

# Glossary

This glossary provides explanations of how particular technical terms are used in the context of statistical confidentiality. Terms generally used in broad fields of statistics, information technology or empirical research are not included. Nor do we include broad terms such as privacy and confidentiality, which are themselves major subjects of the whole of this book.

**Additive table** A contingency table in which the interior cells sum to the associated marginal totals. Tables can become non-additive when some perturbative disclosure limitation techniques are applied to them.

**Aggregate data** Data which summarize information across a population (or sample). Summary statistics, frequency tables, and magnitude tables are all examples of aggregate data.

**Analysis server** An Internet or LAN-based remote access system enabling the analysis of data sets to be conducted remotely.

**Analytical completeness** The extent to which a disclosure-limited data release allows the same analyses to be conducted as prior to the disclosure limitation being applied.

**Analytical validity** The extent to which a disclosure-limited data release leads to the same statistical inferences as prior to the disclosure limitation being applied.

**Anonymization** A process to make practically impossible the identification of subjects in a database, that is, a process of making the risk of identity disclosure negligible.

**A priori knowledge** Knowledge that a data snooper has which would facilitate attempts to disclose information about individual population units. This might be in the form of pre-existing knowledge about the population units, knowledge of the disclosure control process that has been applied to the data set, or response knowledge.

*Synonym: External knowledge*

**Attribute suppression** Statistical disclosure limitation by eliminating all or some values of one or more attributes.

**Attribution** The association of information in a data set with a particular population unit.

**Attribute disclosure** The disclosure of information about a **population unit** without (necessarily) the identification of that population unit within a data set. This typically refers to aggregate data where something can be inferred about individuals who possess a certain combination of characteristics.

**(Disclosure) Auditing** The process of checking whether processed data are protected or not. When the data are tabular, then this process is done by computing the minimum and maximum value of all the risky cells. On large tables the computation of these two extreme values may be through integer linear programming.

**Cell perturbation** A statistical disclosure limitation method for tabular data whereby the values of some (or all) cells are replaced by some value in an interval.

**Cell suppression** A disclosure limitation method for tabular data whereby the values of particular cells are not released.

**Confidentiality promise** An assurance given to respondents that their data, once provided, will not be disseminated in a form in which they are identifiable or through which information can be learnt about them.

**Confidentiality breach** A situation where a confidentiality promise is broken, especially when a data snooper has actually identified a subject in the database and made use of this information.

**Consent** The process through which a respondent gives permission for data about them, collected by a data stewardship organization, to be used in a given way.

**Controlled rounding** A disclosure limitation method for tabular data whereby the values of all cells are replaced by a value in a finite set. Typically, for each cell, this set contains only two values, which are the numbers rounding up and down the original cell value to a multiple of a given base number (for example, 10). Marginal totals are maintained.

**Controlled tabular adjustment (CTA)** A disclosure limitation method for tabular data whereby the values of all cells are replaced by different values. It is a particular method of cell perturbation. CTA adjusts cell values to satisfy disclosure limitation ranges and tabular constraints, such as additivity, while minimizing data loss as measured by some linear measure of overall data distortion, such as the sum of the absolute values of the individual cell value adjustments. CTA replaces each risky cell by either of the two endpoints of its protection range. Certain non-risky cell values are adjusted by small amounts to restore additivity.

**Cyclic perturbation** A disclosure limitation method for tabular data whereby the values of all cells are replaced by different values. It is a particular technique of cell perturbation in which patterned collections of four or more cells, called data cycles, are randomly modified with some cell values increasing by one and others decreasing by one, in such a way that each row and column sum is undisturbed.

**Data dissemination** Any process through which (access to) a set of data are given to data users.

**Data divergence** The difference either between two data sets or a data set and the world in how information about a given population unit is recorded. This can be caused by coding differences, response errors, data aging, data entry errors, and other forms of misclassification.

