

How to predict the Unforeseeable?

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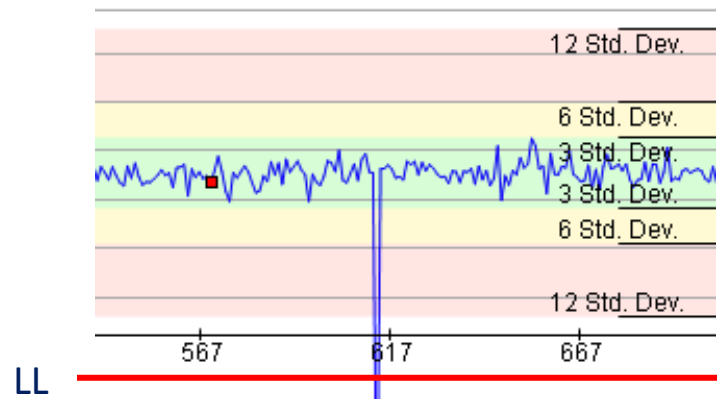
The Power of $[\infty]$

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'Random' Fails – *the* Challenge in Adaptive Test



- Spot Failures in Devices under Test (DUTs) are among the most notorious sources of yield loss in semiconductor production.
- How to 'control' such a test?

By Sampling?

➔ Sampling rate of 10 → Cpk over of 10 fully tested parts before the fail equals 4.71!

➔ incapable of predicting 'random' defects!

The 'Miracle' and what's best done of it

→ 'DUT Response Model'

- Devices, failing in a certain test, often 'announce' that deficiency in specific tests earlier in the flow by showing 'abnormal' readings!
- Thus, not (only) the history of a test may predict its actual result but the outcome of earlier tests on the same DUT!

Our Focus

- Real-time monitoring of specific '**Signature Test**' results
- 'Very short-term' predictions
- Test flows are dynamically changed by adding or removing **Adaptive Test Groups**
- Execution of those groups is a function of prior Signature Test readings on the same device

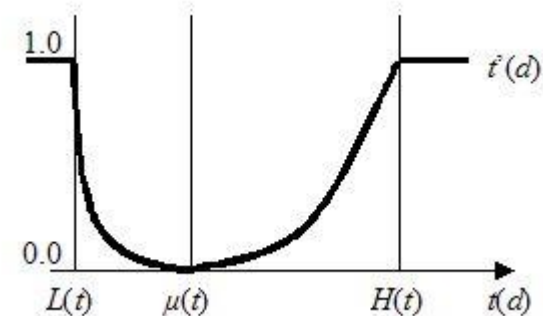
Knowledge Correlations accentuate critical results

- The crucial ingredient in a reliable DUT Response Algorithm is the determination of the Signature Tests!
- Compute Knowledge Correlations from specific test results:

Define the normalized score $t'(d)$ of a test result $t(d)$ for a test t with limits $L(t)$, $H(t)$ and 'target value' of $\mu(t)$ as:

$$t'(d) = \begin{cases} 1 & \text{for } t(d) < L(t) \\ F(t)_{\text{lower}} & \text{for } L(t) \leq t(d) \leq \mu(t) \\ F(t)_{\text{upper}} & \text{for } \mu(t) \leq t(d) \leq H(t) \\ 1 & \text{for } t(d) > H(t) \end{cases}$$

