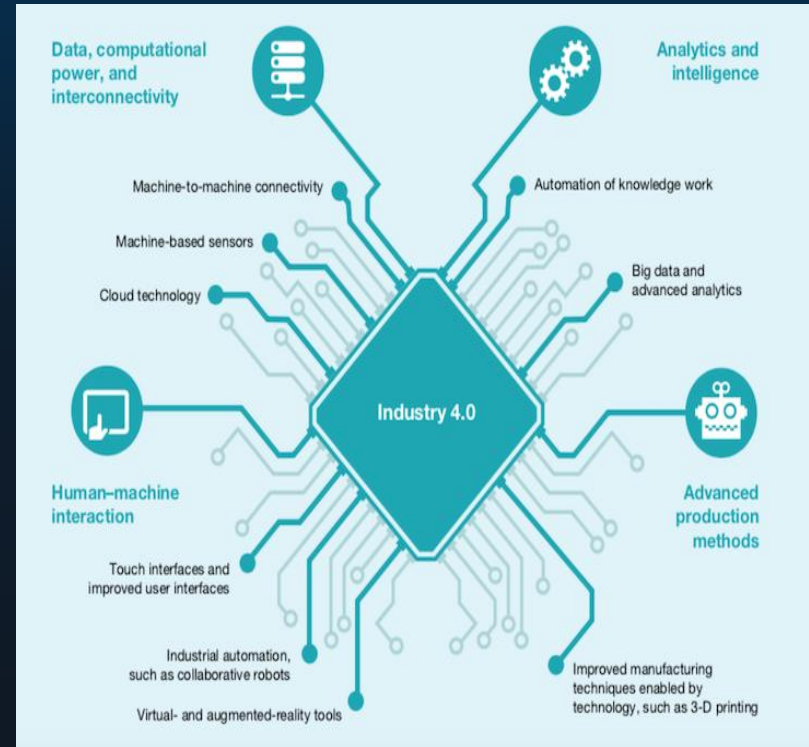


Source: McKinsey - Advanced Analytics in semiconductor manufacturing, October 2017

# Machine Learning Transforming Semiconductor Manufacturing

## Impact of Machine Learning

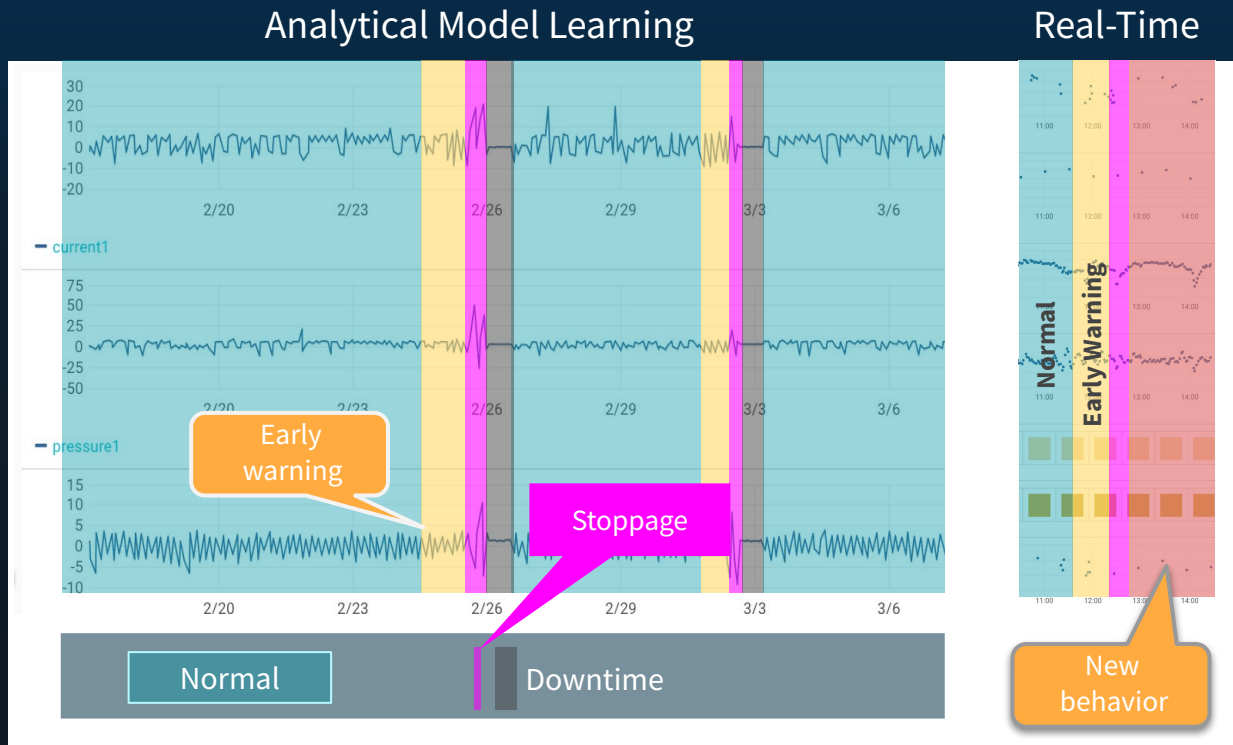
- 1-5% improvement in yield
- 10-20% gains for maintenance productivity
- 20% reduction in equipment downtime
- 25% reduction in inspection costs



