

#### **Self-controlled Case Series (SCCS)**





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#### **Acknowledgements**

Thanks to ALL the people I have worked with on SCCS in the last 15 years

Special thanks to:

Paddy Farrington

Heather Whitaker

Ian Douglas



## **Today**

- Overview of SCCS method
- Example
- Basic assumptions/features of SCCS
- Indication bias



#### **Observational studies**

#### **Cohort studies**

- Compare risks of events between exposed and non exposed individuals
- Event random, exposure fixed
- Poisson regression
  - Condition on time

#### Case control studies

- Compare exposure between cases and controls
- Exposure random, event fixed
- Logistic regression

#### What if....

- We only have people who have the event, but we don't know the exact denominator
  - Hospital data
  - Long-haul flights and thromboembolism (VTE)
- There is a substantial difference between exposed and unexposed individuals
  - Ethnicity
  - Deprivation
  - Indication bias (severe asthma and influenza vaccine)

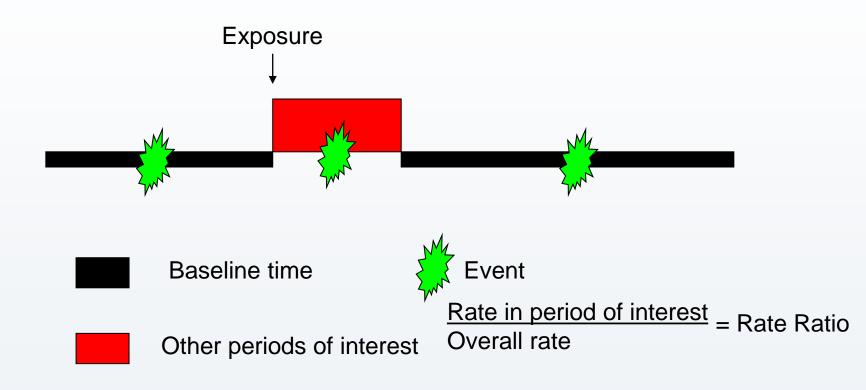


# Self-controlled case series may be the answer?

- Case series analysis only cases are sampled
- Estimation is within individuals rather than between individuals

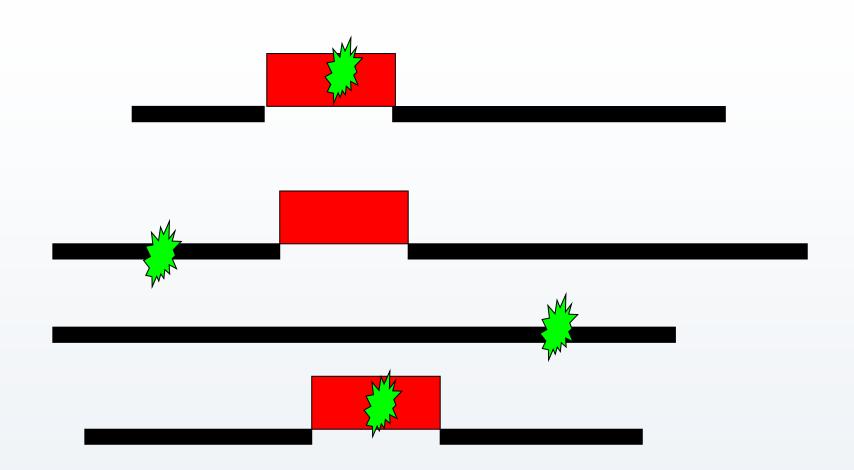


It is a conditional cohort method: exposures are regarded as fixed, event times as random



Farrington CP. Biometrics 1995;51:228-235

# LUCL





# Other key features of case series method

- Follow-up is not censored at event
- Can be used with:
  - independent recurrent events
  - uncommon non-recurrent events
- The analysis is self-matched, thus eliminating the effect of fixed confounders



### Fitting the case series model

#### Conditional Poisson model

A:  $exp(\Phi i) = Baseline incidence of condition$ 

B:  $exp(\alpha 1) = Relative$  incidence exposure group k

We want an estimate of B ©

We can't measure A with SCCS



#### Case series

- We know that all had the event
- Overall Poisson Rate = 1
- Estimate conditional likelihood that event happened in specific period



