# A Fast Algorithm for Optimal Wire-Sizing Under Elmore Delay Model \*

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#### Abstract

In this paper, we present a fast algorithm for continuous wire-sizing under the distributed Elmore delay model. Our algorithm GWSA-C is an extension of the GWSA algorithm (for discrete wire-sizing) in [CL93a]. GWSA-C is an iterative algorithm with guaranteed convergence to optimal wire-sizing solutions. When specialized to discrete wire-sizing problems, GWSA-C is an improved implementation of GWSA, both in terms of runtime and memory usage. GWSA-C is extremely fast even on very large problems. For example, we can solve a 1,000,000 wire-segment problem in less than 3 minutes runtime on an IBM RS/6000 workstation. This kind of efficiency in wire-sizing has never been reported in the literature.

### 1 Introduction

Since interconnect delay has become the dominant factor of circuit delay in deep sub-micron fabrication technology, it is clear that sizing the gates alone is not enough to achieve desirable circuit performance. As a result, wire-sizing for performance optimization is currently a very active research area ([CL93a] [ZDX93] [MBP95] [SP94]).

The wire-sizing optimization problem was first introduced in [CL93a] [CL93b]. The authors in [CL93a] use the monotonicity, separability, and dominance properties under the Elmore delay model to determine an optimal solution for the discrete wire-sizing problem. (By discrete wire-sizing, we mean that each wire has only a finite number of possible widths.) Most of the other wire-sizing algorithms in the literature (e.g. [ZDX93] [MBP95] [SP94]) are designed for continuous wire-sizing (i.e., each wire size can be any real number within a given range), but none of them are efficient when the number of segments is in the thousands. (Note that in practice it is possible to have "large" problems. For example, a 3000-pin clock tree has about 6,000 tree branches; if each branch has two wire segments, we have more than 10,000 wire segments!)

In this paper, we present a fast algorithm GWSA-C for continuous wire-sizing under the distributed Elmore delay model. GWSA-C is an extension of the GWSA algorithm (for discrete wire-sizing) in [CL93a]. Basically, GWSA-C is a greedy algorithm based on iteratively re-sizing the wire segments. In each iteration, the wire segments are examined one at a time; each time a wire segment is re-sized optimally

while keeping the widths of the other segments fixed. The memory requirement by GWSA-C is O(n) and each iteration in re-sizing all wire segments (once per wire segment) takes O(n) time, where n is the number of wire segments. GWSA-C is guaranteed to converge to optimal wire-sizing solutions. When specialized to discrete wire-sizing problems, GWSA-C is an improved implementation of GWSA, both in terms of runtime and memory usage. GWSA-C is extremely fast even on very large problems. For example, we can solve a 1,000,000 wire-segment problem in less than 3 minutes runtime on an IBM RS/6000 workstation. This kind of efficiency in wire-sizing has never been reported in the literature.

The remainder of this paper is organized as follows: In Section 2, we present the formulation of the wire-sizing problem under the Elmore delay model. In Section 3, we present our wire-sizing algorithm GWSA-C. Experimental results are presented in Section 4.

