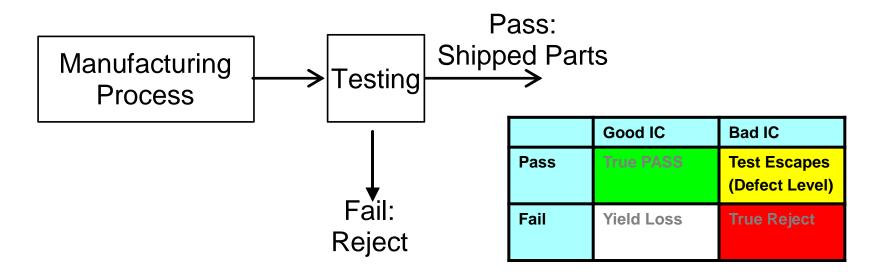
#### Introduction

- What Is Testing
- Types of Testing
- Test Quality
  - Defect per Million (DPM)
  - Models to Predict DPM
  - Models to Predict Yield
- Test Economics
- Issues in Testing
- Conclusion



## **How to Measure Test Quality?**

- Defect Level (DL)
  - Fraction of bad IC passing the test (test escapes)
  - Measured by Defective Parts per Million (DPM)
    - \* < 200 DPM is acceptable for some IC</p>
    - \* >1,000 DPM is very bad for IC



**Low DPM Means Good Test Quality** 

## IC DPM → System DPM

- If you think DPM= 1,000 (0.1%) is good enough, consider this case
- A system has twenty IC (each 1000 DPM), what is system DPM?
  - No IC is defective for system to work correctly
    - \* System DPM =  $1-(1-0.1\%)^{20} = 2\% = 20,000$  DPM!
  - If you sell a million system
    - You have to fix 20,000 of them!

**Low IC DPM is Very Important** 

### Quiz

Q: (Cont'd from last slide)
20 IC in a system.
Suppose we improve IC test quality to 200DPM
What is system DPM?

A:

IC DPM ↓ 5X, System DPM ↓ 5X

#### Introduction

- What Is Testing
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Measure *Real DPM* Is difficult. Can we predict it?



#### Brown & Williams Model [Williams 81]

$$DL = 1 - Y^{(1-FC)}$$

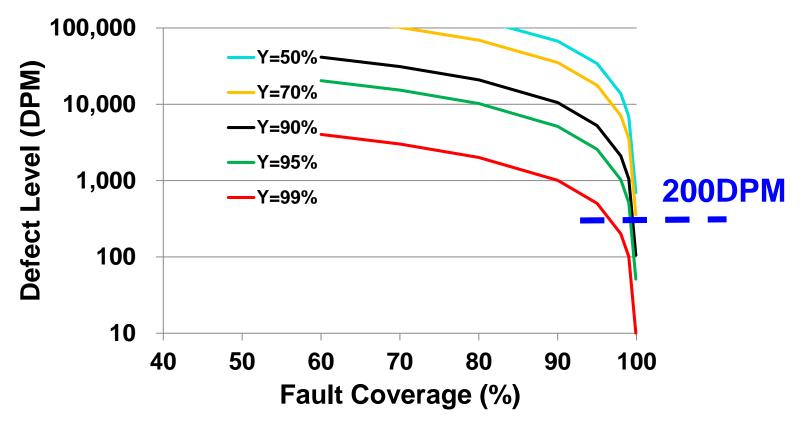
- Y=Yield, fraction of total manufactured IC that are good
  - High Y means good manufacturing quality
  - Unknown parameter, estimated by fab. data or prediction

$$yield, Y = \frac{number\ of\ good\ chips}{number\ of\ total\ chips} \le 100\%$$

- FC = Fault Coverage, fraction of detected faults
  - High FC means good test quality
  - Known parameter, from fault simulator (see fault simulation)

$$fault\ coverage,\ FC = \frac{number\ of\ detected\ faults}{number\ of\ total\ faults} \leq 100\%$$