**Data protection** A broad term which covers a set of legal and ethical principles and instruments which collectively delineate the ways in which data may be fairly processed by data stewardship organizations. Data protection operates at national and international levels.

**Data snooper** An individual, group, or organization that seeks to identify individual population units within a data set and/or discover information about a population unit, usually through a statistical linkage process of information already known to information contained within the data set.

**Data Stewardship Organisation (DSO)** An organization which captures, processes, holds, and/or disseminates data about population units. A DSO is under legal or ethical obligations to maintain confidentiality and provide data releases of high utility to users.

**Data swapping** An SDL method in which certain attribute values for some records in the source data are exchanged. For example, a sample of households may be selected and matched on a set of selected key variables with households in nearby geographic areas that have similar characteristics (such as the same number of adults and same number of children). Values of an attribute such as household income may be swapped.

**Data user** Any person using a data set for a bona fide, typically statistical, purpose.

**Data utility** The usability or research value of a given set of data. In the context of disclosure-limited data, high data utility requires both analytical completeness and analytical validity.

**Data-world divergence** The difference between a data set and reality. Sources of data divergence include data outdated, response errors, coding or data entry errors, differences in coding, and the effects of disclosure limitation.

**Deidentification** The removal of direct identifiers from a data set or individual record.

**Direct identifiers** Information that can identify a data subject directly, such as name or social security number. Also includes combinations of attribute values, such as address and age, that uniquely identify a data subject.

**Disclosure auditing** See auditing.

**External knowledge** Knowledge that a data snooper may have that is external to that in a particular data release.

**Global recoding** Any method of recoding the categories in a whole data set to a smaller set of categories.

**Key variables** A combination of attributes that increase disclosure risk because a data snooper may link values to those in an identified external data set.

**Hierarchical data** Used particularly to describe microdata which has grouping variables for two or more levels of analysis. The vast majority of microdata is geographically hierarchical. Another variant is hierarchical household data which contain all members of a set of sampled households.

**Identification** The association of a population unit whose identity is known with a particular microdata record.

**Identification file** A database, data set, or other form of data storage which contains formal identifiers for individual population units.

**Identification risk** The probability that a data snooper can identify a data subject in a released data product.

**Identity disclosure** Identity disclosure occurs when a data subject is identified from released data.

**Integer linear programming** Mathematical tools for optimization. Typically there is a linear objective function that must be maximized or minimized by choosing values of specified variables. These variables are subject to linear systems of equalities and/or inequalities with integer variables and possibly continuous variables.

**Interval publication** A disclosure limitation method for tabular data whereby the values of particular cells are replaced by intervals of values. For each value in each interval there should exist values in the other intervals such that together with the values which have not been replaced the resulting table is coherent. Each interval should contain the original value that it is replacing.

**Linking** The process of matching a data subject in released data to an identified subject in an external database.

**Local suppression** A disclosure limitation method where particular data units within a microdata file are coded as missing. This method is used within the ARGUS system.

**Lower bound** The minimum possible value of a given cell in a table of aggregate values which has been perturbed so that the exact value is not given (and possibly is even suppressed).

**Macrodata** Summary data, most especially data released in the form of a contingency table, hence with categorical attributes.

**Magnitude tables** Tabular data where each cell value has been obtained by adding the response value of all the contributors that fit within the categories of the key variables describing such cell. The response variable may be continuous or integer numbers.

**Marginal total** A cell value that has been computed by summing other cell values.

**Matrix masking** SDL methods that transform an  $n \times p$  (cases by variables) data matrix  $Z$  through pre- and post-multiplication and the possible addition of noise.

**Masking** Any SDL method that prepares data for release by stochastic or deterministic transformation of the source data.

**Mathematical programming** An area of mathematics devoted to study mathematical models and algorithms to solve optimization problems. Integer linear programming is a sub-area, but there are other areas like convex optimization and quadratic programming.

