

# Threshold Decoding for Disjunctive Group Testing<sup>1</sup>

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**Abstract.** Let  $1 \leq s < t$ ,  $N \geq 1$  be integers and a complex electronic circuit of size  $t$  is said to be an  $s$ -active,  $s \ll t$ , and can work as a system block if not more than  $s$  elements of the circuit are defective. Otherwise, the circuit is said to be an  $s$ -defective and should be replaced by a similar  $s$ -active circuit. Suppose that there exists a possibility to run  $N$  non-adaptive group tests to check the  $s$ -activity of the circuit. As usual, we say that a (disjunctive) group test yields the positive response if the group contains at least one defective element. Along with the conventional decoding algorithm based on disjunctive  $s$ -codes, we consider a threshold decision rule with the minimal possible decoding complexity, which is based on the simple comparison of a fixed threshold  $T$ ,  $1 \leq T \leq N - 1$ , with the number of positive responses  $p$ ,  $0 \leq p \leq N$ . For the both of decoding algorithms we discuss upper bounds on the  $\alpha$ -level of significance of the statistical test for the null hypothesis  $\{H_0 : \text{the circuit is } s\text{-active}\}$  verse the alternative hypothesis  $\{H_1 : \text{the circuit is } s\text{-defective}\}$ .

## 1 Statement of Problem

Let  $N \geq 2$ ,  $t \geq 2$ ,  $s$  and  $T$  be integers, where  $1 \leq s < t$  and  $1 \leq T < N$ . The symbol  $\triangleq$  denote the equality by definition,  $|A|$  – the size of the set  $A$  and  $[N] \triangleq \{1, 2, \dots, N\}$  – the set of integers from 1 to  $N$ . A binary  $(N \times t)$ -matrix

$$X = \|x_i(j)\|, \quad x_i(j) = 0, 1, \quad i \in [N], j \in [t], \quad (1)$$

with  $t$  columns (*codewords*)  $\mathbf{x}(j) \triangleq (x_1(j), x_2(j), \dots, x_N(j))$ ,  $j \in [t]$ , and  $N$  rows  $\mathbf{x}_i \triangleq (x_i(1), x_i(2), \dots, x_i(t))$ ,  $i \in [N]$ , is called a *binary code of length  $N$  and size  $t$*  =  $\lfloor 2^{RN} \rfloor$ , where a fixed parameter  $R > 0$  is called a *rate* of the code  $X$ . The number of 1's in a binary column  $\mathbf{x} = (x_1, \dots, x_N) \in \{0, 1\}^N$ , i.e.,  $|\mathbf{x}| \triangleq \sum_{i=1}^N x_i$ , is called a *weight* of  $\mathbf{x}$ . A code  $X$  is called a *constant weight binary code of weight  $w$* ,  $1 \leq w < N$ , if for any  $j \in [t]$ , the weight  $|\mathbf{x}(j)| = w$ . The conventional symbol  $\mathbf{u} \vee \mathbf{v}$  will be used to denote the disjunctive (Boolean)

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<sup>1</sup>The research is supported in part by the Russian Foundation for Basic Research under Grant No. 16-01-00440.

sum of binary columns  $\mathbf{u}, \mathbf{v} \in \{0, 1\}^N$ . We say that a column  $\mathbf{u}$  *covers* a column  $\mathbf{v}$  ( $\mathbf{u} \succeq \mathbf{v}$ ) if  $\mathbf{u} \vee \mathbf{v} = \mathbf{u}$ .

## 1.1 Disjunctive and Threshold Disjunctive Codes

**Definition 1.** [1]. A code  $X$  (1) is called a *disjunctive  $s$ -code*,  $s \in [t - 1]$ , if the disjunctive sum of any  $s$ -subset of codewords of  $X$  covers those and only those codewords of  $X$  which are the terms of the given disjunctive sum.

