The design of algorithms for the production of training data

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Optimal complexity

All optimal algorithms have the same distribution on the number of queries. The generating function of this distribution

$$P(z,u) = \sum_{ ext{partition } p} u^{ ext{queries}(p)} rac{z^{|p|}}{|p|!}$$

is characterized by P(0, u) = 1 and

$$\partial_z P(z,u) = P(zu,u)e^{zu}.$$

Asymptotic optimal average query number and standard deviation

$$E \sim rac{\binom{n}{2}}{\log(n)}, \quad \sigma \sim rac{n^{3/2}}{\sqrt{\log(n)}}.$$

We expect a Gaussian limit law.

Proof

Analysis of a particular chordal algorithm.

Clique Algorithm.

- Ask queries between the largest vertex and all others
- The block of the largest label is now discovered
- Remove this block and continue with the remaining vertices.

A partition is then decomposed as a pair

(partition, block of the largest vertex).

Example: $\{\{1,3\},\{4\},\{2,5,6\}\} \mapsto (\{\{1,3\},\{4\}\}, \{2,5\})$

Proof (exact results)

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Symbolic method:

$$P(z):=\sum_n B_n \frac{z^n}{n!},$$

$$\partial_z P(z) = P(z)e^z$$

or recurrence

$$B_{n+1} = \sum_{k=0}^n \binom{n}{k} B_k.$$

In fact, we know $P(z) = e^{\exp(z)-1}$.

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Symbolic method:

$$P(z, u) := \sum_{p} u^{\operatorname{queries}(p)} \frac{z^{|p|}}{|p|!}, \qquad \partial_z P(z, u) = P(zu, u) e^{zu}.$$

or recurrence

$$B_{n+1,q} = \sum_{k=0}^{n} \binom{n}{k} B_{k,q-n}.$$

In fact, we know $P(z) = e^{\exp(z)-1}$. An addtional variable is added to mark the query number.

Proof (asymptotics)

The asymptotics number of partitions is classicaly obtained using a saddle-point method on the Cauchy integral representation of the coefficient extraction

$$B_n = n![z^n]e^{\exp(z)-1} = \frac{n!}{2i\pi} \oint e^{\exp(z)-1} \frac{dz}{z^{n+1}}$$

Probability generating function

$$\mathsf{PGF}_n(u) := \sum_{q \ge 0} \mathbb{P}(q \text{ queries } | |p| = n)u^q = \frac{n!}{B_n}[z^n]P(z,u).$$

Expected number of queries

$$E_n = \partial_{u=1} \operatorname{PGF}_n(u) = \frac{n!}{B_n} [z^n] \partial_{u=1} P(z, u)$$

Solve the differential equation in $\partial_{u=1}P(z, u)$

 $\overline{\partial_z \partial_{u=1} P(z, u)} = \partial_{u=1} (P(z, u) e^{uz}) = (\partial_{u=1} P(z, u)) e^z + z P(z) e^z.$

Characterize optimal active clustering algorithms for non-uniform distributions on set partition.

If the queries are chosen randomly (but not trivial), what is their asymptotic expected number?

Given a (non-chordal) agregated graph, how hard is it to find the optimal queries? (NP-complete?)

Best use of a heuristic distance between elements.