

Phases of development of Weighted Cumulative Exposure modeling

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Support

Canadian Institutes of Health Research (CIHR) grant PJT-180634
Natural Sciences and Engineering Research Council (NSERC) of Canada grant 228203

Goal of Weighted Cumulative Exposure (WCE) modeling

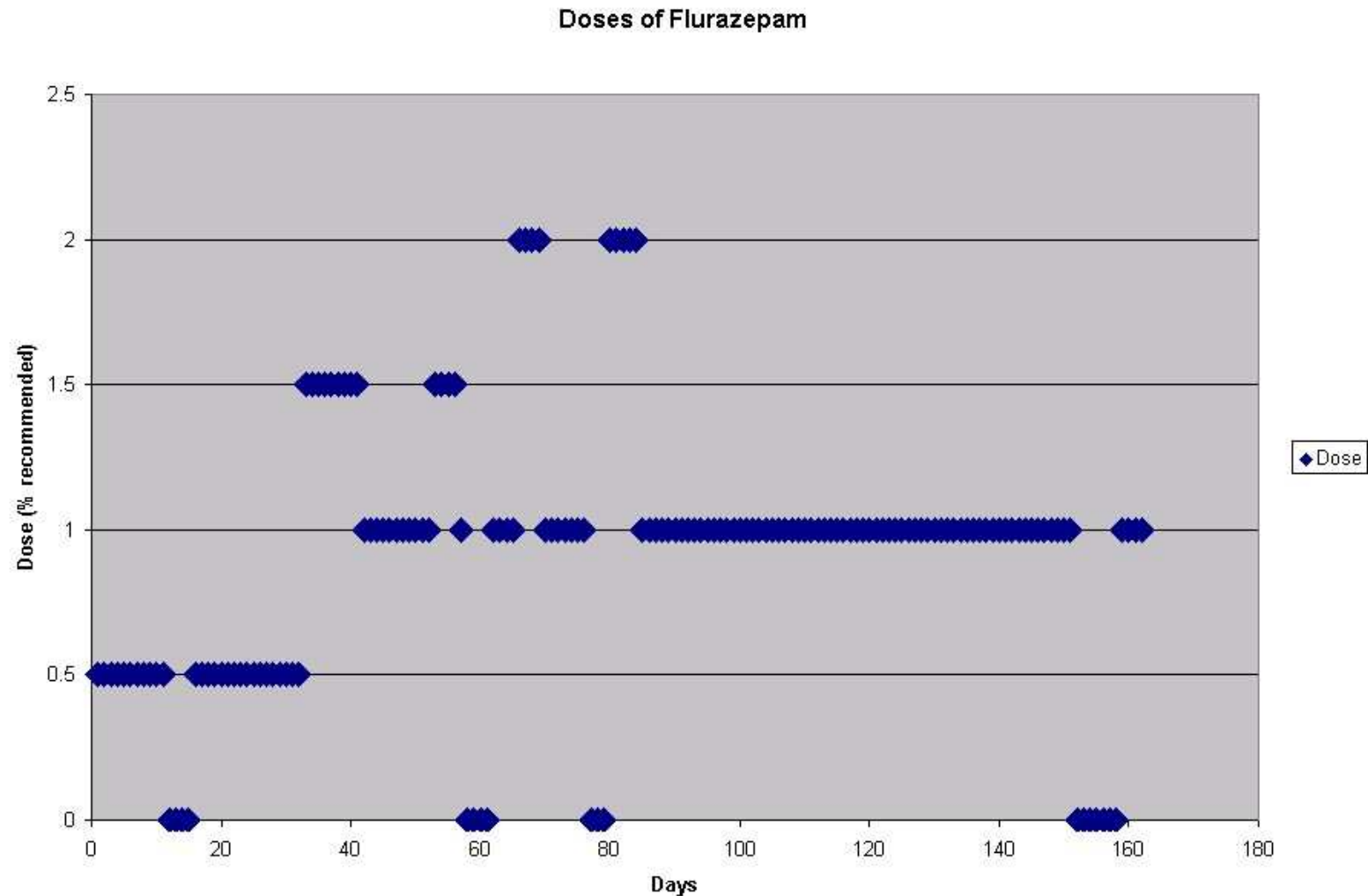
To Assess

**CUMULATIVE Effects of Past Values of
Time-Varying Exposures**

on the current Hazard in Time-to-Event analyses
(focusing mainly on **observational
Pharmaco-epidemiological studies of
Safety or Effectiveness of Medications**)