### 2 Problem Formulation

Given a routing tree T consisting of a driver  $w_0$  (at the root) with resistance  $r_0$ , a set of s sinks  $\{N_1, N_2, ..., N_s\}$ with capacitance load  $c_i^s$ ,  $1 \le i \le s$ , and n wire segments  $\{w_1, w_2, ..., w_n\}$ . Also given is a set of non-negative weights  $\{\lambda_1, \lambda_2, ..., \lambda_s\}$  with  $\sum_{i=1}^s \lambda_i = 1$ , where  $\lambda_i$  is a weight associated with sink N<sub>i</sub> representing the criticality of the sink. Let  $l_i$  be the length of  $w_i$  and  $x_i$  be the wire size of  $w_i$ ,  $1 \le i \le n$ .  $\mathbf{x} = (x_1, x_2, ..., x_n)$  which will be referred to as a wire-sizing solution. We are also given  $U_i$  and  $L_i$  as the respective upper bound and lower bound of the wire size of  $w_i$ , i.e.  $L_i \leq x_i \leq U_i$ ,  $1 \leq i \leq n$ . Let  $dec(w_i)$  be the set of descendants of  $w_i$  (excluding  $w_i$ ). Let  $ans(w_i)$  be the set of ancestors of  $w_i$  (excluding  $w_i$  and including  $w_0$ ). The resistance and capacitance of wire segments  $w_i$  is denoted by  $r_i$  and  $c_i$ , respectively. Let  $c_0 = 0$ . We use  $\mu_i$  to denote  $\sum_{N_j \in dec(w_i)} \lambda_j, \ 1 \le i \le n. \text{ Let } R_i = \sum_{w_j \in ans(w_i)} \mu_j r_j \text{ and }$  $C_i = \sum_{w_j \in dec(w_i)} c_j + \sum_{N_j \in dec(w_i)} c_j^s, 1 \le i \le n.$   $R_i$  and  $C_i$ will be referred to as the weighted upstream resistance and the downstream capacitance, respectively, of  $w_i$ .

We use the  $\pi$ -model to model each wire segment  $w_i$  (see Figure 1). We have the resistance of  $w_i$  is given by  $r_i = \alpha l_i/x_i$ , the capacitance of  $w_i$ , is given by  $c_i = \beta x_i l_i/2$ , where  $\alpha$  is the resistance per unit length at unit width and  $\beta$  is the capacitance per unit square. We use the Elmore delay model [E48] to estimate the delay. The signal delay at sink  $N_i$  is

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wire size x, length l
$$\beta \times 1/2 \stackrel{\alpha}{=} 1/x$$

Figure 1: RC model for wire segment

computed by:

$$D_i = \sum_{w_j \in P_i} r_j \left( C_j + \frac{c_j}{2} \right), \tag{1}$$

where  $P_i$  is the set of wire segments that lie on the path from the root to sink  $N_i$ .

In this paper, we consider the following wire-sizing optimization problem:

Minimize 
$$F(\mathbf{x}) = \sum_{i=1}^{s} \lambda_i D_i$$
Subject to 
$$L_j \le x_j \le U_j, \ 1 \le j \le n.$$

## 3 Wire-sizing Algorithm

Lemma 1 For each  $x_i$ , F(x) can be written in the following form

$$F(\mathbf{x}) = A_i(\mathbf{x})x_i + \frac{B_i(\mathbf{x})}{x_i} + E_i(\mathbf{x}),$$

where  $A_i(\mathbf{x})$ ,  $B_i(\mathbf{x})$ , and  $E_i(\mathbf{x})$  are independent of  $x_i$ ,  $A_i(\mathbf{x}) = \beta l_i R_i$ , and  $B_i(\mathbf{x}) = \alpha l_i \mu_i C_i$ .

Proof:

$$F(\mathbf{x}) = \sum_{j=1}^{s} \lambda_{j} D_{j}$$

$$= \sum_{j=1}^{s} \lambda_{j} \sum_{w_{k} \in P_{j}} r_{k} (C_{k} + \frac{c_{k}}{2})$$

$$= \sum_{w_{k} \in T} r_{k} (C_{k} + \frac{c_{k}}{2}) \sum_{N_{j} \in dec(w_{k})} \lambda_{j}$$

$$= \sum_{c \in T} r_{k} \mu_{k} (C_{k} + \frac{c_{k}}{2}).$$

Note that the terms that involve  $x_i$  come from  $\sum_{w_k \in ans(w_i)} r_k \mu_k C_k$ . In fact, only the term  $\beta l_i x_i$  (the wire capacitance of  $w_i$ ) in  $C_k$  contribute to the terms with  $x_i$ , hence

$$A_i(\mathbf{x}) = \beta l_i \sum_{w_j \in ans(w_i)} r_j \mu_j.$$

Since the terms that involve  $\frac{1}{x_i}$  only come from  $r_i \mu_i C_i = \frac{\alpha l_i}{x_i} \mu_i C_i$ , we have  $B_i(\mathbf{x}) = \alpha l_i \mu_i C_i$ . Let  $E_i(\mathbf{x}) = F(\mathbf{x})$ 