### DL v.s. FC



Required DL = 200 DPM, then we need

Yield	50%	70%	90%	95%	99%
FC	99.97%	99.95%	99.8%	99.6%	98%

\* DL  $\approx$  (1-FC)(1-Y) when Y  $\approx$  1

poor yield good test



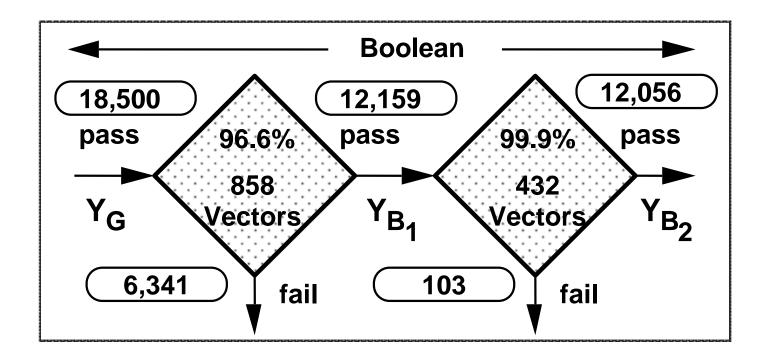
good yield poor test

### **B&W Proof** (simplified!)

- Assume each fault occurrence probability is q (uniform independent)
  - Total t faults in an IC
  - d out of t faults detected by this test  $\rightarrow$  FC = d/t
- Event A = an IC is good (no fault at all)
  - $Prob(A) = probability that an IC is good <math>\rightarrow Prob(A) = (1-q)^t$
  - Y = fraction of IC that are good  $\rightarrow Y = Prob(A)$
- Event B = an IC pass the test
  - Prob(B) = an IC free of d faults  $\rightarrow Prob(B) = (1-q)^d$
- Event AB = an IC is good and pass test
  - Assume no overkill  $\rightarrow Prob(AB) = Prob(A)$
- Quality Level (QL) = fraction of good IC passing the test
  - $QL = Prob(A/B) = Prob(AB)/Prob(B) = (1-q)^{t-d} = (1-q)^{t(1-d/t)} = Y^{(1-FC)}$
- Defect Level (DL) = fraction of bad IC passing the test
  - $DL = 1-QL = 1-Y^{(1-FC)}$  QED

## Motorola 6802 Experiment [McCluskey 88]

- Y = 12,056/18,500 = 65.16%
- test #1 FC = 96.6%
- Theoretical DL = 1-Y<sup>(1-FC)</sup> = 14,454 DPM
- Experimental DL = 8,471 DPM (103 over 12,159 IC passing test #1)



### Quiz

Q1: Use B&W model.

$$DL = 1 - Y^{(1-FC)}$$

A:

Q2: Your boss is not happy about DL. If we add some design for testability (DFT) circuits to improves FC but harms yield a little.

Y=97%, FC=99%. DL=?

A:

**DFT Very Important for DL Improvement** 

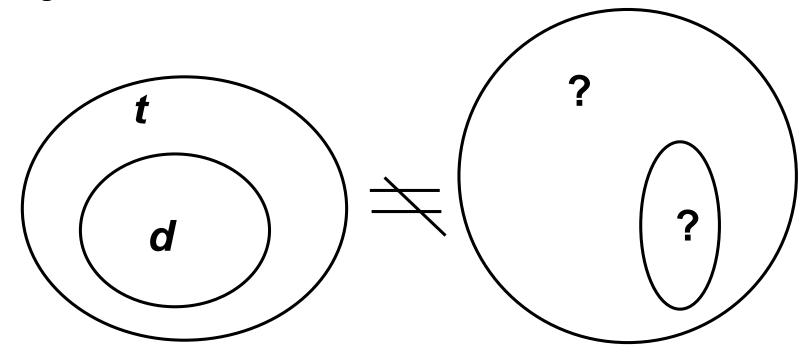
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- Conclusion



### Why B&W Not Accurate?

- 1. Fault coverage does not represent defect coverage
  - 100% FC ≠ 0 DPM (see fault modeling chapter)
- 2. Defects are clustered
  - Agrawal used Poisson distribution



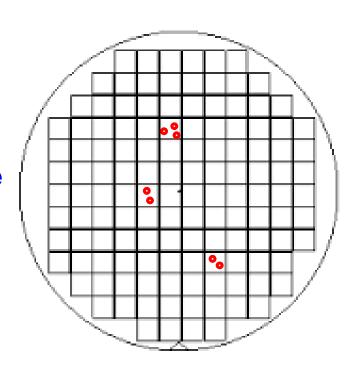
**Faults** 

**Defects** 

### Agrawal Model [Agrawal 82]

$$DL = \frac{(1 - FC)(1 - Y)e^{-(n-1)FC}}{Y + (1 - FC)(1 - Y)e^{-(n-1)FC}}$$

- Actual data show defects are clustered
  - *n* =average num. of defects on a bad die
- We need two unknown parameters
  - n: from experiment
  - Y: from fab. or prediction
- Motorola experiment revisited
  - Experiment DL = 8,471 DPM
  - B&W DL =14,454 DPM
  - Agrawal
    - \* *n* =1, DL = 17,849 DPM
    - \* n = 2, DL = 6,869 DPM
    - experimental n between 1 and 2



7 defects 3 bad die n=7/3=2.3

#### Introduction

- What Is Testing
- Types of Testing
- Test Quality
  - Defect per Million (DPM)
  - Models to Predict DPM
  - Models for Yield
    - \* Simple
    - Critical Area
- Test Economy
- Issues in Testing
- Conclusion

Measure *Real Yield* Is too late. Can we predict it?