- The function F is chosen to stress readings close or beyond test limits

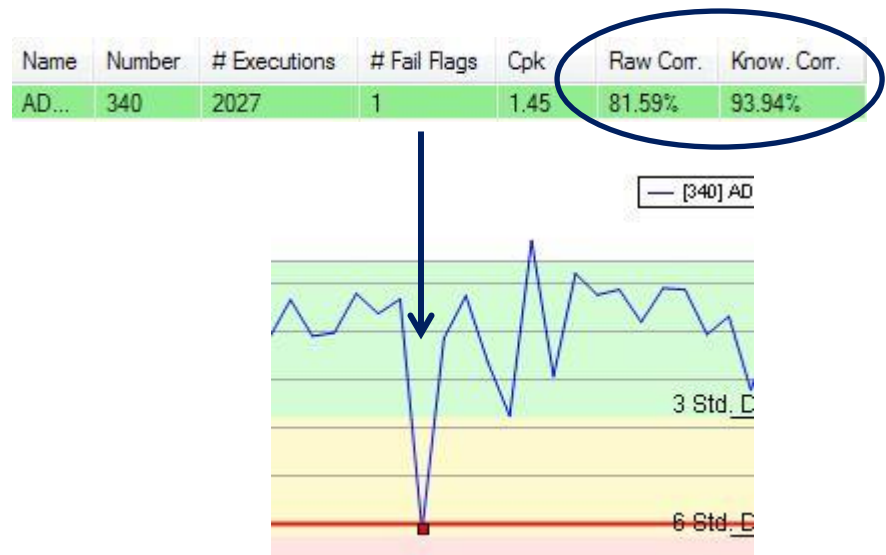


Knowledge Correlations (KCs) point the way

- Strategy: Find Signature Tests by reconstructing measurements in Adaptive Test Groups on a part-by-part scale from the best knowledge-correlated tests
- The reconstruction vehicle is a standard multilayer neural net (→FANN)
- KCs perform better than 'raw' correlations when it comes to spot fails! ('Raw' indicates original test readings.)

- Example:

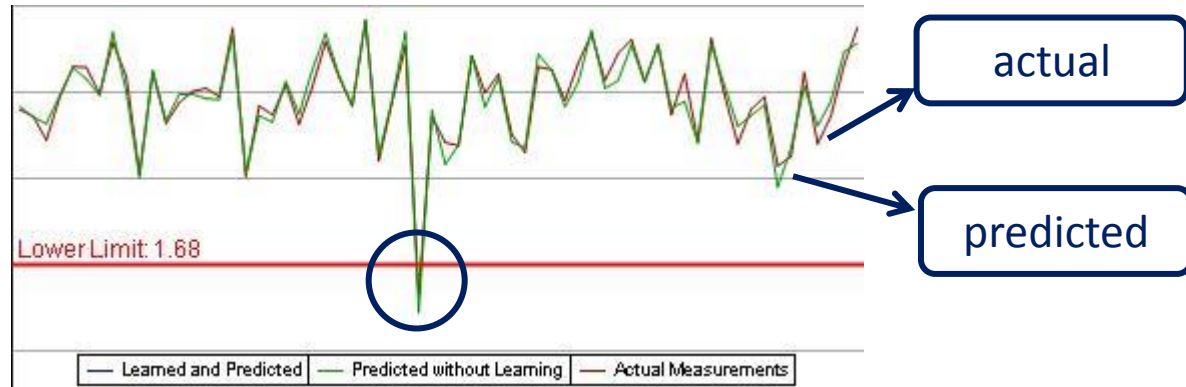
Reconstruct a Test #340 by a Test #300 over one wafer, in particular in the vicinity of the spot fail



Knowledge Correlations have the edge

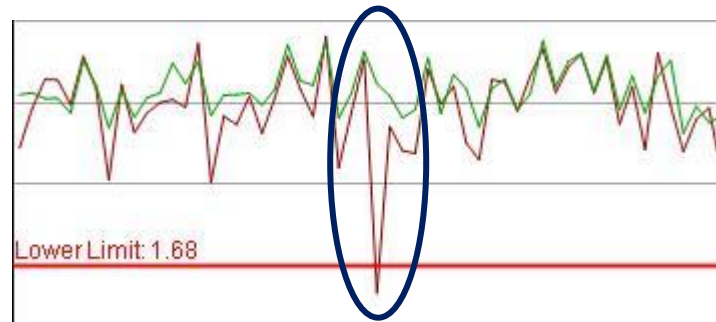
- Training of FANN over half the wafer not including spot fail

Test Reconstruction Partners		
Name	Number	Know. Corr.
VW...	300	93.94%
VCo...	388	66.87%
VW...	315	58.15%



- Same setting, but highest raw-correlated test as predictor:

Test Correlation Partners		
Name	Number	Raw Corr.
VCo...	388	81.59%
MV...	386	74.83%
VD...	384	61.84%



