#### **Predictive Maintenance:** Lessons learned from Semiconductor Manufacturing

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# **Electronic Systems**

From Materials to Power Electronic Applications – Everything from One Source

Automotive Electronics

Devices &

Reliability

# **Semiconductors • Power Electronics**





Energy Electronics

Technologies &

Vanufacturing

### Industry 4.0

History





Production

Dr.

### Industry 4.0

Fraunhofer Layer Model of Industry 4.0 Value Creation



- Three Layers
  - Production
  - ICT
  - Enterprise Transformation
- More than **150 topics**
- Driven by experts from 20 institutes
- Knowhow based on more than 300 projects

http://www.academy.fraunhofer.de/





#### Agenda

#### 1. Industry 4.0 in Semiconductor Manufacturing

2. Dr. Production, Predictive Maintenance and Beyond

3. Outlook



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... what comes to mind





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A semiconductor view on "Industry 4.0"

# The most complex production system is a semiconductor Frontend.



Dr. T. Kaufmann, Infineon

11th Innovationsforum for automation, 2014, Dresden

2014-01-23

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infineon





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Some history on standards

#### Most famous standard: "SECS/GEM"

- 1978: Hewlett-Packard proposed that standards be established for communications among semiconductor manufacturing equipment.
- 1980/1982: SEMI published the SECS-1/SECS-II standards
- 1992: GEM standard published
- Continued: HSMS, GEM300, EDA/Interface A, …



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Referring to: http://www.semi.org/en/sites/semi.org/files/docs/AUX023-00-1211.pdf (15.03.2012)



Bridge the "productivity gap" by Advanced Process Control (APC)













"Big data" and Advanced Process Control

- Objective: Ensure high productivity and product quality
- Fundamental goals of APC ("Advanced Process Control")
  - to apply measures for process control close to the process
  - to automate control actions
- Typical APC methods (SEMI E133):
  - SPC, FDC, FP, RtR, VM, PdM
- Basis for APC:
  - Metrology data
  - Data from equipment & processes
    - Logistics data



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#### Agenda

1. Industry 4.0 in Semiconductor Manufacturing

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Part of the Fraunhofer Layer Model of Industry 4.0 Value Creation



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The Building Blocks – Industry 4.0 Components in a Nutshell







The Building Blocks: Analysis of Manufacturing Processes & Data Acquisition/Analysis

Data-driven manufacturing optimization provides a variety of optimization methods we will help to find the right one. Feedback Process data Fault detection and classification Run-to-run uncovers anomalies of systems/processes control Feed Process Download Run-to-run control automatically of parameters data forward modifies process parameters to improve process results Predictive maintenance predicts the need (Virtual) (Virtual) Process n-1 Sensors Process n+1 for service and maintenance measures Metroloav Metrology Process n Virtual metrology enables the data-Process go / no go driven prediction of quality parameters data Predictive Fault detection and Maintenance (PdM) classification (FDC) 🗾 Fraunhofer Production Dr © Fraunhofer IISB, 01-2018

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The Building Blocks: Consulting & Conception

We develop concepts for step-by-step implementation of data-driven production optimization and we clarify necessary conditions.



**Example:** The benefits of implementing predictive maintenance through introduction of data-driven production optimization is greatly depending on the target equipment (right: analysis in semiconductor manufacturing).



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The Building Blocks: Development of Intelligent Algorithms (1/5)

We develop intelligent algorithms that analyze production data in real time and suggest or take necessary measures.

#### Example 1: Predictive maintenance in ion-implantation

Relevant data are continuously collected and analyzed using a Bayesian network. The network predicts the real time remaining until the break of the filament with an accuracy of 10-20 hours. -Prognosis -Real time to breakdown -Real time to breakdown -Real time to breakdown -Real time to breakdown

Based on these forecasts, the maintenance tasks can be scheduled exactly. Thus, no "mere" preventive maintenance after a predetermined operating time or number of processes needs to be carried out, and no system failures are risked by missed maintenance steps.



The Building Blocks: Development of Intelligent Algorithms (2/5)

We develop intelligent algorithms that analyze production data in real time and suggest or take necessary measures.

#### Example 2: Virtual metrology for deep-trench etching

Relevant data are continuously collected and evaluated using a "gradient boosting tree" algorithm. The algorithm predicts the actual depth of the trench after the etching process with a deviation of less than 4 nm compared to values obtained from physical metrology.



The application of virtual metrology allows "virtual" control of every single wafer, while regular, costly and time-consuming physical measurements can be limited.

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The Building Blocks: Development of Intelligent Algorithms (3/5)

We develop intelligent algorithms that analyze production data in real time and suggest or take necessary measures.

#### Example 3: Predictive maintenance in wire-bonding

Relevant data are continuously collected and evaluated to correlate quality parameters and equipment parameters with the wear status of the wedge.

First results show potential for an optimized wedge tool usage.









The Building Blocks: Development of Intelligent Algorithms (4/5)

We develop intelligent algorithms that analyze production data in real time and suggest or take necessary measures.

#### Example 4: Detection of critical equipment states related to manual interaction

Relevant data are continuously collected and correlated to equipment states that cannot be measured directly (e.g. state of clamping, torque).

First results show that the algorithm is able to detect critical equipment states and to give measures for optimization.







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The Building Blocks: Development of Intelligent Algorithms (5/5)

We develop intelligent algorithms that analyze production data in real time and suggest or take necessary measures.

#### Example 5: Predictive Probing to reduce time for device test

Relevant data from upstream processes are continuously collected and analyzed to predict device properties without actually measuring them.

Accurate interpolation of LED properties with < 5% measured chips achieved, with significant time and cost savings.



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Opto Semiconductors

LED-chips to be measured LED-chips not to be measured

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The Benefits

We have validated the monetary benefits through the use of data-driven production optimization in several reference projects.



Typical benefits from the introduction of predictive maintenance.

**Example:** Amortization of the cost of predictive maintenance (hardware, software, implementation costs, ...) in various systems in semiconductor manufacturing in less than 24 months.









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#### **Cognitive Power Electronics 4.0**



## **Semiconductors • Power Electronics**





### Outlook

#### Cognitive Power Electronics 4.0



#### Vacuum System

- Condition Monitoring
- Predictive Maintenance

#### Equipment

- Fault Detection and Prediction
- Process Optimization
- Predictive Maintenance

#### Intelligent Grid

Office Test B

- Control, Stability
- Optimiziation of Interaction: Source – Storage – Load

DC-Grid

DC-Labor

tationäre Speich





600 ... 800 V



#### Outlook The Chance of Working Together

#### Lessons learned from 14.0 in semiconductor manufacturing

- 1. Collaborate (competitors, universities, ...)
- 2. Know your process
- Make use of standards 3
- 4. Good to have data from >1 year of production
- Take care of data quality 5.
- 6. Combine knowledge of data experts and process experts
- 7. Go for low-hanging fruits ...
- 8. ... but avoid "island-solutions"
- 9. Collaborate (beyond industry segments)



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# Thank you for your interest!



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