

# Towards a Reliable Statistical Oracle and its Applications

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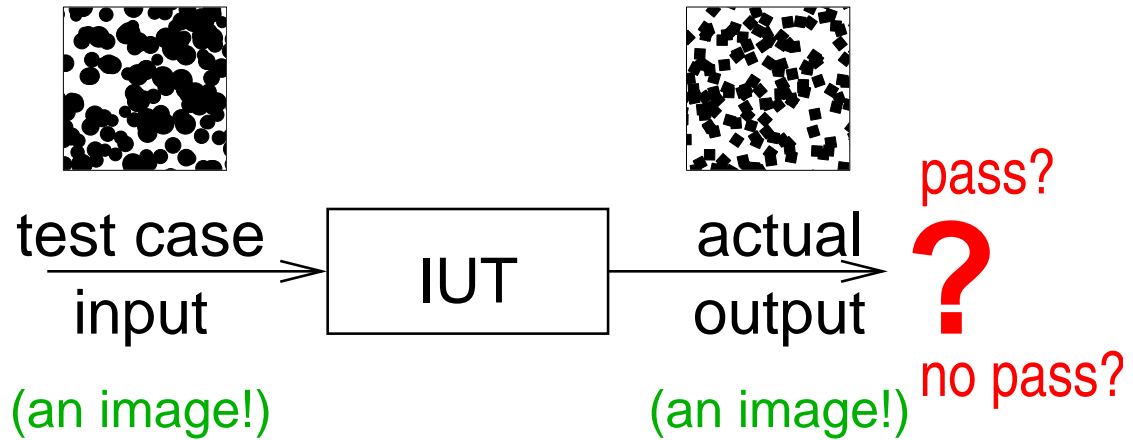
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Universität Ulm

## Overview

1. Motivation: How to test image processing applications?
2. What is an oracle?
3. Typical examples of oracles
4. What about random testing?
5. **A pattern for the Statistical Oracle**
6. Necessary image processing basics
7. **How to apply the Statistical Oracle to test image processing applications?**
8. Conclusion

## Motivation

*How to test an image processing application?*

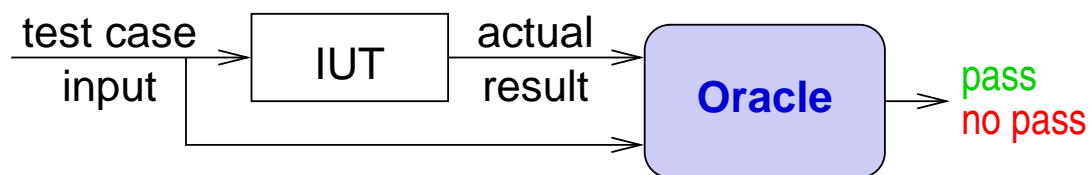


- Complex input and output
- How to determine the expected output?
- Also: How to generate many complex inputs?

## Oracles

*What is an oracle?*

### Overview:



An **oracle** is necessary to **decide** whether a test case passes or not.

# Oracles

## Examples of Oracles

- **Judging:**  
Tester evaluates pass/no pass
- **Prespecification:**  
Solved examples, approximation, parametric, ...
- **Gold Standard:**  
Trusted system, voting, regression, ...
- **Organic:**  
Reversing, built-in test, ...

~> all oracles only applicable in special cases

# Random Testing

**Random Testing** is testing using **randomly generated inputs**.

**Complex inputs** (e. g. images) are not easy to specify

~> **random generation** would pay off

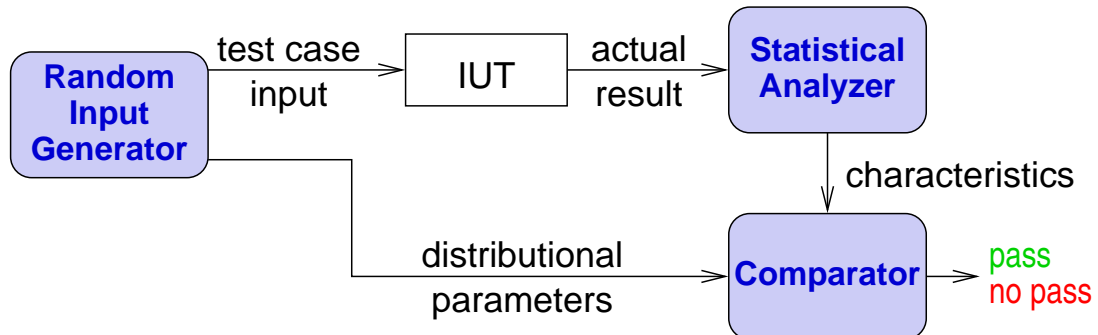
But what about the **outputs**?