Let  $\mathcal{S}$ ,  $\mathcal{S} \subset [t]$ , be an arbitrary fixed collection of defective elements of size  $|\mathcal{S}|$ . For a binary code  $X$  and collection  $\mathcal{S}$ , define the binary *response vector* of length  $N$ , namely:

$$\mathbf{x}(\mathcal{S}) \triangleq \bigvee_{j \in \mathcal{S}} \mathbf{x}(j), \quad \text{if } \mathcal{S} \neq \emptyset \quad \text{and} \quad \mathbf{x}(\mathcal{S}) \triangleq (0, 0, \dots, 0) \quad \text{if } \mathcal{S} = \emptyset. \quad (2)$$

In the classical problem of *non-adaptive group testing*, we describe  $N$  tests as a binary  $(N \times t)$ -matrix  $X = \|\mathbf{x}_i(j)\|$ , where a column  $\mathbf{x}(j)$  corresponds to the  $j$ -th element, a row  $\mathbf{x}_i$  corresponds to the  $i$ -th test and  $x_i(j) \triangleq 1$  if and only if the  $j$ -th element is included into the  $i$ -th testing group. The result of each test equals 1 if at least one defective element is included into the testing group and 0 otherwise, so the column of results is exactly equal to the response vector  $\mathbf{x}(\mathcal{S})$ . Definition 1 of disjunctive  $s$ -code  $X$  gives the important sufficient condition for the evident identification of any unknown collection of defective elements  $\mathcal{S}$  if the number of defective elements  $|\mathcal{S}| \leq s$ . In this case, the identification of the unknown  $\mathcal{S}$  is equivalent to discovery of all codewords of code  $X$  covered by  $\mathbf{x}(\mathcal{S})$ , and its complexity is equal to the code size  $t$ . Note that this algorithm also allows us to check  $s$ -activity of the circuit defined in the abstract. Moreover, it is easy to prove by contradiction that every code  $X$  which allows to check  $s$ -activity of the circuit without error is a disjunctive  $s$ -code.

**Definition 2.** Let  $s, s \in [t - 1]$ , and  $T, T \in [N - 1]$ , be arbitrary fixed integers. A disjunctive  $s$ -code  $X$  of length  $N$  and size  $t$  is said to be a *disjunctive  $s$ -code with threshold  $T$*  (or, briefly,  $s^T$ -code) if the disjunctive sum of any  $\leq s$  codewords of  $X$  has weight  $\leq T$  and the disjunctive sum of any  $\geq s + 1$  codewords of  $X$  has weight  $\geq T + 1$ .

Obviously, for any  $s$  and  $T$ , the definition of  $s^T$ -code gives a sufficient condition for code  $X$  applied to the group testing problem described in the abstract of our paper. In this case, only on the base of the known *number of positive responses*  $|\mathbf{x}(\mathcal{S})|$ , we decide that the controllable circuit identified by an unknown collection  $\mathcal{S}$ ,  $\mathcal{S} \subset [t]$ , is  $s$ -active, i.e., the unknown size  $|\mathcal{S}| \leq s$  ( $s$ -defective, i.e., the unknown size  $|\mathcal{S}| \geq s + 1$ ) if  $|\mathbf{x}(\mathcal{S})| \leq T$  ( $|\mathbf{x}(\mathcal{S})| \geq T + 1$ ).

**Remark 1.** The concept of  $s^T$ -codes was motivated by troubleshooting in complex electronic circuits using a non-adaptive identification scheme which was considered in [2].