Source: McKinsey - Optimizing semiconductor manufacturing

# Machine Learning Enables Predictive Operations

Enable time series pattern discovery and early warning



# Predictive Operations Use Cases

## ✔ Yield & Quality Improvement    ✔ Predictive Maintenance    ✔ Root Cause Analysis

Improve yield by identifying process variables that affect defect density and other on-wafer outcomes

Identify excursions by classifying known events as they recur

Identify early lifetime failures to avoid unscheduled downs

Identify early periodic failures to schedule proactive maintenance

Optimize preventative maintenance schedule to reduce consumables costs and minimize lost production

Signal contribution indicates subsystems involved

Signal clustering allows efficient investigation of new phenomena

# Quality Improvement / Yield Improvement



## Yield Improvement with Defect Density Analysis

New monitoring method to reduce inspection rate

- **Problem:** Defect inspection is an expensive process with inherent delays  
Using sensor data to reduce inspection rate can save a fab money
- **Cost:** Reduced fab throughput due to inspection time, capital cost of inspectors, WIP at risk due to inspection queuing delay
- **Solution:** Falconry LRS can identify correlations between sensor levels and defectivity. Customers can then use those correlations to drive a more efficient defect inspection sampling plan
- **Benefit:** Reduced inspection costs and yield loss



# Defectivity Analysis

- Wafers are etched and then cleaned
- Signal data from a number of high and low defectivity lots measured post-etch were analyzed by Falconry
- Falconry performed unsupervised modeling using ~75 signals and found correlation between cluster groups and defect density
  - Bias/clamp voltage, coolant flow, current, gas flow, pressure, temperature, throttle valve position, pump speed, ring position
- Falconry's predictive analytics helps customer understand which wafers are likely to have a high defect density
- Those wafers can be preferentially sampled thereby decreasing total inspection sampling rate and cost while maintaining low risk of missed defect excursions.



# Pattern Discovery: Two Chambers From The Same Equipment



## Note:

- Each batch comprises of a single wafer that undergoes etch followed by a wash cycle
- Lower % occurrence of a pattern suggests lower quality/yield and higher defect density

# Discovering Important Signals For “Low Defect Density”

-11.09 minutes 2018/01/20 20:55:12

M

[1]

*High defect density batch*

Show Explanation Scores

Signal	Explanation Score ^
IB5Value	1.00
StepID	1.00
RF27MHzGen...	0.73
RF27MHzGen...	0.68
RF2MHzGenR...	0.68
ForelineMano...	0.63
Gas_12_Flow...	0.63
IB4Value	0.63
IB7Value	0.63
RF2MHzGenD...	0.63
RF2MHzPACu...	0.63
TopPlateHeat...	0.63
Gas_11_Flow...	0.57
IB1Value	0.57
RF2MHzPADi...	0.57
ThrottleValve...	0.57
Gas_7_Flow_AI	0.52
Gas_9_Flow_AI	0.52
RF27MHzFre...	0.52
RF2MHzSetp...	0.52
TuningGas_Fl...	0.52
Confinement...	0.47
Gas_8_Flow_AI	0.47

