5. Random Sampling and Arrangement of Lines

### A central concept of statistics is

# A random sample is a good estimator for statistical population

The concept of randomized divide-and-conquer Quick-Sort

- Let N be any set of points in the real line.
- If we pick a random element S from N, then S probably divides the line into interval of roughly equal size. The size mean the number of unchosen points lying in the interval.

## Random Sampling without replacement

- Given a set N of objects, a r-element subset R of N is a random sample if every element in N is equally likely to be in R.
  - Choose the first element in  ${\cal R}$  randomly from  ${\cal N}$
  - Choose the seond element in R from the remaining n-1 elements independently and randomly.
  - Repeat the process untail r elements from N are chosen.

An interesting and important question:

Given a set N of n points in the real line, does a random sample R of N of size r divide the real line into roughly equal size?

- Let H(R) be the partition of the real line formed by R.
- For each interval I in H(R), the conflict size of I is the number of points in  $N \setminus R$  lying in I.

# Is the conflict size of each interval in H(R) O(n/r) with high probability?

Most researchers conjecture the positive answer, but no one can prove it over several centries.

# Main Theorem

For a set N of n points on the real line and a random sample R of N of size r, with probability greather than 1/2, the conflict size of each interval in H(R) is  $O([n/r] \log r)$ .

More generally, for any fixed c > 2 and any  $s \ge r > 2$ , with probability  $1 - O(1/s^{c-2})$ , the conflict size of each interval in H(R) is less than  $c(n \ln s)/(r-2)$ . In other words, the probability of some conflict size exceeding  $c(n \ln s)/(r-2)$  is small,  $O(1/s^{c-2})$  to be precise.

# Proof of Main Theorem

# Terminology

- $\Pi = \Pi(N)$  is the set of all pairs the form (p,q) where p, as well as q, is a point in N or a point at infinity.
- A point at infinity means either  $-\infty$  or  $+\infty$
- $\sigma$  is any such pair in  $\Pi$ , and thus defines an interval on the real line.
- $D(\sigma)$  is  $\{p,q\} \cap N$ , and consists of the endpoints of  $\sigma$  not at the infinity. The points in  $D(\sigma)$  is said to define  $\sigma$ .
- $d(\sigma)$  is the size of  $D(\sigma)$  and is called the *degree* of  $\sigma$ .  $d(\sigma)$  is 0, 1, or 2. - d((p,q)) = 2,  $d((-\infty, p))$ , and  $d((-\infty, +\infty))$ .
- L(σ) is the set of points in N that lies in the interior of σ. The points in L(σ) is said to conflict with σ
- $l(\sigma)$  is the size of  $L(\sigma)$  and called the *conflict size* of  $\sigma$ .
- $\Pi$  is a configuration space of N
  - An interval  $\sigma\in\Pi$  is active over a subset  $R\subseteq N$  if  $\sigma$  is an interval of H(R)
  - $-\sigma$  is an interval of H(R) if and only R contains all points in  $D(\sigma)$  but no poin in  $L(\sigma)$ .

# Conditional Probability

- Let  $R \subseteq N$  denote a random sample of N of size r.
- Let  $p(\sigma, r)$  denote the *conditional probability* that R contains no point in conflict with  $\sigma$ , given that it contains the points defining  $\sigma$ .

# Claim

$$p(\sigma, r) \le (1 - \frac{l(\sigma)}{n})^{r - d(\sigma)}$$

#### Intuition

- Since R must contain  $D(\sigma)$ , the remaining  $r d(\sigma)$  can be thought of as resulting from independent random draws.
- The probability of choosing a conflicting point in any such draw is greater than or equal to  $l(\sigma)/n$ .

Rigorous justification

- $\bullet$  Let R' be  $R \setminus D(\sigma)$
- R' is a random sample of the set  $N' = N \setminus D(\sigma)$  of size  $n d(\sigma)$
- R' is obtained from N' by  $r d(\sigma)$  successive random drwas without replacement.
- For each  $j \geq 1$ , the probability that the point chosen in the  $j^{\text{th}}$  draw does not conflict with  $\sigma$ , given that no point chosen in any previous draw conflicts with  $\sigma$ , is

$$1-\frac{l(\sigma)}{n-d(\sigma)-j}\leq 1-\frac{l(\sigma)}{n}$$

• Then

$$p(\sigma, r) = \prod_{j=1}^{r-d(\sigma)} 1 - \frac{l(\sigma)}{n-d(\sigma)-j} \le (1 - \frac{l(\sigma)}{n})^{r-d(\sigma)}$$

**Proof of Main Theorem**(continue)

• Since  $1 - l(\sigma)/n \le \exp(-l(\sigma)/n)$ , the claim implies  $p(\sigma, r) \le \exp(-\frac{l(\sigma)}{n}(r - d(\sigma)))$ ,

where  $\exp(x)$  denotes  $e^x$ .