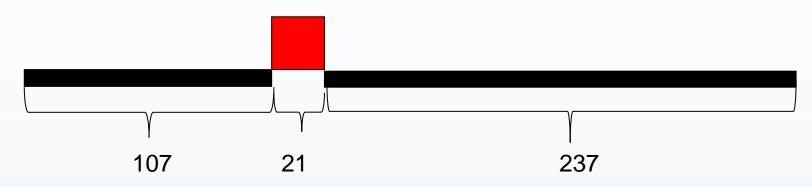
# **Example: MMR vaccine and Viral meningitis example**

- Meningitis in second year of life
- 1 Oct 1988 and 31 December 1991

10 children with viral meningitis

Don't know how many were vaccinated

#### MMR vaccine and viral meningitis



Three exposure periods:

1: 107 days 2: 21 days 3: 237 days

Estimate likelihood of event in each of those periods:

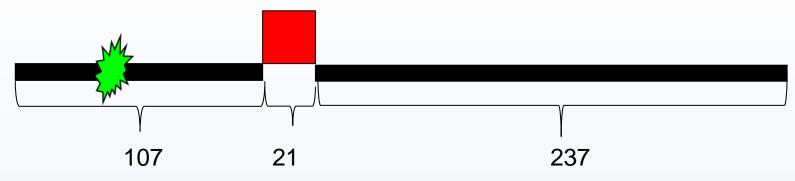
$$\lambda 1 = 107A$$
  $\lambda 2 = 21AB$   $\lambda 3 = 237A$ 

Overall Poisson rate for child 1:

$$\Lambda = 107A + 21AB + 237A$$



#### MMR vaccine and viral meningitis



Child 1 had event in first period - estimate conditional likelihood of event in this period:

$$\lambda 1/\Lambda = 107A/(107A + 21AB + 237A)$$

A is cancelled out = self-controlled

$$\lambda 1/\Lambda = 107/(344 + 21B)$$

Log  $(\lambda 1/\Lambda) = \log(107/(344 + 21B))$  and repeat for other children

### Baseline risk of event may change over time

- Risk of event may change with age
- Introduce different risk periods

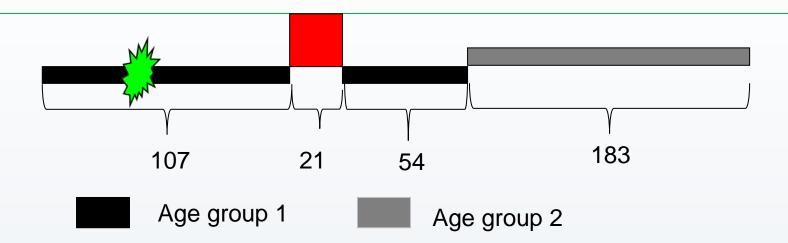
A:  $exp(\Phi i) = baseline incidence of viral meningitis$ 

B:  $exp(\alpha_1)$  = relative incidence exposure group k

C:  $exp(\beta 1) = relative incidence age group j$ 

## **L**

## MMR vaccine and viral meningitis



- Four exposure periods:
- 1: 107 days 2: 21 days 3: 54 days 4: 183 days
- Estimate likelihood of event in those periods
- $\lambda 1 = 107A$ ,  $\lambda 2 = 21AB$ ,  $\lambda 3 = 54A$ ,  $\lambda 4 = 183AC$

#### Multinormial likelihood for all children

- Repeat for other children as for child 1
- log and estimate log-likelihood
- β1: -1.491 α1: 2.488