# **Yield Estimation (Simple)**

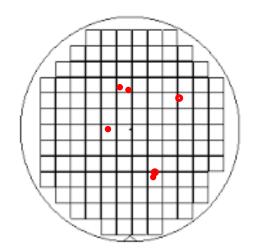
- Number of defects k on a die obeys *Poisson distribution* with mean  $\lambda$ 
  - Prob(k) = prob exactly k defects on a die
  - A die is good when k=0
  - Yield = Prob(k=0)

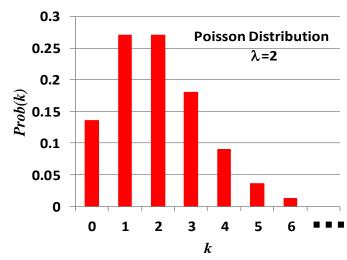
$$Prob(k) = \frac{\lambda^k e^{-\lambda}}{k!}$$

$$Y = Prob(k = 0) = e^{-\lambda}$$

- A = die area, D = defect density
- $\lambda = AD$  = ave. number of defects per die

$$Y=e^{-AD}$$
 Largerarea,lower yield





### Quiz

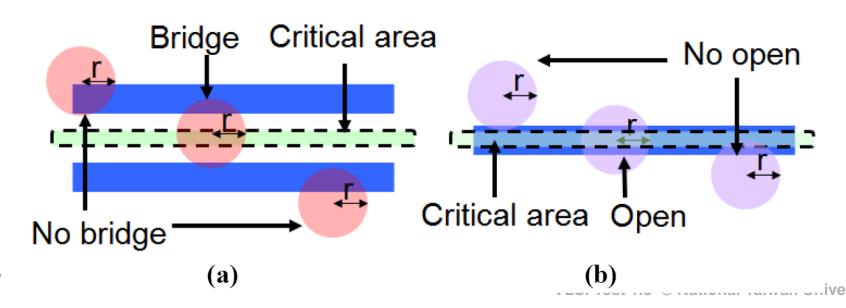
Q: A=0.8cm<sup>2</sup>, D=0.5/cm<sup>2</sup>. What is yield?

$$Y = e^{-AD}$$

A:

### Critical Area, CA

- Problem with Y-AD model
  - Some area NOT sensitive to defect. Use total area is pessimistic
- Critical area, CA(r)
  - Area where center of defect (radius r) land to create a fault
- Example:
  - (a) bridging fault (b) open fault

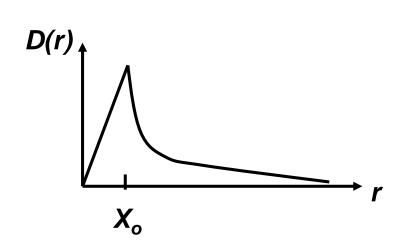


### Defect Size and Density Distribution, DSDD

- D(r) = defect density per unit defect radius r
- Typical DSDD curve:

$$D(r) = \begin{cases} D_0 \frac{2(p-1)r}{(p+1)X_o^2}, & 0 \le r < X_o \\ D_0 \frac{2(p-1)X_o^{p-1}}{(p+1)r^p}, & r \ge X_o \end{cases}$$

- D(r) is process dependent.
  - Example values:
  - $4.5 \times 10^{-6}$  cm  $\leq X_0 \leq 5.5 \times 10^{-6}$  cm
  - 2.0 ≤ p ≤ 3.0
  - $0.25 \text{ /cm}^2 \le D_0 \le 0.6 \text{ /cm}^2$



Total Defect Density = 
$$D_0 = \int_0^\infty D(r) dr$$

## **Yield Estimation (CA)**

Expected number of defects per die landing in CA

$$\lambda = \int_0^\infty CA(r) \times D(r) dr$$

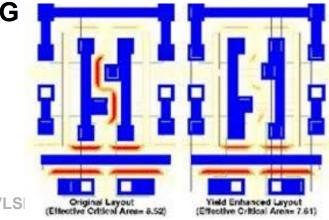
- Again, *Poisson distribution* with mean  $\lambda$ 
  - Yield = Prob(k=0), zero defect in CA

$$Y = e^{-\lambda} = e^{-\int_0^\infty CA(r) \times D(r) dr}$$

simple model

$$\Leftrightarrow Y = e^{-AD}$$

- Critical area analysis (CAA) is useful tool to estimate yield
  - 1. modify design for yield improvement
  - 2. extract bridging or open faults for ATPG