 $A_i(\mathbf{x})x_i - \frac{B_i(\mathbf{x})}{x_i}$ . It is clear that  $A_i(\mathbf{x})$ ,  $B_i(\mathbf{x})$ , and  $E_i(\mathbf{x})$  are independent of  $x_i$ 

If we re-size  $w_i$  while keeping the wire-sizes of all the other wire segments fixed, we say that it is a local re-sizing of  $w_i$ . An optimal local re-sizing of  $w_i$  is a local re-sizing that minimize  $F(\mathbf{x})$ .

Lemma 2 Let  $\hat{\mathbf{x}} = (\hat{x_1}, \hat{x_2}, \dots, \hat{x_n})$  be a wire-sizing solution. An optimal local re-sizing of  $w_i$  is given by changing the width of  $w_i$  to

$$x_i^* = min\left(U_i, max\left(L_i, \sqrt{\frac{B_i(\hat{\mathbf{x}})}{A_i(\hat{\mathbf{x}})}}\right)\right).$$

**Proof:** Suppose we change the wire-size of  $w_i$  to  $x_i$ . It follows from Lemma 1 that the new delay cost is given by

$$G(x_i) = A_i(\hat{\mathbf{x}})x_i + \frac{B_i(\hat{\mathbf{x}})}{x_i} + E_i(\hat{\mathbf{x}})$$

Differentiating G with respect to  $x_i$ , we get

$$\frac{dG}{dx_i} = A_i(\hat{\mathbf{x}}) - \frac{B_i(\hat{\mathbf{x}})}{x_i^2}$$

Let 
$$\theta(\hat{\mathbf{x}}) = \sqrt{\frac{B_i(\hat{\mathbf{x}})}{A_i(\hat{\mathbf{x}})}}$$
. Note that

$$\begin{cases} \frac{dG}{dx_i} = 0, & x_i = \theta(\hat{\mathbf{x}}), \\ \frac{dG}{dx_i} < 0, & x_i < \theta(\hat{\mathbf{x}}), \\ \frac{dG}{dx_i} > 0, & x_i > \theta(\hat{\mathbf{x}}). \end{cases}$$

Hence  $G(x_i)$  is decreasing when  $x_i < \theta(\hat{\mathbf{x}})$ ,  $G(x_i)$  is increasing when  $x_i > \theta(\hat{\mathbf{x}})$ , and  $G(x_i)$  is minimum at  $x_i = \theta(\hat{\mathbf{x}})$ .

Let  $x_i^*$  be the wire-size of  $w_i$  in an optimal local re-sizing of the wire segments. We consider three cases.

Case 1:  $\theta(\hat{\mathbf{x}}) \in [L_i, U_i]$ . In this case, we have  $x_i^* = \theta(\hat{\mathbf{x}})$ .

Case 2:  $\theta(\hat{\mathbf{x}}) > U_i$ .

Since  $G(x_i)$  is decreasing on  $[L_i, U_i]$ , we have  $x_i^* = U_i$ .

Case 3:  $\theta(\hat{\mathbf{x}}) < L_i$ .