## 1.2 Hypothesis Test

Let a circuit of size  $t$  is identified by an unknown collection  $\mathcal{S}_{un}$ ,  $\mathcal{S}_{un} \subset [t]$ , of defective elements of an unknown size  $|\mathcal{S}_{un}|$  and  $X$  be a code (1) of size  $t$  and length  $N$ . Introduce the null hypothesis  $\{H_0 : |\mathcal{S}_{un}| \leq s\}$  (the circuit is  $s$ -active) verse the alternative  $\{H_1 : |\mathcal{S}_{un}| \geq s + 1\}$  (the circuit is  $s$ -defective). In this paper we focus on the testing of statistical hypotheses  $H_0$  and  $H_1$ . The similar problem related to constructing of a confidence interval for  $|\mathcal{S}_{un}|$  was considered in [3], where the authors construct the interval  $[\hat{s}/c; \hat{s}]$ , such that given a *random code*  $X$ , the statistic  $\hat{s}$ , i.e., a function of the random response vector  $\mathbf{x}(\mathcal{S}_{un})$ , satisfies the following properties:  $\Pr\{\hat{s} < |\mathcal{S}_{un}|\}$  is upper bounded by a small parameter  $\epsilon \ll 1$  and the expected value of  $\hat{s}/|\mathcal{S}_{un}|$  is upper bounded by a number  $c > 1$ .

For fixed parameters  $s$ ,  $1 \leq s < t$ , and  $T$ ,  $1 \leq T < N$ , consider the following *threshold decision rule* motivated by Definition 2, namely:

$$\begin{cases} \text{accept } \{H_0 : |\mathcal{S}_{un}| \leq s\} & \text{if } |\mathbf{x}(\mathcal{S}_{un})| \leq T, \\ \text{accept } \{H_1 : |\mathcal{S}_{un}| \geq s + 1\} & \text{if } |\mathbf{x}(\mathcal{S}_{un})| \geq T + 1. \end{cases} \quad (3)$$

For the conventional *statistical* interpretation of the decision rule (3), it is reasonable to assume that the different collections of defective elements of the same size are *equiprobable*. That is why, we set that the *probability distribution* of the random collection  $\mathcal{S}_{un}$ ,  $\mathcal{S}_{un} \subset [t]$ , is identified by an unknown probability vector  $\mathbf{p} \triangleq (p_0, p_1, \dots, p_t)$ ,  $p_k \geq 0$ ,  $k = 0, 1, \dots, t$ ,  $\sum_{k=0}^t p_k = 1$ , as follows:

$$\Pr\{\mathcal{S}_{un} = \mathcal{S}\} \triangleq \frac{p_{|\mathcal{S}|}}{\binom{t}{|\mathcal{S}|}} \quad \text{for any subset } \mathcal{S} \subseteq [t]. \quad (4)$$

Introduce a *maximal error probability* of the decision rule (3) :

$$\varepsilon_s(T, \mathbf{p}, X) \triangleq \max \{ \Pr\{\text{accept } H_1 | H_0\}, \Pr\{\text{accept } H_0 | H_1\} \} \quad (5)$$

where the conditional probabilities in the right-hand side of (5) are identified by (3)-(4). Note that the number  $\varepsilon_s(T, \mathbf{p}, X) = 0$  if and only if the code  $X$  is an  $s^T$ -code. Denote by  $t_s(N, T)$  the maximal size of  $s^T$ -codes of length  $N$ . For a parameter  $\tau$ ,  $0 < \tau < 1$ , introduce the rate of  $s^{\lfloor \tau N \rfloor}$ -codes:

$$R_s(\tau) \triangleq \overline{\lim}_{N \rightarrow \infty} \frac{\log_2 t_s(N, \lfloor \tau N \rfloor)}{N} \geq 0, \quad 0 < \tau < 1. \quad (6)$$

**Definition 3.** Let  $\tau$ ,  $0 < \tau < 1$ , and a parameter  $R$ ,  $R > R_s(\tau)$ , be fixed. For the maximal error probability  $\varepsilon_s(T, \mathbf{p}, X)$ , defined by (3)-(5), consider the function

$$\varepsilon_s^N(\tau, R) \triangleq \max_{\mathbf{p}} \left\{ \min_X \varepsilon_s(\lfloor \tau N \rfloor, \mathbf{p}, X) \right\}, \quad R > R_s(\tau), \quad (7)$$

where the minimum is taken over all codes  $X$  of length  $N$  and size  $t = \lfloor 2^{RN} \rfloor$  with parameter  $R > R_s(\tau)$ . The number  $\varepsilon_s^N(\tau, R) > 0$  does not depend on the unknown probability vector  $\mathbf{p}$  and can be called the *universal* error probability of the decision rule (3). The corresponding *error exponent*