## **Summary**

- Test quality measured by Defect per Million (DPM)
- Models for DPM
  - Brown & Williams :

$$DL = 1 - Y^{(1-FC)}$$

Agrawal

$$DL = \frac{(1 - FC)(1 - Y)e^{-(n-1)FC}}{Y + (1 - FC)(1 - Y)e^{-(n-1)FC}}$$

- Models for Yield
  - Simple model

$$Y = e^{-AD}$$

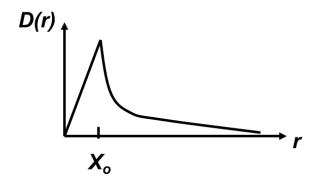
Critical area

$$Y = e^{-\int_0^\infty CA(r) \times D(r) dr}$$



# Food for Thoughts, FFT

- Q: What is unit of *D(r)*?
  - D(r) = defect density per unit defect radius r



Total Defect Density =  $D_0 = \int_0^\infty D(r) dr$ 



# **APPENDIX#1: Agrawal Proof (1)**

• Number of defects k on a bad IC obeys Poisson distribution with mean n p(k) = probability exactly k faults in a bad IC (k > 0)

$$p(k) = (1-Y)\frac{(n-1)^{k-1}e^{-(n-1)}}{(k-1)!} \qquad k = 1, 2, 3... \text{ (shifted Poisson)}$$

$$p(0) = Y$$

- Suppose total t defects, d is detected by test, so FC = d/t
- $q_m(k)$  = prob. detecting exactly m defects when an IC has k defects

$$q_{m}(k) = \frac{\binom{k}{m}\binom{t-k}{d-m}}{\binom{t}{d}} \quad hypergeometric distribution$$

•  $q_0(k)$  = prob. of passing a bad IC, having exactly k defects

$$q_0(k) = \frac{\binom{t-k}{d}}{\binom{t}{d}} \approx (1 - \frac{d}{t})^k = (1 - FC)^k$$

# **Agrawal Proof (2)**

TE = Prob. bad IC passing the test

$$TE = \sum_{k=1}^{t} q_0(k) p(k) = \sum_{k=1}^{t} (1 - FC)^k (1 - Y) \frac{(n-1)^{k-1}}{(k-1)!} e^{-(n-1)}$$

$$= (1 - Y) \sum_{k=1}^{t} (1 - FC) \frac{\left[ (1 - FC)(n-1) \right]^{k-1}}{(k-1)!} e^{-(n-1)(1 - FC)} e^{-(n-1)FC}$$

$$= (1 - Y)(1 - FC) e^{-(n-1)FC} \sum_{k=1}^{t} \frac{\left[ (1 - FC)(n-1) \right]^{k-1}}{(k-1)!} e^{-(n-1)(1 - FC)}$$

$$\approx (1 - Y)(1 - FC) e^{-(n-1)FC}$$

$$DL = \frac{TE}{Y + TE} = \frac{(1 - FC)(1 - Y)e^{-(n-1)FC}}{Y + (1 - FC)(1 - Y)e^{-(n-1)FC}}$$

## **APPDENDIX #2: B&W Approximation**

Tylor Exapnsion

$$DL(Y) = \sum_{n=0}^{\infty} \frac{f^{(n)}(Y)}{n!} (Y - a)^n = f(1) + \frac{f'(1)}{1!} (Y - 1) + \frac{f''(1)}{2!} (Y - 1)^2 + \dots$$

when  $Y \approx 1$ , we expand DL(Y) around a=1

$$DL(Y) = 1 - Y^{1-FC}$$

$$= (1 - 1^{1-FC}) - (1 - FC) \cdot 1^{-FC} (Y - 1) - \frac{(-FC)(1 - FC)1^{-FC-1}}{2!} (Y - 1)^2 - \dots$$

$$= 0 + (1 - FC)(1 - Y) + \frac{(1 - FC)FC}{2} (1 - Y)^2 + \dots$$

$$\approx (1 - FC)(1 - Y)$$

Example FC=98.5%, Y=95% 
$$DL(Y) = 1 - Y^{1-FC} = 1 - 0.95^{0.015} = 769 \text{DPM}$$
 
$$DL(Y) \approx (1 - FC)(1 - Y) = 750 \text{DPM}$$
 
$$DL(Y) \approx (1 - FC)(1 - Y) + \frac{FC(1 - FC)}{2}(1 - Y)^2 = 768 \text{DPM}$$

**DL** ≈ (1-FC)(1-Y)