**Microaggregation** An SDL method in which records are aggregated into groups. Instead of releasing the actual values for individual records, the mean (typically) of the group is released. Confidentiality is protected by each group having at least a minimum number of observations.

**Microdata** Data composed of records on individual data subjects. Each record might refer to an individual person, household, business, or other entity. The data may be directly collected for statistical purposes or obtained from other sources, such as administrative sources.

**Minimal sample unique** A combination of values within a data set that is unique within the data set and for which no subset of values is also unique.

**Multivariate additive noise** A statistical disclosure limitation method in which the released data are created by adding a random vector of disturbances to the source data. The multivariate distribution of the random vector is chosen to lower disclosure risk while maintaining adequate data utility.

**Noise addition** A statistical disclosure limitation method in which the released data are created by adding a random disturbance to some or all values in the source data.

**$n$  rule** A procedure to detect the risky cells (those requiring protection) in a table, so a procedure to solve the so-called primary problem. This procedure is a simple mechanism that takes into account only the number of contributors to each cell. Given a threshold value  $n$ , a cell is classified as “risky” when the number of contributors to this cell is less than or equal to  $n$ .

**Parallel divergence** A situation where two data sets both contain the same value for a given population unit/variable but both differ from a veridical representation of that population unit/variable. For example, somebody who

lies to the tax collector about their income may well also lie to the census agency.

**Perturbation** Any disclosure limitation process where values for given data units are changed either systematically or randomly.

**Primary problem** Checking whether there is something to protect or not in a dataset. When there is something that needs protection, it must be individuated. For example, given a table, the primary problem is the problem of finding the cells (if any exist) that will need protection. The cells individuated by solving the Primary Problem are called “Primary Cells” (or risky cells). The secondary problem is protecting the risky cells.

**Poisson model** A model based on the assumption that the counts in a contingency table are Poisson distributed.

**Population uniqueness** The proportion of population units, within a given population, which has a unique combination of values, for a given set of variables.

**Population unit** A socioeconomic entity about which data may be collected and which is a member of a set of such entities. The set is referred to as the population. Entities may be individuals, households, families, businesses, or other organizations.

***p/q* rule** It is a procedure to detect which cells require protection in a table. In other words, it is a procedure to solve the so-called primary problem. This procedure is a mechanism more elaborated than the so-called *N* rule. In particular, the *p/q* rule takes into account the number of contributors to each cell and also the value of the major contributors.

**Post-tabular SDL** Statistical disclosure limitation processes applied to dataset once it has been aggregated into tabular format.

**Pre-tabular SDL** Statistical disclosure limitation processes applied to dataset prior to it being aggregated into tabular format.

**Primary suppressions** The cells individuated by a rule to solve the Primary Problem on a table. They are called suppressions when the method to solve the Secondary Problem is the cell suppression technique.

**Probabilistic record linkage** Record linkage processes where non-exact linkages are allowed.

**Random rounding** A statistical disclosure limitation method for tabular data that rounds each cell of a table up or down to a multiple of a given base number (say 5). It makes use of a probability function to decide the direction of the rounding for each value. For example, a cell value *i* will go down with probability  $(1-i/5)$  and up with probability  $i/5$ . The main advantages of random rounding are the simplicity of the approach and the unbiased feature of the resulting table. The main disadvantage is that the resulting table may not be additive when the marginal cells are also modified with the same approach.

**Random unique** A record which is not special unique on a given set of variables.

**Record linkage** The process of linking records on different data sets. This can be done by researchers for bona fide analytical reasons or by a data snooper attempting to link identification information to an anonymized data set.

**Record suppression** A disclosure control process whereby whole records are removed from a data set before dissemination (or, in the cases of samples from census data, deliberately not sampled).

**Reidentification** The process of determining the identity of data subjects in data releases that have been subject to statistical disclosure limitation, that is, have been deidentified.

**Research Data Center** A physical facility maintained by a data stewardship organization where data users may make use of the data for statistical purposes under restricted access conditions.