• Since  $d(\sigma) \leq 2$ ,

$$p(\sigma, r) \le \exp(-\frac{l(\sigma)}{n}(r-2)).$$

• If  $l(\sigma) \ge c(n \ln s)/(r-2)$ , for some c > 1, then

$$p(\sigma, r) \le \exp(-c \ln s) = \frac{1}{s^c}.$$

#### Combined probability

- Let  $q(\sigma, r)$  denote the probability that R contains all points in  $D(\sigma)$ .
- The probability that  $\sigma$  is active over R is precisely  $p(\sigma, r)q(\sigma, r)$ .

The probability that some  $\sigma \in \Pi$ , with  $l(\sigma) > c(n \ln s)/(r-2)$ , is active over R is bounded by

$$\sum_{\sigma\in\Pi: l(\sigma)>\frac{cn\ln s}{r-2}} p(\sigma,r)q(\sigma,r) \leq \sum_{\sigma\in\Pi: l(\sigma)>\frac{cn\ln s}{r-2}} q(\sigma,r)/s^c \leq \frac{1}{s^c} \sum_{\sigma\in\Pi} q(\sigma,r).$$

Summary

- Let  $\pi(R)$  denote the number of intervals in  $\Pi$  whose defining points are in R.
- $\sum_{\sigma \in \Pi} q(\sigma, r)$  is  $\pi(R)$ .
- For a random sample R of N, the probability that some  $\sigma \in \Pi$ , with  $l(\sigma) > cn \ln s/(r-2)$ , is active over R is bounded by

$$\frac{1}{s^c} E[\pi(R)].$$

• Since R has r points,  $\pi(R) = \binom{r}{2} + 2r + 1 = O(r^2)$ .

$$\frac{1}{s^c} E[\pi(R)] = O(\frac{r^2}{s^c}) = O(\frac{1}{s^{c-2}}).$$

#### Arrangement

Given a set N of hyperplane in  $\mathbb{R}^d$ , the arrangement G(N) formed by N is the natural partition of  $\mathbb{R}^d$  by N into faces of varying dimensions together with the adjacencies among them.

- A face of j dimensions is called a j-face
- A *d*-face is called a cell
- A (d-1)-face is called a facet
- A 1-face is called an edge
- A 0-face is called a vertex

# General Position Assumption

- No two hyperplane are parallel to each other
- For  $2 \leq j \leq d+1$ , the intersection among j hyperplane is exactly a (d+1-j)-face

## Arrangement in the plane

An arrangement of n lines is one of the simplest geometric structure

•  $O(n^2)$  faces in total



### Facial lattice of an arrangement

- The lattice contains a node for each face of G(N)
- Each node contains auxiliary information, such as pointers to the hyperplanes containing the corresponding face
- A node for a j-face f is linked to a node for a (j 1)-face g if f and g are adjacent



#### Fact

Cells of an arragement of lines in the plane does not allows the random sampling technique

• When all lines in N are tangent to the same circle, for any subset R of N, the central cell of the arragnement of R is intersected by all lines in  $N \setminus R$ .



A cell of an arrangement G(R) does not satisfied the *bounded* degree property.

That is, the collection of cells is not a configuration space.



H(R): the vertical trapezoidal decomposition of G(R)

### **Bounded Valence**

A configuration space  $\Pi(N)$  is said to have *bounded valence* if the number of configurations in  $\Pi(N)$  sharing the same trigger sets is bounded by a constant

# General Form for Main Theorem

Given a set N of n objects, a configuration space  $\Pi(N)$  of N with bounded valance, and the maximum degree d of a configuration in  $\Pi(N)$ , for any random sample R of N of size r, with probability greater than 1/2, the conflict size for each active configurations over R is at most  $c(n/r) \log r$ , where c is a large enough constant.

More generally, fixed any c > d, for any  $s \ge r > d$ , with probability 1- $O(1/s^{c-d})$ , the conflict size of each active configuration over R is less than  $c(n\log s)/(r-d)$ 

#### Sketch of Proof:

For the same reasoning, we have the following fact.

#### Fact

The probability that some  $\sigma \in \Pi(N)$ , with  $l(\sigma) \geq c(\ln s)/(r-d)$ , is active over a random sample R is bounded by  $E[\pi(R)]/s^c$ , where  $\pi(R)$  is the number of configurations in  $\Pi(N)$  whose defining objects are in R.

# $\pi(R) = O(r^d)$

- For each  $b \leq d$ , there are at most  $\binom{r}{b} \leq r^b$  trigger sets contained in R
- Since  $\Pi(N)$  has bounded valence, only a constant number of configurations in  $\Pi(N)$  share the same trigger set.

$$\frac{E[\pi(R)]}{s^c} = O(\frac{r^d}{s^c}) = O(\frac{1}{s^{c-d}}).$$