~> An appropriate **oracle** is needed for **automation**

# Statistical Oracle

## Pattern

### Overview:



### Necessary:

**Mean, Variance**, and so on of **characteristics** must be known (explicit formula or algorithm)

# Statistical Oracle

## Necessary Statistical Basics (1)

### Notation:

- Let  $X_1, \dots, X_n$  denote i. i. d. **random variables** (the **inputs of the comparator** –  $X_i$  belongs to the  $i$  th test case)
- The mean  $\mu$  and variance  $\sigma^2$  of  $X_i$  is **unknown**.
- The **sample mean** of these random variables  $X_1, \dots, X_n$  is

$$\bar{X}_n := \frac{1}{n} \sum_{i=1}^n X_i.$$

## Statistical Oracle

### Necessary Statistical Basics (2)

#### Notation (continued):

- According to the **central limit theorem**, it holds that

$$\frac{\overline{X}_n - \mu}{\sigma/\sqrt{n}} \xrightarrow{d} \mathcal{N}(0, 1)$$

for  $n \rightarrow \infty$ .

- Thus,  $\overline{X}_n$  can be regarded as **approximately normally distributed** with mean  $\mu$  and variance  $\sigma^2/n$  if  $n \geq 30$  (a common rule of thumb)
- The **greater**  $n$  gets, the **less likely deviations from  $\mu$**  become

## Statistical Oracle

### Necessary Statistical Basics (3)

#### Notation (continued):

- The **sample variance**

$$S_n^2 := \frac{1}{n-1} \sum_{i=1}^n (X_i - \overline{X}_n)^2$$

of the random variables  $X_1, \dots, X_n$  (**approaches  $\sigma^2$**  as  $n$  goes to infinity)

- $x_i, \overline{x}_n, s_n^2$  denote **realizations**
- $\mu_0$  denotes the **expected mean** of  $X_i$

# Statistical Oracle

## Statistical Decision Test

Given the (maximum) probability of  $\alpha \in (0, \frac{1}{2})$  of a Type I error (i. e.  $\alpha$  can be **controlled**)

**Type I error:** reject  $H_0$  under  $H_0$

**Type II error:**  $H_0$  not rejected under  $H_1$

**Construction:**

**Intersection-union test** with two one-sided t-tests

$H_0 : \mu \notin [\mu_0 - \Delta, \mu_0 + \Delta]$  (buggy)

$H_1 : \mu = \mu_0$  (correct)

( $\Delta > 0$  can be chosen arbitrarily)

**Type I error:** false rejection of  $H_0$  (i. e. a buggy IUT passes)

# Statistical Oracle

## Statistical Decision Test: Decision Rule

Reject  $H_0$ , i. e. the **implementation passes**, if

$$\frac{\bar{x}_n - (\mu_0 - \Delta)}{s_n/\sqrt{n}} \geq t_{n-1,\alpha}$$

and

$$\frac{\bar{x}_n - (\mu_0 + \Delta)}{s_n/\sqrt{n}} \leq -t_{n-1,\alpha}$$

where  $t_{n-1,\alpha}$  denotes the  $(1 - \alpha)$ -quantile of the t-distribution with  $n - 1$  degrees of freedom

## Statistical Oracle

### Statistical Decision Test: Type II Error

- Probability  $\alpha$  of **Type I error** already controlled
- Probability  $\beta$  of **Type II error**:

$$\begin{aligned} & \mathbb{P}(\text{decision for } H_0 \mid H_1) \\ &= \mathbb{P}\left(t < t_{n-1,\alpha} - \frac{\Delta}{s_n/\sqrt{n}} \vee t > -t_{n-1,\alpha} + \frac{\Delta}{s_n/\sqrt{n}} \mid \mu = \mu_0\right) \end{aligned}$$

with  $t = \frac{\bar{x}_n - \mu_0}{s_n/\sqrt{n}}$

- **Consequences:**

- ◊ **Necessary:**  $\frac{\Delta}{s_n/\sqrt{n}} > t_{n-1,\alpha}$
- ◊  $\beta$  decreases, when
  - $\alpha$ ,  $\Delta$ , or  $n$  increases
  - $\sigma$  decreases (and consequently  $s_n$  decreases)