Since  $G(x_i)$  is increasing on  $[L_i, U_i]$ , we have  $x_i^* = L_i$ . From the above discussion, we have

$$x_{i}^{*} = min\left(U_{i}, max\left(L_{i}, \sqrt{\frac{B_{i}(\hat{\mathbf{x}})}{A_{i}(\hat{\mathbf{x}})}}\right)\right) \square$$

Theorem 1  $\mathbf{x}^* = (x_1^*, \dots, x_n^*)$  is an optimal wire-sizing solution if and only if

$$x_i^* = min\left(U_i, max\left(L_i, \sqrt{\frac{B_i(\mathbf{x}^*)}{A_i(\mathbf{x}^*)}}\right)\right), 1 \le i \le n.$$

Proof:

 $(\Rightarrow)$  This follows directly from Lemma 2 and the fact that optimal local re-sizing of  $w_i$  would not change its width.

 $(\Leftarrow)$  If  $x_i^* = min\left(U_i, max\left(L_i, \sqrt{\frac{B_i(\mathbf{X}^*)}{A_i(\mathbf{X}^*)}}\right)\right)$ , we show that  $\mathbf{x}^*$  is a global minimum point of  $F(\mathbf{x})$  over  $\Omega_x = {\mathbf{x} | L_i \leq }$  $x_i \leq U_i, 1 \leq i \leq n$ ; i.e.  $\mathbf{x}^*$  is an optimal wiresizing solution. Note that F(x) is a posynominal in x, and that under the transformation  $x_i = e^{z_i}$ ,  $1 \le i \le n$ ,  $G(\mathbf{z}) = F(e^{z_1}, \dots, e^{z_n})$  is convex over  $\Omega_z = \{ \mathbf{z} \mid L_i \leq e^{z_i} \leq e^{z_i} \}$  $U_i, 1 \le i \le n$  (see [DPZ67]). Let  $\mathbf{z}^* = (z_1^*, \dots, z_n^*)$  where  $x_i^* = e^{z_i^*}, 1 \le i \le n$ . We now consider three cases:

Case 1: 
$$x_i^* = \sqrt{\frac{B_i(\mathbf{X}^*)}{A_i(\mathbf{X}^*)}}$$
.

In this case, we have  $\frac{\partial F}{\partial x}(\mathbf{x}^*) = 0$ . Thus

$$\frac{\partial G}{\partial z_i}(\mathbf{z}^*) = \frac{\partial F}{\partial x_i}(\mathbf{x}^*) \frac{\partial x_i}{\partial z_i}(\mathbf{z}^*) = \frac{\partial F}{\partial x_i}(\mathbf{x}^*) e^{z_i^*} = 0.$$

In this case,  $L_i \geq \sqrt{\frac{B_i(\mathbf{X}^*)}{A_i(\mathbf{X}^*)}}$ . We have  $\frac{\partial F}{\partial x_i} \geq 0$ , and  $z_i$  $z_i^* > 0, \forall z \in \Omega_z$ . Hence

$$\frac{\partial G}{\partial z_i}(\mathbf{z}^*)(z_i - z_i^*) = \frac{\partial F}{\partial x_i}(\mathbf{x}^*)e^{z_i^*}(z_i - z_i^*) \ge 0, \, \forall \mathbf{z} \in \Omega_z.$$

Case 3:  $x_{i}^{*} = U_{i}$ .

In this case,  $U_i \leq \sqrt{\frac{B_i(\mathbf{X}^*)}{A_i(\mathbf{X}^*)}}$ . We have  $\frac{\partial F}{\partial x_i}(\mathbf{X}^*) \leq 0$  and  $z_i - z_i^* \leq 0$ ,  $\forall \mathbf{z} \in \Omega_z$ . Hence

$$\frac{\partial G}{\partial z_i}(\mathbf{z}^*)(z_i - z_i^*) = \frac{\partial F}{\partial x_i}(\mathbf{x}^*)e^{z_i^*}(z_i - z_i^*) \ge 0, \, \forall \mathbf{z} \in \Omega_z.$$

