



# Microarray Layout and the Quadratic Assignment Problem

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## Outline

- 1 Introduction to Microarray Layout
- Conflict Index Model
- 3 New Approach: Quadratic Assignment Problem (QAP)

#### Outline

Introduction

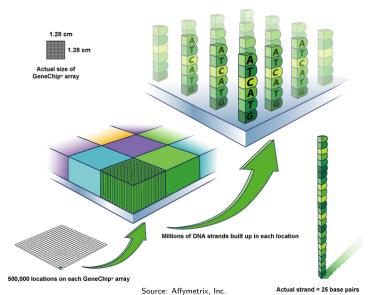
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- 1 Introduction to Microarray Layout
- Conflict Index Model
- 3 New Approach: Quadratic Assignment Problem (QAP)

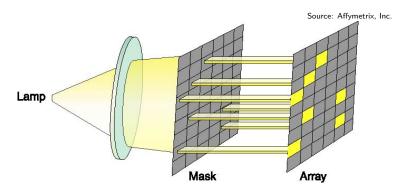
# High-Density Oligonucleotide Microarrays

Introduction

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# Probe Synthesis with Photolitographic Masks



- Probes are synthesized on the chip in a series of steps
- Each step appends a particular nucleotide to selected regions
- Selection occurs by exposure to light directed by a mask

$\rho_1$	$p_2$	$p_3$
ACT	CTG	GAT
$\rho_4$	<i>p</i> <sub>5</sub>	$p_6$
TCC	GAC	GCC
$p_7$	$p_8$	$p_9$
TAC	CGT	AAT

```
ACGTACGTACGT
8 9
```

**QAP** Formulation

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$p_1$	<b>p</b> <sub>2</sub>	$p_3$
ACT	CTG	GAT
$p_4$	$p_5$	$p_6$
TCC	GAC	GCC
P <sub>7</sub>	p <sub>8</sub>	p <sub>9</sub>

```
S = ACGTACGTACGT
```

$\rho_1$	$p_2$	$p_3$
ACT	CTG	GAT
<i>p</i> <sub>4</sub>	<b>p</b> <sub>5</sub>	<b>p</b> <sub>6</sub>
TCC	GAC	GCC
$p_7$	$p_8$	$p_9$

```
S = ACGTACGTACGT
\varepsilon_1 = A - - - - - -
```

$\rho_1$	$p_2$	$p_3$
ACT	CTG	GAT
$p_4$	$p_5$	$p_6$
TCC	<b>G</b> AC	GCC
<i>p</i> <sub>7</sub>	<i>p</i> <sub>8</sub>	$p_9$
TAC	CGT	AAT

```
S = ACGTACGTACGT
\mathcal{E}_1 = A - - - - - -
\varepsilon_2 = -C -----
\mathcal{E}_3 = --G-----
\varepsilon_5 = --G-----
\varepsilon_6 = --G-----
\mathcal{E}_{8} = -CG-----
\mathcal{E}_{9} = A - - - - - - -
```

$p_1$	<i>p</i> <sub>2</sub>	$p_3$
ACT	CTG	GAT
<i>p</i> <sub>4</sub>	$p_5$	$p_6$
TCC	GAC	GCC
p <sub>7</sub>	<i>p</i> <sub>8</sub>	$p_9$

```
S = ACGTACGTACGT
\varepsilon_1 = A^{-----}
\varepsilon_2 = -C -----
\mathcal{E}_3 = --G-----
\mathcal{E}_4 = ---T
\mathcal{E}_{5} = --G-----
\varepsilon_6 = --G------
\mathcal{E}_7 = ---T
\varepsilon_8 = -CGT - - - - -
\mathcal{E} \circ = A - - - - - - - -
```

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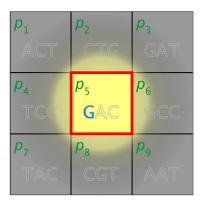
$\rho_1$	$p_2$	$p_3$
ACT	CTG	GAT
<i>p</i> <sub>4</sub>	$p_5$	$p_6$
TCC	GAC	GCC
<i>p</i> <sub>7</sub>	$p_8$	$p_9$
TAC	CGT	AAT

ρ <sub>1</sub> ACT	p <sub>2</sub> CTG	ρ <sub>3</sub> GAT
P <sub>4</sub> TCC	p <sub>5</sub> GAC	p <sub>6</sub> GCC
P <sub>7</sub> TAC	p <sub>8</sub>	ρ <sub>9</sub> AAT

```
S = ACGTACGTACGT
\varepsilon_1 = A - - - C - T - - - -
\varepsilon_2 = -C - - - T - G -
\mathcal{E}_3 = --G-A--T---
\varepsilon_4 = ---T-C---C--
\varepsilon_5 = --G-A----C--
\varepsilon_6 = --G - -C - - -C - -
\mathcal{E}_7 = ---TAC-----
\varepsilon_8 = -CGT - - - - -
\xi_9 = A - - A - - - - T
e'_{0} = A - - - - - A - T
```