$$E_s(\tau, R) \triangleq \overline{\lim}_{N \rightarrow \infty} \frac{-\log_2 \varepsilon_s^N(\tau, R)}{N}, \quad s \geq 1, \quad 0 < \tau < 1, \quad R > R_s(\tau) \quad (8)$$

identifies the asymptotic behavior of  $\alpha$ -level of significance for the decision rule (3), i.e.,

$$\alpha \triangleq \exp_2\{-N [E_s(\tau, R) + o(1)]\}, \quad \text{if } E_s(\tau, R) > 0, \quad N \rightarrow \infty. \quad (9)$$

Along with (3) we introduce the *disjunctive decision rule* based on the conventional algorithm:

$$\begin{cases} \text{accept } H_0 & \text{if } \mathbf{x}(\mathcal{S}_{un}) \text{ covers } \leq s \text{ codewords of } X, \\ \text{accept } H_1 & \text{if } \mathbf{x}(\mathcal{S}_{un}) \text{ covers } \geq s + 1 \text{ codewords of } X. \end{cases} \quad (10)$$

For a fixed code rate  $R$ ,  $R > 0$ , the error exponent for disjunctive decision rule (10)  $E_s(R)$  is defined similarly to (5)-(8). The function  $E_s(R)$  was firstly introduced in our paper [1], where we proved

**Theorem 1.** [1]. *The function  $E_s(R) = 0$  if  $R \geq 1/s$ .*

## 2 Lower Bounds on Error Exponents

In this Section, we formulate and compare the random coding lower bounds for the both of error exponents  $E_s(R)$  and  $E_s(\tau, R)$ . These bounds were proved applying the random coding method based on the ensemble of constant-weight codes. A parameter  $Q$  in formulations of theorems 2-3 means the relative weight of codewords of constant-weight codes. Introduce the standard notations

$$h(Q) \triangleq -Q \log_2 Q - (1 - Q) \log_2 [1 - Q], \quad [x]^+ \triangleq \max\{x, 0\}.$$

In [1], we established

**Theorem 2.** [1]. *The error exponent  $E_s(R) \geq \underline{E}_s(R)$  where the random coding lower bound*

$$\underline{E}_s(R) \triangleq \max_{0 < Q < 1} \min_{Q \leq q < \min\{1, sQ\}} \{ \mathcal{A}(s, Q, q) + [h(Q) - qh(Q/q) - R]^+ \},$$

$$\mathcal{A}(s, Q, q) \triangleq (1 - q) \log_2 (1 - q) + q \log_2 \left[ \frac{Qy^s}{1 - y} \right] + sQ \log_2 \frac{1 - y}{y} + sh(Q),$$

and  $y$  is the unique root of the equation

$$q = Q \frac{1 - y^s}{1 - y}, \quad 0 < y < 1.$$

In addition, as  $s \rightarrow \infty$  and  $R \leq \frac{\ln 2}{s}(1 + o(1))$ , the lower bound  $\underline{E}_s(R) > 0$ .

**Theorem 3.** 1. The error exponent  $E_s(\tau, R) \geq \underline{E}_s(\tau, R)$  where the random coding bound  $\underline{E}_s(\tau, R)$  does not depend on  $R > 0$  and has the form:

$$\begin{aligned} \underline{E}_s(\tau, R) &\triangleq \max_{1-(1-\tau)^{1/(s+1)} < Q < 1-(1-\tau)^{1/s}} \min \{ \mathcal{A}'(s, Q, \tau), \mathcal{A}(s+1, Q, \tau) \}, \\ \mathcal{A}'(s, Q, q) &\triangleq \begin{cases} \mathcal{A}(s, Q, q), & \text{if } Q \leq q \leq sQ, \\ \infty, & \text{otherwise.} \end{cases} \end{aligned}$$