- If above is scalable → #300 is potential Signature Test

Na...	Number	# Executions	# Fail Flags
AD...	340	245827	53

Line #	Name	Number	Know. Corr.
<input type="checkbox"/> 1	VW...	300	77.71%
<input type="checkbox"/> 2	VCo...	388	53.45%
<input type="checkbox"/> 3	VW...	315	34.62%

Line #	Name	Number	Raw Corr.
1	hvk...	330	82.48%
2	DA...	419	69.57%
3	DA...	405	63.02%

The Decision Criterion

- Test #300 displays distinct spike at the part failing in Test #340, compared to its results just before



To qualify as a Numerical Indicator for future potential fails in #340, the graph suggests

- to evaluate the result of #300 at the fail device against a certain 'local average region' of #300
- to calculate that average from the #300-readings at a collection of parts tested just before the defective

The number of free parameters is just two:

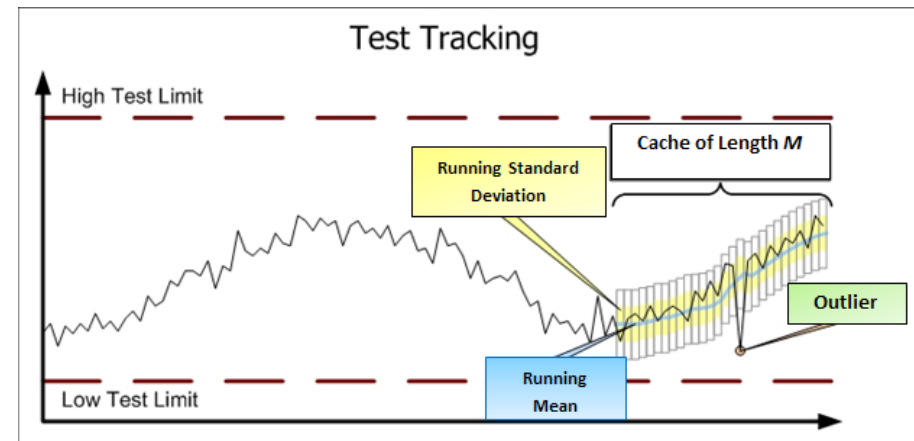
- The weight of the 'average' against the spike
- The size of the device window before the fail

→ Ockham's razor ...

Feed-Forward Decisions in Real Time

The algorithm is executed in a die-by-die rhythm:

- The most recent results, say $[1, N - 1]$, of Signature Tests are kept in memory
- Those readings determine the 'local average regions', against which the results of the Signature Tests at Part N are assessed
- Depending on the 'being inside or outside' scores of the latter, a potential fail in a defect-correlated Adaptive Test Group at the Part N is predicted
- On Device $N+1$ the procedure repeats itself, now with the set $[2, N]$ of the Signature Test results as entry of the 'local average region' calculation
- And it keeps on going that way...



The Adaptive Group Test Controller

The decision algorithm is the core of a Feed-Forward Controller, called Adaptive Group Test Controller (AGTC), where the latter

- generates the decision to execute Adaptive Groups or not on account of the Signature Test results received in real-time from the tester. Either uni- or multivariat, i.e., '∨' or '∧' combinations of Signature Tests
- hands back in real-time the instruction to execute Adaptive Groups or not to the tester
- however, does not get information about success (i.e., avoiding escapes)

NOTE

- No additional hardware is required to run the AGTC, only a minor one-time adaption of the test program
- The control over the Adaptive Group Testing remains with the owner of the test program → Interesting for fabless companies...

Actual Realisation of the AGTC

- On a Catalyst, for example, a test group can be easily handled inside an “if” statement (1=execute, 0=skip), where *oatcDo* is the automaton call:

```
if (oatcDo(oatc,340)==1)
    seq Adaptive Group()
    {
        TestNr(340)...
    }
```

- What about test time overhead due to the AGTC ?
- On a standard Catalyst, a automaton decision takes about 0.3 ms
- Typically, two to four Signature Tests control an Adaptive Group
- The method works independently of the test regime!