Right-most: 
$$\mathcal{E}_{9}^{"} = ----A---A--T$$

Left-most:  $\mathcal{E}_{9}^{""} = A - - A - T - - -$ 



- Untargeted spots can be accidentally activated
  - Diffraction of light
  - Internal reflection
- Production of defective probes
- More likely near the borders between masked and unmasked spots: border conflict

#### Border Length Minimization Problem (Hannenhalli et al., 2002)

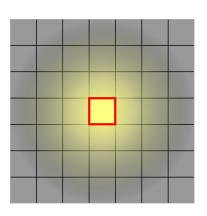
Find arrangement of the probes and embeddings with minimum number of border conflicts over all masks

#### Outline

- Conflict Index Model

#### Motivation

- Border Length measures the quality of a particular mask
  - We are more interested in a per-probe measure
- Practical considerations:
  - a) Stray light might damage probes as far as three cells away from the targeted spot
  - b) Imperfections in the middle of a probe are more harmful than in its extremities



# ATGACTACCATGCAGTACAACATAC

#### Definition

Introduction

#### Conflict Index of a probe p

$$\mathcal{C}(p) := \sum_{t=1}^T \Bigl( \omega(p,t) \sum_{\mathit{nbs.p'}} \delta(p,p',t) \Bigr)$$

#### Distance-dependent weights

$$\delta(p, p', t) := \begin{cases} (d(p, p'))^{-2} & \text{if } p' \text{ is unmasked at step } t, \\ 0 & \text{otherwise,} \end{cases}$$

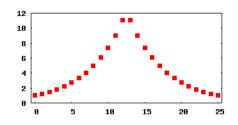
where d(p, p') is the Euclidean distance between the spots of p and p'.

0.06	0.08	0.10	0.11	0.10	0.08	0.06
0.08	0.13	0.20	0.25	0.20	0.13	0.08
0.10	0.20	0.50	1.00	0.50	0.20	0.10
0.11	0.25	1.00	p	1.00	0.25	0.11
0.10	0.20	0.50	1.00	0.50	0.20	0.10
0.08	0.13	0.20	0.25	0.20	0.13	0.08
0.06	0.08	0.10	0.11	0.10	0.08	0.06

### Definition

### Conflict Index of a probe p

$$\mathcal{C}(p) := \sum_{t=1}^{T} \left( \omega(p, t) \sum_{p'} \delta(p, p', t) \right)$$



#### Position-dependent weights

$$\omega(p,t) := \left\{ egin{array}{ll} c \cdot \exp\left(\theta \cdot \lambda(p,t)
ight) & ext{if $p$ is masked at step $t$,} \\ 0 & ext{otherwise,} \end{array} 
ight.$$

where

$$\lambda(p,t) := 1 + \min(b_{p,t}, \ell_p - b_{p,t}),$$

 $b_{p,t}$  denotes the number of nucleotides synthesized up to and including step t,  $\ell_p$  is the length of probe p, c>0 and  $\theta>0$  are constants.

# Border Length and Conflict Index

#### Redefine $\delta$ and $\omega$ as

Introduction

$$\delta(p,p',t) := \left\{ \begin{array}{l} 1 \quad \text{if } p' \text{ is a direct neighbor of } p \\ \quad \text{and is unmasked at step } t, \\ 0 \quad \text{otherwise} \end{array} \right.$$
 
$$\omega(p,t) := \left\{ \begin{array}{l} 1/2 \quad \text{if } p \text{ is masked at step } t, \\ 0 \quad \text{otherwise} \end{array} \right.$$

- Then  $\sum_{p} C(p) = \sum_{t=1}^{T} \mathcal{B}_{t}$
- $\bullet$  Border length and conflict indices are equivalent for this choice of  $\delta$  and  $\omega$
- For our choices, they are not equivalent but still correlated: a good layout has low border lengths and conflict indices

#### New Problem

#### Conflict Index Minimization Problem

Find placement of the probes and embeddings such that

$$\sum_{p} \mathcal{C}(p) o \mathsf{min}$$

#### Outline

- 3 New Approach: Quadratic Assignment Problem (QAP)

#### Previous Work: Place and Re-embed

The problem has been traditionally approached in two phases:

- 1) Placement of probes given a fixed embedding
- 2) Re-embedding of probes once a placement is fixed

#### Placement: Row-epitaxial (Kahng et al., 2003)

- Spots are filled in a pre-defined order
  - Select probe from a list Q such that conflicts with filled spots are minimized
- Restrict the maximum size of Q (e.g. Q = 20000)

#### Re-embedding: several algorithms (Kahng et al., 2002, 2003)

- Based on the Optimum Single Probe Embedding (OSPE)
  - Re-embed a probe optimally in regards to its neighbors
  - Dynamic programming, like a sequence alignment