2. As  $s \rightarrow \infty$  the optimal value of  $\underline{E}_s(\tau, R)$  satisfies the inequality:

$$\underline{E}_{\text{Thr}}(s) \triangleq \max_{0 < \tau < 1} \underline{E}_s(\tau, R) \geq \frac{\log_2 e}{4s^2}(1 + o(1)), \quad s \rightarrow \infty.$$

It is possible to use decision rule (3) with any value of parameter  $T$ . The numerical values of the optimal error exponent  $\underline{E}_{\text{Thr}}(s)$  along with the corresponding optimal values of threshold parameter  $\tau$  and the constant-weight code ensemble parameter  $Q$  are presented in Table 1. Besides in Table 1 the values of  $\underline{E}_s(0)$  and  $R_{\text{cr}} \triangleq \sup\{R : \underline{E}_s(R) > \underline{E}_{\text{Thr}}(s)\}$  are shown. Theorems 1-3 show that, for large values of  $R$ , the threshold decision rule (3) has an advantage over the disjunctive decision rule (10) as  $N \rightarrow \infty$ .

Table 1: The numerical values of  $\underline{E}_{\text{Thr}}(s)$

$s$	2	3	4	5	6	7	8
$\underline{E}_{\text{Thr}}(s)$	0.1380	0.0570	0.0311	0.0196	0.0135	0.0098	0.0075
$\tau$	0.2065	0.1365	0.1021	0.0816	0.0679	0.0582	0.0509
$Q$	0.1033	0.0455	0.0255	0.0163	0.0113	0.0083	0.0064
$\underline{E}_s(0)$	0.3651	0.2362	0.1754	0.1397	0.1161	0.0994	0.0869
$R_{\text{cr}}$	0.2271	0.1792	0.1443	0.1201	0.1027	0.0896	0.0794

### 3 Simulation for finite code parameters

For finite  $N$  and  $t$ , we carried out a simulation as follows. The probability distribution vector  $\mathbf{p}$  (4) is defined by

$$p_k \triangleq \binom{t}{k} p^k (1-p)^{t-k}, \quad p \triangleq \frac{s+1/2}{t}, \quad 0 \leq k \leq t,$$

i.e. the number of defective elements  $|\mathcal{S}_{un}|$  is binomially distributed and has the expected value  $s + 1/2$ . A code  $X$  is generated randomly from the ensemble of constant-weight codes, i.e. for some weight parameter  $w$ , every codeword of  $X$  is chosen independently and equalprobably from the set of all  $\binom{t}{w}$  codewords. For every weight  $w$  and every decision rule, we repeat generation 1000 times and choose the code with minimal error probability. Note that for disjunctive decision rule  $\Pr\{\text{accept } H_0|H_1\} = 0$ . In Table 2 the best values of maximal error probability for fixed parameters  $s$ ,  $t$  and  $N$  are shown in bold.

Table 2: Results of simulation

$N$	Threshold decision rule			Disjunctive decision rule		
	$\Pr\{\text{acc. } H_1 H_0\}$	$\Pr\{\text{acc. } H_0 H_1\}$	$w$	$T$	$\Pr\{\text{acc. } H_1 H_0\}$	$w$
$s = 2, \quad t = 15$						
5	<b>0.1366</b>	0.1355	2	3	0.4780	2
8	0.0732	<b>0.0824</b>	3	5	0.3610	2
10	0	<b>0.0744</b>	1	2	0.2390	3
12	0	<b>0.0440</b>	1	2	0.1220	3
14	0	<b>0.0349</b>	2	4	0.0537	3
15	0	0.0258	2	4	<b>0.0195</b>	3
$s = 2, \quad t = 20$						
5	<b>0.1398</b>	0.1365	2	3	0.5356	2
8	<b>0.0897</b>	0.0890	3	5	0.4169	2
10	<b>0.0897</b>	0.0858	3	5	0.3008	3
12	<b>0.0580</b>	0.0576	4	7	0.1979	3
15	0	<b>0.0324</b>	2	4	0.0792	4

## References

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