## Statistical Oracle

### Statistical Decision Test: Type II Error (2)

#### Solution:

- For  $\alpha = 1/4$  determine  $n$  such that  $\beta \leq 1/4$  ( $\Delta$  and  $\sigma$  are given)  
( $\rightsquigarrow$  numerically!)
- **BPP** algorithm: each answer is correct with probability at least  $3/4$
- **Repeat this  $m$  times** and make a **majority decision** to achieve arbitrarily small error probabilities

# Statistical Oracle

## Consequences

- **Cannot decide** for a **single test case**, but for a set of test cases
- **Decision** is not always correct, but with arbitrarily high probability
- Does not check the **whole result**, but **only some properties**  
( $\rightsquigarrow$  only correctness with respect to the checked property and with respect to the used inputs)

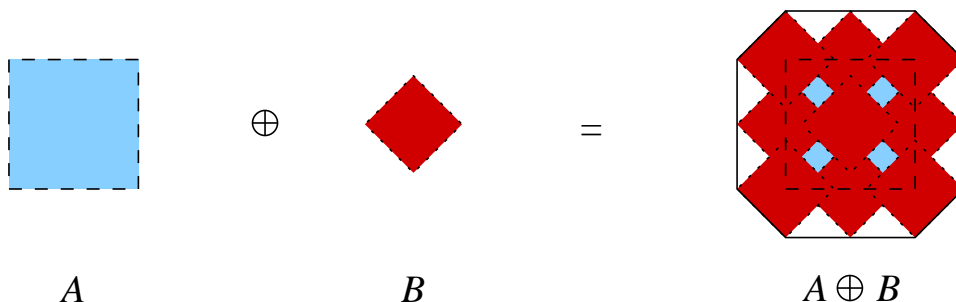
# Image Processing

## Preliminaries: Minkowski addition and dilation

The **Minkowski addition**  $A \oplus B$  is defined as

$$A \oplus B := \{x + y : x \in A, y \in B\}.$$

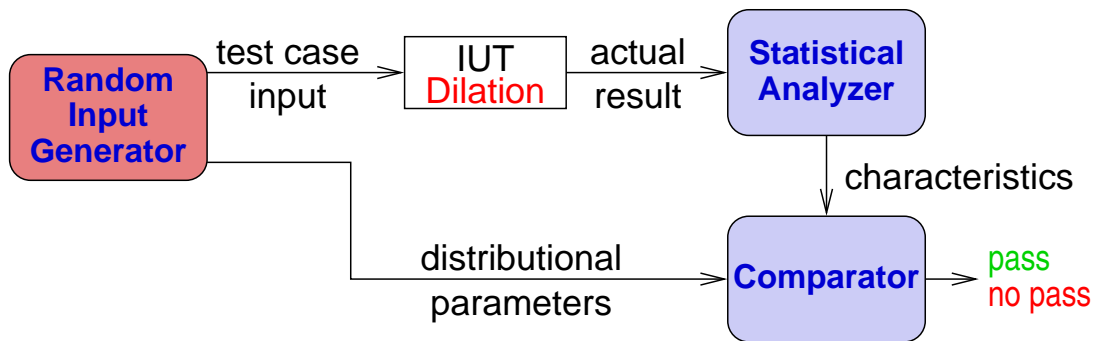
The set  $B$  is in this case called the **structuring element**.



$\delta_B(A)$  denotes the **dilation** of  $A$  with the structuring element  $B$  and is defined as  $\delta_B(A) := A \oplus \check{B}$