Combining the three cases, we get  $\frac{\partial G}{\partial z_i}(\mathbf{z}^*)(z_i - z_i^*) \geq 0$ ,  $\forall \mathbf{z} \in \Omega_z$ . Since G is convex (see [L84]), we have, for all  $z \in \Omega_z$ ,

$$G(\mathbf{z}) - G(\mathbf{z}^*) \geq \nabla G(\mathbf{z}^*)(\mathbf{z}) - \mathbf{z}^*)$$

$$= \sum_{i=1}^{n} \frac{\partial G}{\partial z_i}(\mathbf{z}^*)(z_i - z_i^*)$$

$$\geq 0.$$

Thus  $F(\mathbf{x}) - F(\mathbf{x}^*) = G(\mathbf{z}) - G(\mathbf{z}^*) \ge 0, \forall \mathbf{x} \in \Omega_x$ . Therefore x\* is a global minimum point.

We now describe our algorithm GWSA-C. Basically, GWSA-C is a greedy algorithm based on iteratively re-sizing the wire segments. In each iteration, the wire segments are examined one at a time; each time a wire segment is re-sized optimally while keeping the sizes of the other segments fixed. Lemma 2 gives the formula to compute the new size of a wire segment. Note that we can pre-compute the set of all edge weights  $\mu_i$ 's in O(n) time by a bottom-up traversal of the tree. The computation of each  $\mu_i$  is based on previously computed  $\mu_j's$ , using the formula  $\mu_i = \sum_{w_j \in children(w_i)} \mu_j$ where  $children(w_i)$  is the set of the children of  $\mu_i$ . At the beginning of each iteration, we compute all downstream capacitances  $C_i$ 's by a bottom-up traversal of T in O(n) time. The formula for computing  $C_i$  is  $C_i = \sum_{w_j \in children(w_i)} (C_j + c_j)$ . Each iteration of GWSA-C consists of a top-down traversal

of T; each time we visit a wire segment  $w_i$ , we re-size  $w_i$ according to the formula in Lemma 2 where we need  $A_i(\mathbf{x})$ and  $B_i(\mathbf{x})$ . Observe that  $R_i = R_{p_i} + \mu_{p_i} r_{p_i}$ , where  $w_{p_i}$  is the parent of  $w_i$ . Thus,  $A_i(\mathbf{x}) = A_{p_i}(\mathbf{x}) + \beta l_i \mu_{p_i} r_{p_i}$ . Since  $C_i$  is independent of the wire sizes of the ancestors of  $w_i$ , there is no need to update  $C_i$  during the top-down traversal of T. We have,  $B_i(\mathbf{x}) = \alpha l_i \mu_i C_i$ . Clearly, each iteration of re-sizing the wire segments takes O(n) time. It is also easy to see that the memory usage by GWSA-C is also O(n). We summarize the description of GWSA-C in Figure 2. Since we terminate the program when no further improvement is possible, it follows from Theorem 2 that GWSA-C converges to an optimal wire-sizing solution. We have the following theorem.

Theorem 2 Algorithm GWSA-C converges to an optimal wire-sizing solution. It runs in O(rn) time using O(n) storage, where n is the number of wire segments and r is the number of iterations.

## Algorithm: GWSA-C

- **S1.**  $x_i := L_i, 1 \le i \le n;$
- Compute all  $\mu_i$ 's by a bottom-up traversal of T using the following formula:

$$\mu_i := \sum_{w_j \in children(w_i)} \mu_j;$$

Compute all  $C_i$ 's by a bottom-up traversal of T using the following formula:

$$C_i := \sum_{w_j \in children(w_i)} (C_j + c_j);$$

Perform a top-down traversal of T:

For each 
$$w_i$$
,  

$$A_i(\mathbf{x}) := A_{p_i}(\mathbf{x}) + \beta l_i \mu_{p_i} r_{p_i};$$

$$B_i(\mathbf{x}) := \alpha l_i \mu_i C_i;$$

$$x_i = min\left(U_i, max\left(L_i, \sqrt{\frac{B_i(\mathbf{x})}{A_i(\mathbf{x})}}\right)\right);$$

S5. Repeat S3-S4 until no improvement.