# Previous Work: Partitioning

- The placement problem can be partitioned
  - Divide the chip into sub-regions; assign sub-sets of probes to each sub-region
  - Sub-regions are processed independently, and can be recursively partitioned
  - A placement algorithm is called on each final sub-region

#### Pivot Partitioning (Carvalho & Rahmann, 2006)

- Alternate horizontal and vertical partitions
- Allow sub-regions to have different sizes









# Quadratic Assignment Problem

#### Definition

- Given  $n \times n$  real-valued matrices  $F = (f_{ij}) \ge 0$  and  $D = (d_{kl}) \ge 0$
- Find a permutation  $\pi$  of  $\{1, 2, \dots n\}$  such that

$$\sum_{i=1}^n \sum_{j=1}^n f_{ij} \cdot d_{\pi(i)\pi(j)} \to \min$$

#### Example: Facility Location Problem

- Assign n facilities to n locations
- $f_{ii}$ : flow of materials from facility i to j
- $d_{kl}$ : distance between locations k and l
- $\bullet$   $\pi$ : one-to-one assignment with minimum cost

# QAP Formulation of Placement Problem

#### Goal: find a placement with

$$\sum_k \mathcal{C}(k) \to \min$$

#### Flow

Introduction

$$f_{ij} := \begin{cases} (d(i,j))^{-2} & \text{if spot } j \text{ is "near" spot } i, \\ 0 & \text{otherwise} \end{cases}$$

#### Distance

$$d_{kl} := \sum_{t=1}^{T} d_{klt},$$

$$d_{klt} := \left\{ egin{array}{ll} c \cdot \exp( heta \cdot \lambda(p_k, t)) & ext{if } p_k ext{ is masked and } p_l ext{ unmasked in step } t, \\ 0 & ext{otherwise} \end{array} 
ight.$$

- The placement problem can be modeled as a QAP
- But QAP is known to be NP-hard
  - Generally impossible to solve (to optimally) for  $n \ge 20$
- Several heuristics exist

#### GRASP (Li, Pardalos and Resende, 1994)

- Greedy Randomized Adaptive Search Procedure
- Comprised of two phases
  - 1) Construction: buids a random feasible solution
  - 2) Local search: search a local optimum in the neighborhood
- GRASP with Path-Relinking (Oliveira et al., 2004)

Introduction

# Results on Small Artificial Chips

#### Border Length minimization Random Row-epitaxial GRASP with Path-Relinking Time Dim Cost Cost Red. Time Cost Red. $6 \times 6$ 1 989.20 1714.60 13.80 0.01 1672.20 15.94 2.73 0.02 $7 \times 7$ 2783.20 2354.60 15.40 2332.60 16.19 6.43 $8 \times 8$ 3721.20 3123.80 16.05 0.03 3099.13 16.72 12.49 4762.00 16.53 0.05 3 967.20 16.69 $9 \times 9$ 3 974.80 25.96 $10 \times 10$ 5 985.20 4895.60 18.20 0.06 4911.40 17.94 47.57 $11 \times 11$ 7 288 40 5 954 40 18.30 0.10 5 990.73 17.80 87.48 $12 \times 12$ 8714.00 7 086.20 18.68 0.117 159.80 17.84 152.42

Dim: chip dimension Cost: total border length Red.: reduction in %

Time: running time in seconds

# Results on Small Artificial Chips

#### Conflict Index minimization

	Random	Row-epitaxial			GRASP	with Path	-Relinking
Dim	Cost	Cost	Red.	Time	Cost	Red.	Time
6×6	524.28	495.15	5.56	0.05	467.08	10.91	3.68
7×7	558.25	521.90	6.51	0.07	489.32	12.35	8.84
8×8	590.51	551.84	6.55	0.09	515.69	12.67	19.48
$9 \times 9$	613.25	568.62	7.28	0.11	533.79	12.96	38.83
$10 \times 10$	628.50	576.49	8.28	0.11	539.69	14.13	73.09
$11{ imes}11$	642.72	588.91	8.37	0.12	551.41	14.21	145.67
12×12	656.86	598.21	8.93	0.12	561.21	14.56	249.19

Dim: chip dimension

Cost: average conflict index

Red.: reduction in %

Time: running time in seconds

# Summary

- Conflict Index
  - New model for evaluating microarray layouts
- New approach to placement
  - Based on the Quadratic Assignment Problem
  - Good for very small regions... but too slow!
- Challenges
  - Make it faster?
  - Use it as a post-placement optimization
  - Formulation considering all embeddings

#### Auf Wiedersehen!

Introduction

#### More info on

http://gi.cebitec.uni-bielefeld.de/assb/chiplayout

#### **QAPLIB**

http://www.seas.upenn.edu/gaplib

- Thanks to Peter Hahn (University of Pennsylvania, USA)
- And thank you for your attention!