**Response knowledge** Knowledge that a particular population unit is represented in a given data set. This usually is in the form of knowledge that the population unit responded to the survey. Response knowledge in respect of a sample survey data set can greatly increase the risk of disclosure.

**Restricted access** An approach to disclosure limitation in which data access is controlled by the data stewardship organization. Access controls can be physical, technical or administrative.

**Restricted data** An approach to disclosure limitation in which the data are transformed, reduced, or otherwise masked and could be used on its own or in combination with restricted access.

**Risk metrics** Numerical indices (possibly estimated probabilities of identification or attribution) designed to indicate the likelihood of disclosure events under a given set of assumptions.

**Risky cells** Cells in a table that have been deemed to have high disclosure risk.

**Rounding** A method of disclosure limitation whereby values are rounded to a particular base. This is usually applied to frequencies within a table of counts.

**R-U confidentiality map** A graph of the trajectory in (disclosure risk, data utility) space as the extent of statistical disclosure protection of a given method increases.

**Safe setting** A form of restricted access where the data analysis environment is highly controlled. This will typically be within the offices of the data stewardship organization or in satellite offices, such as research data centers, where it has assurance that confidentiality standards are maintained.

**Sample uniqueness** The proportion of records, within a sample data set, which have a unique combination of values, for a given set of variables.

**Sensitive cells** See risky cells

**Secondary problem** The problem that must be solved after the Primary Problem has been solved and has generated a set of risky cells. The Secondary Problem consists of protecting the risky cells. The way to proceed depends on the chosen method (like cell suppression and controlled rounding).

**Secondary suppressions** Used in disclosure limitation for aggregate data these are cell suppressions that are necessary in order to stop a data snooper (or data user) from recovering the original cell values. They are the output of the Secondary Problem when the chosen method is cell suppression. “*Complementary suppressions*” is a synonym.

**Sensitivity** A measure, using qualitative and often subjective, of the damage that would be caused (usually to the respondent concerned) if a given piece of information is disclosed about either a given population unit or an imaginary “typical” population unit.

**Social acceptability** A conceptual component of sensitivity which considers how responses to surveys might vary in their compliance with social norms and expectations. Consider, for example, a survey of sexual practices or attitudes toward political extremism.

**Special unique** A record within a microdata set which is sample unique on a given set of variables and also on a subset of those variables.

**Special unique detection algorithm** A method for calculating and grading the disclosure risk for individual-level categorical microdata by determining the number and scale of special unique combinations of variables for each record within a data set within some defined key.

**Statistical disclosure** The process by which information is inferred about individual population units.

**Statistical disclosure limitation (SDL)** The process of lowering the disclosure risk of data releases intended for statistical analysis purposes. “*Statistical disclosure control*” is a synonym.

**Subtraction** A method of attacking a table-aggregated dataset (typically a table of counts) by removing population units for which data are already known for the dimensions of the table and thereby reducing the table to a residual, possibly disclosive, form.

**Swapping key** A set of key variables used for pairing records in record swapping disclosure limitation regimes.

**Swapping partner** Records within a data swapping system which are paired on the basis of having the same (or possibly similar) values on a swapping key.

**Synthetic data** Data to be released which have been probabilistically generated based on a model estimated from the source data.



**Target variables** Variables within a data set deemed likely to be of interest to an intruder. Target variables will generally be (i) unavailable to an intruder and (ii) sensitive in some way.

**Topcoding** An SDL method used with interval-scale or many-category ordinal variables where values above a certain threshold are revealed only in aggregate form. Age and income often are topcoded, with only an indication that the values are above the threshold.

**Unrestricted access** A data dissemination process whereby no limits are placed on access nor is such access monitored. This now is mainly used for the delivery of aggregate statistics over the web.

**Upper bound** The maximum possible value of a given cell in a table of aggregate values which has been perturbed so that the exact value is not given (and possibly is suppressed).



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