75 chamber signals



**Operational  
Machine  
Learning**


Top 10 signals



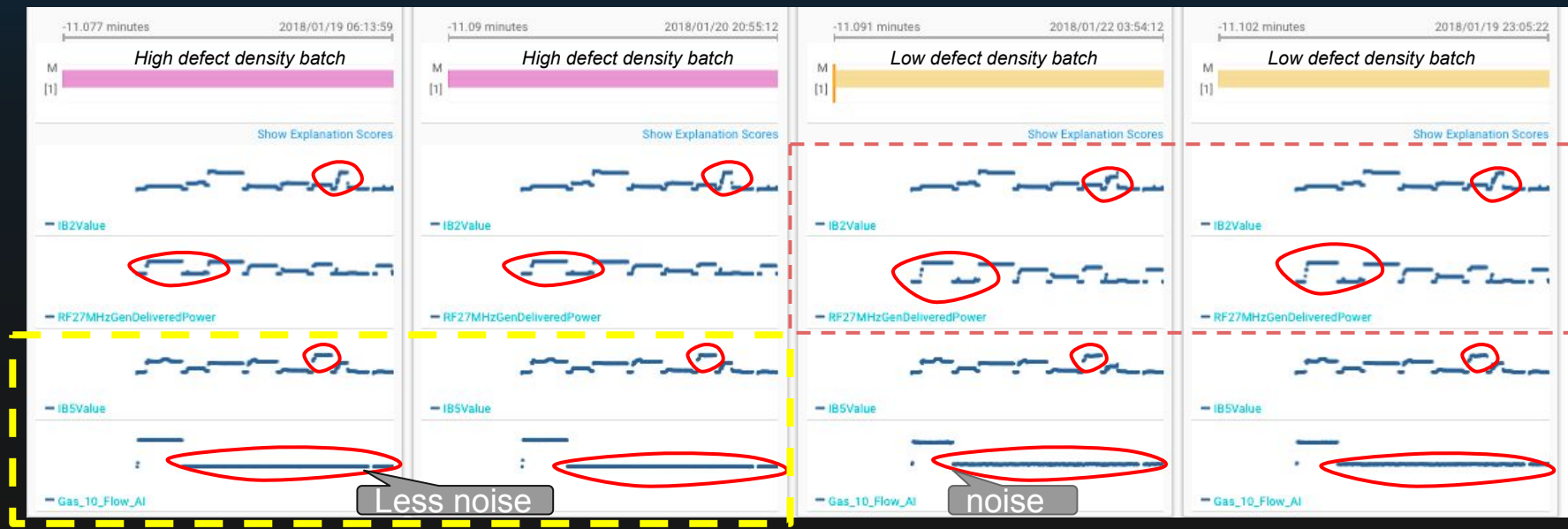
1. IB2 Valve
2. Foreline Manometer Adjusted Pressure
3. ESC Coolant Flow
4. Gas 11 Flow
5. RF 2 MHz PA Dissipation
6. RF 27 MHz PA Dissipation
7. Top Plate Heater Temperature Output Value
8. Gas 1 Flow
9. Gas 3 Flow
10. RF 2 MHz Gen Forward Power

# Run-to-run Comparison of Discovered Patterns

Wafers have process signatures not easily discernible to the human eye

 Higher impact signals for LOW defect density wafer batches

 Higher impact signals for HIGH defect density wafer batches



# Predictive Maintenance

## Detect Impending Mass Flow Controller (MFC) failures

### Preventative maintenance for production efficiency (Dry Etching)

- **Problem:** An imbalance in gas flow can result in too many reactive species in the chamber and loss of anisotropy in the etch process
- **Cost:** Unscheduled maintenance and yield degradation
- **Solution:** Detect etchant MFC anomalies days in advance of potential failure
- **Benefit:** Convert unscheduled downtime into scheduled downtime for more efficient fab operations



falconry

# Predicting Malfunction of MFC in Etch Chamber

- Dry etchers used to fabricate wafers have several MFCs that control the release of gas etchants with precise measure
- These MFCs may start malfunctioning and eventually fail
  - In the meantime, they also impact the quality of the wafers being etched
- **Cost of undetected failure:**
  - Reduced yield
  - Unscheduled downtime waiting for maintenance team slot to open or for long lead time parts to be obtained
- Customer used LRS to detect patterns of MFC behavior which indicated failure of the MFC up to a week in advance
- This allowed the customer to schedule equipment maintenance, minimize the time equipment was running in poor state and avoid an unplanned system down

# Detecting Gas Valve Malfunction

Failure detection one week in advance

Model trained on one etch chamber



Actual failure



Failure identified by Falkonry  
(Early Warning)

Same model applied on another etch chamber with no facts provided





# Summary of Use Cases

Applying machine learning to optimize semiconductor manufacturing