B:  $exp(\alpha 1) = relative incidence exposure group k: 12.037$ 

C:  $exp(\beta 1) = relative incidence age group j: 0.225$ 

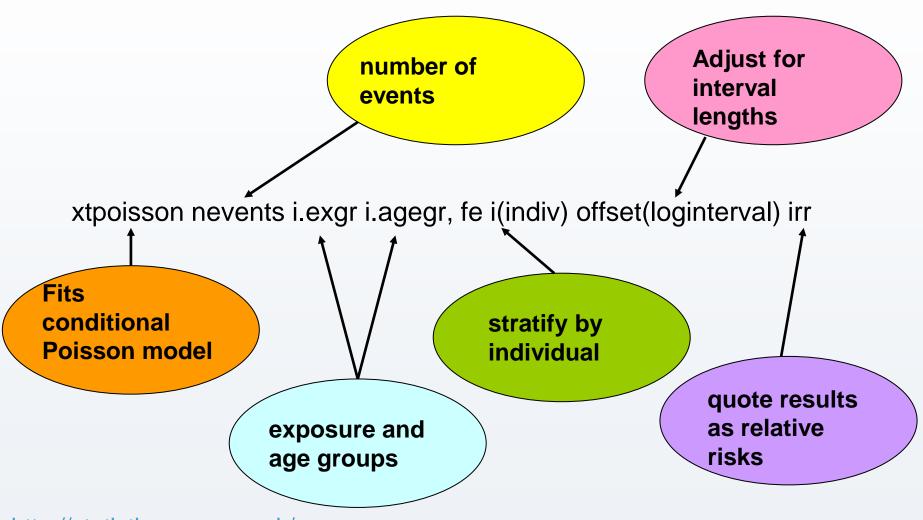
#### **OBS OBS**

A:  $exp(\Phi i) = Baseline incidence of condition - eliminated!$ 

Whitaker et al., Statist. Med. 2006, 25: 1768 – 1797



#### Statistical analysis in STATA



http://statistics.open.ac.uk/sccs



### The assumptions and features of SCCS

- 1. Events must be:
  - Independently recurrent
  - Rare
- 2. Occurrence of an event should *not* (appreciably) affect subsequent exposures
  - Occurrence of an event should not (appreciably) increase mortality
- 3. Events cannot happened at exact same time or age
- 4. Does not produce estimates of absolute incidence



# Assumption 1 - Events must be independently recurrent or rare

- The method was developed for independent recurrent events
- Bias is ignorable if the method is used with non-recurrent events with risk of occurrence < 10% over the observation period
- A test for independence has been developed (Farrington & Hocine, Applied Statistics 2010 59: 457 – 475)
- If events dependent you can use just first event
  - fractures
  - MI



# Assumption 2 - Occurrence of an event should not affect subsequent exposures

 Conditioning on full exposure histories is only valid provided events do not affect subsequent exposures



 In other words, exposure must be an exogenous variable (coming from outside a system)



#### **Direction of bias**

- The direction of bias is predictable:
  - If the event reduces the chance of exposure, the Relative Incidence will be biased upwards
  - If the event increases the chance of exposure, the Relative Incidence will be biased downwards
- This may be helpful to know

### Assumption 2 - cont.

Occurrence of an event should not (appreciably) increase mortality

- The observation period is assumed to be independent of the event
- Failure of the assumption can cause bias, in an unpredictable direction
- However, see Farrington et al, JASA 2011, 106: 417 426



# When can events influence subsequent exposures?

- Occurrence of an event may delay exposure:
  - Treatment (eg vaccination) may be deferred until recovery
- The event may be a contra-indication for treatment:
  - For example, intussusception and rotavirus vaccine
- If the event is death:
  - No subsequent observation can occur

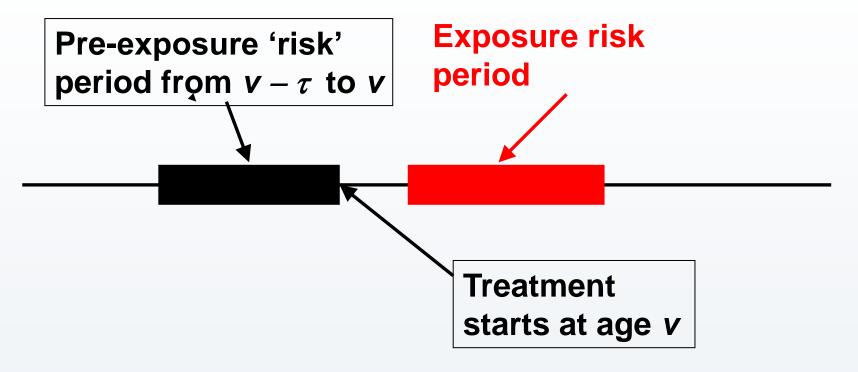


## **Short-term impact**

- The event may only temporarily delay exposure
- This will result in a deficit of events in that period
- The RI will be biased upwards
- This can be corrected by including a pre-exposure 'risk' period