Figure 2: The GWSA-C algorithm.

Remark GWSA-C can be specialized to solve the discrete wire-sizing problem. Let  $s_1 \leq s_2 \leq ... \leq s_k$  be the set of feasible wire sizes for  $w_i$ . We modify GWSA-C as follows. In S4, after computing  $x_i$ , we let  $s_j$  and  $s_{j+i}$  be the two consective feasible wire sizes such that  $s_j \leq x_i \leq s_{j+1}$ . It is not hard to see that the optimal local re-sizing of  $w_i$  is obtained by checking the delay costs  $G(s_i)$  and  $G(s_{i+1})$  and pick the smaller one. For discrete wire-sizing problems, GWSA-C is an improved implementation of GWSA in [CL93a], both in terms of runtime and memory usage.

## Experimental Results

We implemented and tested our algorithm on an IBM/RS6000 workstation. We first compare the runtime of GWSA-C with GWSA in [CL93a] on a single wire with 100 to 10<sup>6</sup> segments. The GWSA package we used was obtained from the authors of [CL93a] in July 1995. The parameters used are shown in Table 1. Table 2 shows the runtime and memory usage comparisons between GWSA and GWSA-C. It shows that GWSA-C runs much more efficiently than GWSA, and that the number of iterations grows very slowly. Due to the requirement of large memory usage, GWSA algorithm could not get any results when the number of wires segments exceeded 2000. Figures 3 and 4 show that the runtime and memory usage of GWSA-C are close to linear. Table 3 shows that GWSA-C improved the delay cost in the clock tree r1-r5 [T93] by up to 89%. Table 4 showsthat GWSA-C produced near-optimal results for discrete wiresizing problems. The lower bound was obtained by running the continuous version of GWSA-C.

Unit Resistance:	0.003	$\Omega/\mu m$
Unit Capacitance:	$2 \times 10^{-17}$	$F/\mu m$
Minimum Wire Width:	1	$\mu m$
Maximum Wire Width:	20	$\mu m$

Table 1: RC Parameters.

# of wire	GWSA [CL93a]		GWSA-C		No of
segments	Time	Memory	Time	Memory	Iters
100	0.15	1968	0.01	32	9
200	0.76	2408	0.02	32	10
1000	22.50	17464	0.08	40	10
2000	106.68	64500	0.17	48	12
10000	-	-	1.12	108	15
20000	-	-	2.35	188	16
100000	-	-	13.52	812	18
1000000	-	-	141.42	7884	19

Table 2: Runtime and memory usage comparisons between GWSA and GWSA-C. (Unit:Runtime(second), Memory(kbytes), Step Width( $1\mu m$ )

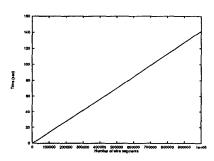


Figure 3: Runtime of GWSA-C.

Circuit Name	No of Wire	Delay (ns)		Improve%
	Segments	Initial	Final	1
r1	533	0.77	0.12	84.42
r2	1195	2.11	0.27	87.20
r3	1723	3.38	0.43	87.28
r4	3805	9.09	1.03	88.67
r5	6201	15.86	1.65	89.04

Table 3: Result Comparisons between with and without wire-sizing by GWSA-C.

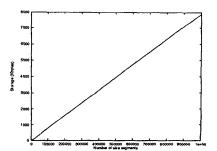


Figure 4: Memory Usage of GWSA-C.

Wire	Delay (ps)			Ratio
Length	Initial	GWSA-C	lower bound	
100	0.52	0.12	0.11	1.09
200	2.66	0.47	0.47	1.00
500	15.72	3.01	2.99	1.01
1000	63.45	12.07	11.98	1.01
10000	6051.50	1209.54	1200.73	1.01
50000	150365.34	30244.42	30024.13	1.01

Table 4: Results for GWSA-C for discrete wire sizes. (Unit:wire length( $\mu m$ ))

## 5 Acknowledgment

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