# Testing of Dilation

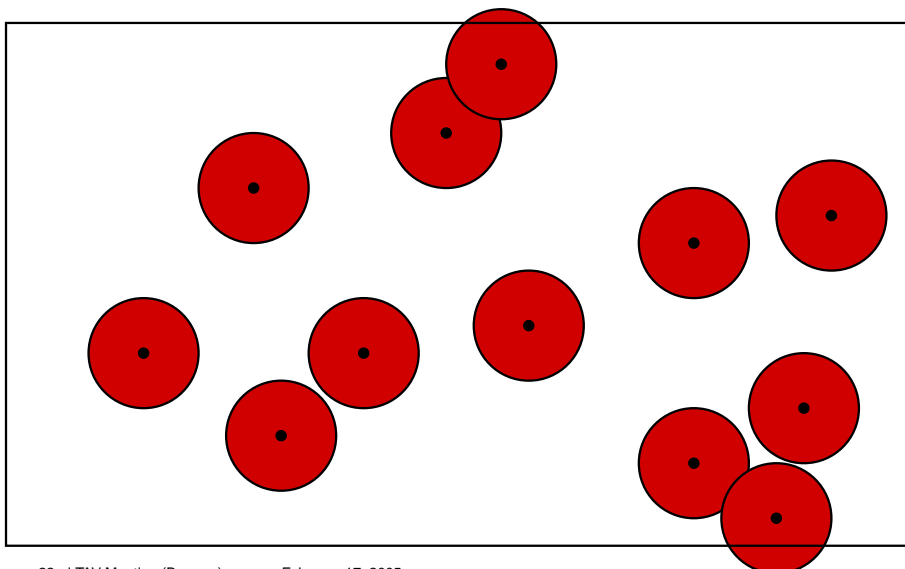
## Random Input Generator: Overview



# Testing of Dilation

## Random Input Generator: Boolean Model

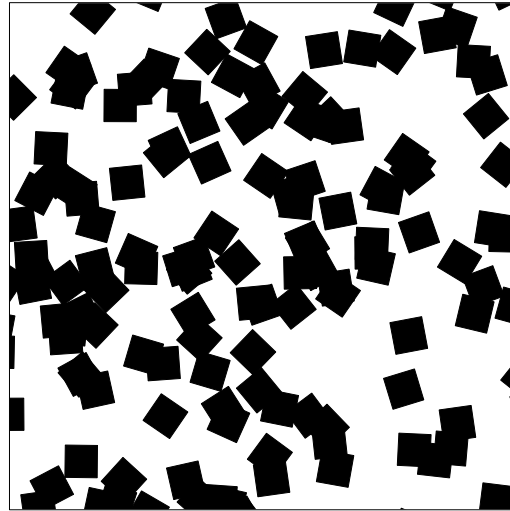
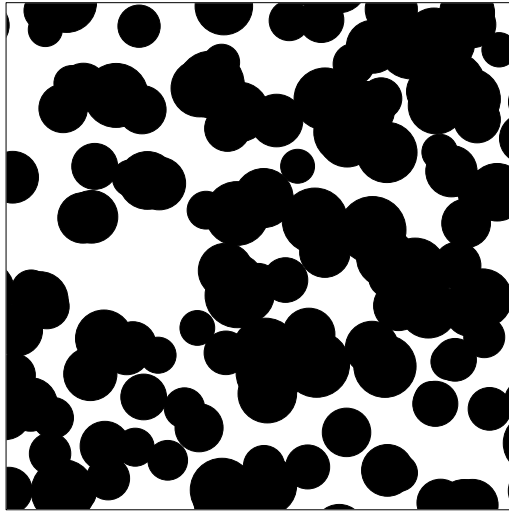
The **Boolean model** is simply the union  $\bigcup_i (B_i + X_i)$  of i. i. d. **random grains**  $B_i$  each translated into another point  $X_i$  of the underlying **Poisson process**.



# Testing of Dilation

## Boolean Model: Examples

Examples:



# Testing of Dilation

## Boolean Model: Useful Property

What happens, when the **Boolean model**  $\bigcup_i (B_i + X_i)$  is **dilated by  $B$** ?

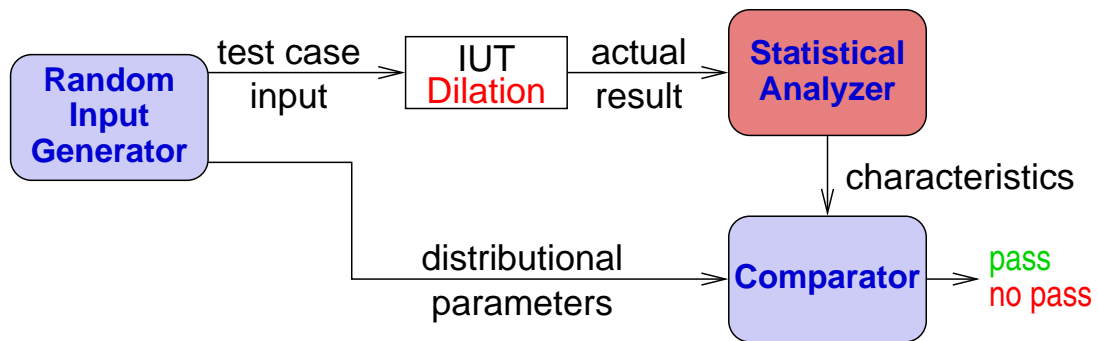
$$\begin{aligned}\delta_B \left( \bigcup_i (B_i + X_i) \right) &= \left( \bigcup_i B_i + X_i \right) \oplus \check{B} \\ &= \bigcup_i ((B_i + X_i) \oplus \check{B}) = \bigcup_i (B_i \oplus \check{B}) + X_i,\end{aligned}$$

The **result is the Boolean model** with grains  $B_i \oplus \check{B}$ .