# Including a pre-exposure 'risk' period



Under suitable conditions, the trick of including a preexposure 'risk' window will correct the bias in the *RI* 

Farrington & Whitaker 2006, Applied Statistics



#### Longer term dependence

- The pre-exposure 'risk' window trick only works if t is relatively short
- If it is not then only post-treatment time can be used



# Indication bias - Asthma exacerbation and flu vaccine

- Cohort and case series studies in asthmatic children
- Aged 1 6 years in 1995/6
- Risk period: 2 weeks after flu vaccine.

• Kramarz et al, Arch. Fam. Med. 2000, 9: 617 – 623



#### Asthma exacerbation and flu vaccine

Method	Sample size	RI	95% <i>CI</i>
Cohort, unadjusted	70 753	3.29	(2.55, 4.15)
Cohort, adjusted	70 753	1.39	(1.08, 1.77)

The cohort results are subject to indication bias? Children with severe asthma more likely to have flu vaccine?

Kramarz et al, Arch. Fam. Med. 2000, 9: 617 – 623



#### Asthma exacerbation and flu vaccine

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Cohort, unadjusted	70 753	3.29	(2.55, 4.15)
Cohort, adjusted	70 753	1.39	(1.08, 1.77)
Case series	2075 cases	0.98	(0.76, 1.27)

The cohort results are subject to indication bias - The case series results are unaffected by this bias.



#### **Summary**

- SCCS builds on the cohort methodology
  - Event is random, exposure is fixed
- Ultimate matching (within individuals)
  - Cancel out all fixed characteristics
- Statistical powerful
- Beware of the assumptions



#### RESEARCH METHODS AND REPORTING





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# Self controlled case series methods: an alternative to standard epidemiological study designs

Irene Petersen,<sup>1,2</sup> Ian Douglas,<sup>3</sup> Heather Whitaker<sup>4</sup>

The self controlled case series (SCCS) method is an epidemiological study design for which individuals act as their own control—ie, comparisons are made within individuals. Hence, only individuals who have experienced an event are included and all time invariant confounding is eliminated. The temporal association between a transient exposure and an event is\_ estimated. SCCS was originally developed for evaluation of vaccine safety, but has since been applied in a range of settings where exact information on the size of the population at risk is lacking or

such as cohort and case-control studies, have been widely applied in medical research (see supplementary web table w1). There are several situations, however, where standard epidemiological study designs fall short. For example, in the research of adverse effects of vaccines it can be difficult to identify suitable comparison groups (eg, if most of the population receives the vaccine). Likewise, studies on hospital data may not have information on the exact catchment areas and hence it is a struggle to find suitable controls for cases of a particular event. In these situations the self controlled case series (SCCS) method provides an alternative epidemiological study design to investigate the association between a transient exposure and an outcome event. The SCCS method is a case only method; it has the advantages that no separate controls are required and any fixed confounder is automatically controlled for.23

In this paper we provide an overview of SCCS methodology and some examples of how the method has been applied, in order to give an idea of the potentials

## **L**

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- Petersen et al. Self controlled case series methods: an alternative to standard epidemiological study designs BMJ 2016

#### Case-crossover approach

- Based on case-control methods
- Exposures in the period immediately preceding the event are compared to exposures at earlier 'control' times
- Assumptions exposure distribution in successive time periods is exchangeable – i.e. cannot account for change in effects of age.
- Maclure M. The case-crossover design: A method for studying transient effects on the risk of acute events AJE 1991: 133(2):144-153
- Greenland S. A unified approach to the analysis if case-distribution (case-only) studies. Statistics in Medicine 1996: 18:1-15.