

Nested case-control and case-subcohort studies

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Studies within prospective cohort studies

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- Case-subcohort studies

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Considerations for their design and analysis.

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A problem might be that it is not feasible to collect data on the additional exposures on all individuals, especially if the original cohort is large, due to cost, time, not wanting to use up existing biological samples.

Even if it is feasible, it may not be necessary.

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The underlying cohort need not be an assembled cohort of individuals participating in a study but may be any well defined sampling frame, such as a country-wide health registry.

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- how to treat individuals who leave/join the study population
- what to do about exposures and other explanatory variables that may change appreciably over time

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Why not use as controls a group of individuals who survive to the end of the observation period?

- if the observation period had been shorter then, for cases occurring early in the observation time period, the pool of potential controls would be different from that if the observation period had been longer
- concerns whether individuals who survive over a long time period without being censored for some reason intrinsically different from the rest of the underlying population of non-cases

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Risk set: the set of all individuals still in the study at that time (still eligible to experience and have the event observed at that time, that is, still event-free and not censored).

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Individuals can be sampled as controls for more than one case and individuals sampled as controls may subsequently become cases.

Nested case-control studies

Risk sets differ depending on whether time since entry into the study or age is used as the underlying time scale

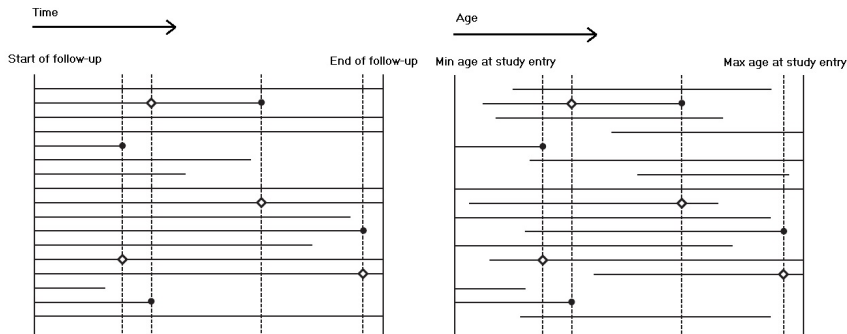


Figure: Sampling a nested case-control study with one control per case, using time since study recruitment or age as the timescale. The solid lines represent the time period over which individuals are observed. Cases occur over the course of follow-up (●), individuals may leave the population or may survive to the end of the follow-up period or maximum age of observation. The dotted lines pass through members of the risk set at each event time. One control (○) is selected for each case from its risk set. Adapted from Keogh and Cox (2014).

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- Can accommodate **time-varying exposures** and **recurrent events**
- Other survival models are possible (e.g. fully parametric models, additive hazards models)

Example

Pai et al. (2004)

Nurses' Health Study and Health Professionals Follow-up Study

Large US prospective cohort studies

Blood samples collected a few years after recruitment

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At each case's event time 2 controls were randomly sampled from the risk set, excluding the case, matched to the case on age, smoking status, date of blood sample and, in the male study only, fasting status at the time of the blood sample.

Function `ccwc` (by David Clayton) in package `Epi` (Carstensen et al., 2017) samples controls for a nested case-control study

Function `clogit` (by Thomas Lumley) in package `survival` (Therneau, 2015) fits a conditional logistic regression model

```
ccwc(entry = 0, exit, fail, origin = 0, controls  
      = 1, match = list(), include = list(), data  
      = NULL, silent = FALSE)
```

entry: Time of entry to follow-up

exit: Time of exit from follow-up

fail: Status on exit (1 = fail, 0 = censored)

origin: Origin of analysis time scale

controls: The number of controls to be selected for each case

match: List of categorical variables on which to match cases and controls

include: List of other variables to be carried across into the case-control study

data: Data frame in which to look for input variables

silent: If FALSE, echos a . to the screen for each case-control set created; otherwise produces no output.

In R

Fitting a conditional logistic regression model

```
clogit(formula, data, weights, subset, na.action
, method = c("exact", "approximate", "efron",
"breslow"), ...)
```

formula: Model formula

data: data frame

weights: optional, names the variable containing case weights

subset: optional, subset the data

na.action: optional na.action argument. By default the global option na.action is used.

method: use the correct (exact) calculation in the conditional likelihood or one of the approximations

...: optional arguments, which will be passed to `coxph.control`

Can we do better?

- in some cases efficiency can be increased using techniques such as **countermatching** and **quota sampling** (also **backwards and forwards use of controls**)
- in some cases multiple imputation can be used to utilise information on the full cohort

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What if we want to study another outcome in the same study?

- methods for re-using controls for other outcomes exist
- however for multiple outcomes a more natural choice is to use a **case-subcohort design**

Case-subcohort studies

Also called *case-cohort* studies

- Disregarding times – *case-base sampling* (or *hybrid epidemiological design*)
- Using times

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The primary feature of a case-subcohort study is the **subcohort**, which is a random sample from the cohort at study baseline, selected ignoring any information obtained during follow-up, and which serves as the set of potential controls for all cases.

The study comprises the subcohort plus all additional cases, that is, those not in the subcohort.