A Showcase Simulation

A Group 'ADC...' is taken as **Adaptive Test Group**:

- Test Time \cong 2 sec (app. 11% of total TT)
- 6 fails over 5 Wafer Sort lots (app.85k parts)

Two **Signature Tests** cover those fails:

	# Direct Hits	99[3]	596[3]	35[0]	35[0]	199[1]	199[1]
[10546]	4	0	1	1	0	1	1
[2003] IF	4	0	0	1	1	1	1
[10510]	3	1	1	0	1	0	0

Monitoring of #2003 and #10510 permits the dynamic sampling of 'ADC...' over another set of five lots



Start with controller parameters on 'save side':

Line	File	# Devices	Yield	# Esc's (ppm)	TTR [10738]	# Esc's [10738]	# Fails [10738]
60	sf...	700	83.14 %	0	10.86 %	0	0 / 8
102	sf...	691	85.53 %	0	11.58 %	0	0 / 7
62	sf...	692	85.69 %	0	12.72 %	0	0 / 7
59	sf...	685	85.11 %	0	8.18 %	0	0 / 7
66	sf...	692	83.09 %	0	10.40 %	0	0 / 5
53	sf...	685	87.45 %	0	11.68 %	0	0 / 5
61	sf...	685	85.99 %	0	15.77 %	0	0 / 3
64	sf...	693	82.11 %	0	13.85 %	0	0 / 3
100	sf...	691	87.70 %	0	15.05 %	0	0 / 3
101	sf...	691	84.80 %	0	9.84 %	0	0 / 3

app. 280 ms

Summary of Adaptive Test Simulation	
ATA Performance	
Adaptive Test Efficiency	100.00 %
Gain	
Global TTR Average	13.96 %
Loss	
Total Escapes (ppm)	0
Underlying Data	
Number of Devices analysed	83750

Is that Result good enough? More is in Store...

- Very little fails → Statistics is a somewhat 'lean'... Nevertheless, widening of controller parameters yields

Summary of Adaptive Test Simulation	
<input type="checkbox"/> ATA Performance Adaptive Test Efficiency 87.50 %	
<input type="checkbox"/> Gain Global TTR Average 44.87 %	
<input type="checkbox"/> Loss Total Escapes (ppm) 23.9	
<input type="checkbox"/> Underlying Data Number of Devices analysed 83750	

What about the two escapes?

Simulator details show:

WARNING!
 Defect devices escape in Adaptive Test Model!

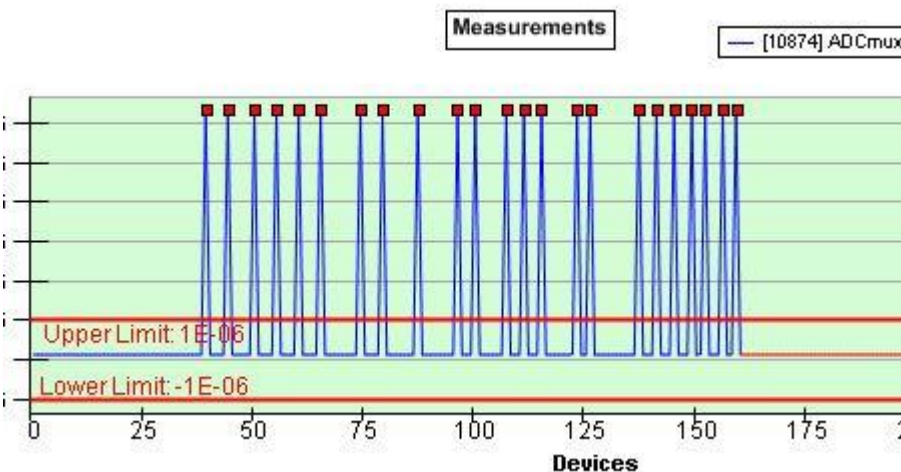
Test 10874 is not executed at
 Failing Device 625
 Failing Device 652

Slightly more risk:
 app. 900 ms

- A 'local single wafer, single site' problem:



→ New analysis, add one Signature Test...o.k.!



Summary of Adaptive Test Simulation	
<input type="checkbox"/> ATA Performance Adaptive Test Efficiency 100.00 %	
<input type="checkbox"/> Gain Global TTR Average 34.14 %	
<input type="checkbox"/> Loss Total Escapes (ppm) 0	
<input type="checkbox"/> Underlying Data Number of Devices analysed 83750	