For example, if  $B_i$  is a disc with random radius  $R_i$  and  $B$  is a disc with radius  $r$ ,  $B_i \oplus \check{B}$  is simply a disc with random radius  $R_i + r$ .

# Testing of Dilation

## Characteristics: Overview



# Testing of Dilation

## Characteristics: Formulae

Choosing **specific area**  $A_A$ , **specific boundary length**  $L_A$ , and **specific Euler number**  $\chi_A$

~> explicit formulae are known:

$$A_A = 1 - \exp(-\lambda \bar{A})$$

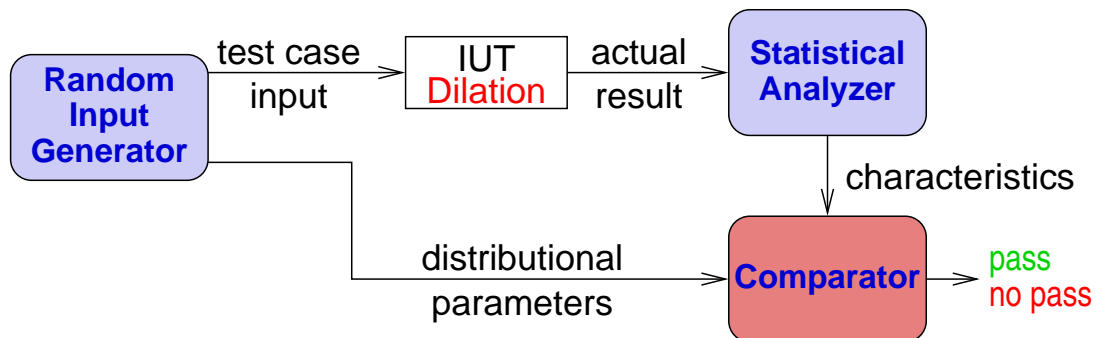
$$L_A = \lambda \bar{L} \exp(-\lambda \bar{A})$$

$$\chi_A = \lambda \left( 1 - \frac{\lambda \bar{L}^2}{4\pi} \right) \exp(-\lambda \bar{A})$$

- $\bar{A}$  **mean area** of the primary grain
- $\bar{L}$  **mean boundary length** of the primary grain
- $\lambda$  **intensity** of the Poisson process

# Testing of Dilation

## Comparator



The comparison is done with the presented **intersection-union test**

# Testing of Dilation

## Putting the Pieces together

1. The **Random Test Input Generator** generates realizations of the **Boolean model**. Passes the characteristics of the Poisson process ( $\lambda$ ) and of the primary grain ( $\bar{A}$  and  $\bar{L}$ ) to the Comparator.
2. The **Statistical Analyzer** computes the **estimators for  $A_A$ ,  $L_A$  and  $\chi_A$** . Each such estimator can be modeled as a random variable  $X_i$ . Then  $A_A$ ,  $L_A$  and  $\chi_A$  are the expected means, respectively, of these random variables (i. e.  $\mu_0$ ).
3. The **Comparator** accumulates the realizations  $x_i$  of each such random variable  $X_i$  for  $n$  outputs and computes the realization  $\bar{x}_i$  of the sample mean. Finally, it decides using an **intersection-union test** — as described — whether the IUT passes or not. (The **error probability  $\alpha$**  can also be given to the Comparator.)

## Conclusion

- **Statistical Oracle** for random testing if mean etc. is known
- Based on an **intersection-union decision test**
- **Error probabilities** can be chosen as needed through repetition
- Random testing of **image processing applications** pays off, due to complex inputs
- **Boolean model** can be used to test **dilation**
- **Relevant characteristics**: area, boundary length, Euler number

## Perspectives

- **Duality** between dilation and erosion resp. Minkowski addition and Minkowski subtraction  
~> Erosion can also be tested
- **Distance transform** usually used to compute dilation and erosion  
~> Distance transform can be tested indirectly

## References

- Robert V. Binder: *Testing Object-Oriented Systems*. Addison-Wesley, 1999.
- Johannes Mayer, Ralph Guderlei: *Test Oracles Using Statistical Methods*. In: Lecture Notes in Informatics P-58, Gesellschaft für Informatik, Köllen Druck+Verlag GmbH, 2004, pp. 179–189.
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