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Case-subcohort studies

Disregarding time

- Time period, preferably short, during which cases are observed
- Assumes that individuals who do not become cases are observed for the entire time period
- Odds ratios and risk ratios, logistic regression

Case-subcohort studies

Using time

From now on we consider the analysis using time.

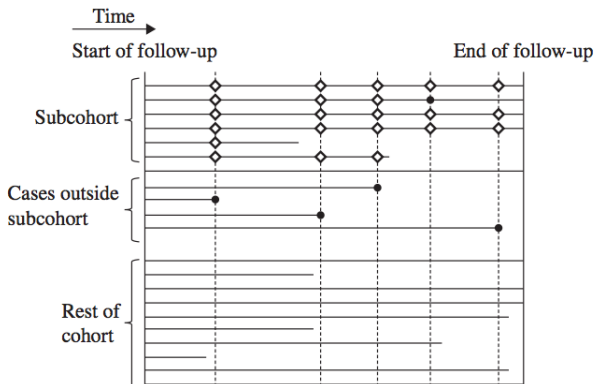


Figure: A case-subcohort study using event times. The horizontal lines show individual follow-up. The cases are indicated by ● and the non-cases used in the comparison set at each failure time are indicated by ◊. From Keogh and Cox (2014).

Case-subcohort studies

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- The cases are compared with members of the subcohort who are at risk at their event time, using a **pseudo-partial likelihood** (also called **weighted Cox regression**) – estimates of hazard or rate ratios

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- **Sandwich** standard errors (correlations between risk sets)

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- The cases are compared with members of the subcohort who are at risk at their event time, using a **pseudo-partial likelihood** (also called **weighted Cox regression**) – estimates of hazard or rate ratios
- **Sandwich** standard errors (correlations between risk sets)
- Variations of this method exist, differing mainly on how risk sets are formed (how the cases outside the subcohort are treated) and on weighting

- Can use the subcohort for assessing associations with other measures collected at study baseline

Case-subcohort studies

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Function `cch` (by Norman Breslow, modified by Thomas Lumley) in package `survival` (Therneau, 2015) fits weighted Cox regression models for case-subcohort studies

In R

Fitting a weighted Cox regression model to case-subcohort data

```
cch(formula, data = sys.parent(), subcoh, id,
     stratum = NULL, cohort.size, method = c("
     Prentice", "SelfPrentice", "LinYing", "I.
     Borgan", "II.Borgan"), robust = FALSE)
```

formula: A formula object that must have a **Surv** object as the response. The **Surv** object must be of type "right", or of type "counting".

subcoh: Vector of indicators for subjects sampled as part of the sub-cohort. Code 1 or TRUE for members of the sub-cohort, 0 or FALSE for others. If **data** is a data frame then **subcoh** may be a one-sided formula.

id: Vector of unique identifiers, or formula specifying such a vector.

stratum: A vector of stratum indicators or a formula specifying such a vector

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```

cohort.size: Vector with size of each stratum in the original cohort from which the subcohort was sampled

data: An optional data frame in which to interpret the variables occurring in the formula.

method: Three procedures are available. The default method is "Prentice", with options for "SelfPrentice" or "LinYing".

robust: For "LinYing" only, if **robust** = TRUE, use design-based standard errors even for phase I

The data argument must not have missing values for any variables in the model.

Can we do better?

- **stratified sampling** (for example when some strata are rare in the population, but in some cases restricts reusability), **stratified analysis** (each case is compared with individuals in the same stratum of the subcohort as the case – hazards proportional within strata) or both
- in some cases so-called 'optimal' weights can be used to utilise information on the full cohort
- in some cases multiple imputation can be used to utilise information on the full cohort

Example

Associations of metabolomics and proteomics with risk of pancreatic cancer and diabetes in the China Kadoorie Biobank

China Kadoorie Biobank (CKB) – prospective study of 0.5M Chinese adults

- blood samples collected at baseline (2004–08)
- linkage to health insurance records, disease and death registries

700 pancreatic cancer incident cases accumulated by 2015, as well as a few thousand diabetes cases

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Want to use NMR metabolomics assay and proteomics assay on samples collected at baseline to identify markers associated with

1. pancreatic cancer
2. diabetes

and possibly several other diseases in the future

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Case-subcohort study

- all 700 pancreatic cancer cases
- a subset of 1000 diabetes cases
- a subcohort of size 1050 (randomly selected from the cohort at baseline)

Example

Armstrong et al. (1994)

Case-subcohort study to investigate whether aluminium plant workers exposed to coal tar pitch volatiles were at increased risk of death from lung cancer.

Source population: over 16,000 men who had worked for at least one year in an aluminium production plant between 1950 and 1979 in Quebec, Canada.

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338 lung cancer deaths observed during follow-up (from 1950 to 1988)

Subcohort: random sample of 1138 men

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Time scale: age








Each case was compared with members of the subcohort still alive at the case's age. This required estimation of the cumulative exposures for all at-risk subcohort members at each case's age at death.

- Two broad types of sub-studies within prospective cohort studies: nested case-control and case-subcohort studies
- Careful design to ensure reliable results

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