“Phase 0”: Look at relevant **Real-World Data**
example of a Time-Varying Drug Exposure:

6-months Variations of DAILY DOSE of a Psychotropic Drug (Flurazepam) for a Single Elderly Patient



Phase "I A": Identify the Methodological Challenge

- Challenge in Modeling Time-Varying Exposures (TVE):
how to Assess 'current' Relative Risk (e.g. HR) at time u
as a Function of the History of Past Values [$X(t)$ for $0 < t \leq u$]:
 $HR [u \mid X(1), X(2), \dots, X(T-1), X(u)]$?
- Needs a 2-Step Solution:
 1. Define Time-Varying aggregate covariate $M(u)$ representing
Current Value of an '**Etiologically Correct Exposure Metric**':
 $M(u) = f [X(1), X(2), \dots, X(u-1), X(u)]$
 2. Use appropriate regression methods (e.g., Cox PH model for time-to-event analyses) to
Estimate HR associated with $M(u)$

Phase “I B”: Identify the **NEED for NEW METHOD(s)**

- Most Pharmaco-Epi studies rely on **Arbitrarily chosen, very simple *Ad Hoc* “Conventional” Time-Varying Exposure Metrics**
- * **EXAMPLE** of **Alternative (formally Incompatible!), Arbitrarily chosen Time-Varying Exposure metrics** used in published studies **of the Same Association (Glucocorticoids use vs. risk of Infections in Rheumatoid Arthritis)**:
 - ‘*Current* use’ (Binary)
 - ‘*Recent* use’ (Binary: Any use in *last month* or *last 3 months* or *last 9 months*)
 - ‘*Ever* use’ (Binary: Use at least once during the *entire past follow-up*, **often many yrs!**)
 - ‘Total past dose’ (Continuous, weakly *Monotone Increasing*)

[Franklin J et al, *Ann Rheum Dis* 2007; Lacaille D et al, *Arthritis Rheum* 2008; Smitten AK et al, *J Rheumatol* 2008; Schneeweiss S et al, *Arthritis Rheum* 2007; Bernatsky S et al, *Rheumatology (Oxford)* 2007; Saag KG et al, *Am J Med* 1994]

Phase "I B": (cont-d)

Identify the **NEED** for **NEW METHOD(s)**

- Need to accurately **assess Cumulative Effects of past Drug Use**

[Pazzagli et al, *Pharmacoepidemiol Drug Saf* 2018; WHO, *Organ Tech Rep Series* 1972; Abrahamowicz & Tamblyn, *Enc Biost* 2005]

- **Risks** (and benefits) of using a specific drug likely **depend on** the treatment's: (i) **dose**, (ii) **duration and/or** (iii) its **recency** **

[Edwards & Aronson, *Lancet* 2000; Perucca & Gilliam, *Lancet Neurol* 2012]

**** This implies that conventional Cumulative Dose (Sum of all past doses) or Cumulative Duration of Use may NOT be accurate TVE Metrics!**

Phase "I C": Concept & Formulation of the New **Weighted Cumulative Exposure (WCE) Model**

$$WCE(u) = \sum_{t \leq u} w(u-t) * X(t) \quad (1)$$

u = current time (when Risk is being assessed)

$WCE(u)$ = Weighted **Cumulative Effect** of the Past Exposures on hazard at time u

$X(t)$ = exposure intensity (**dose**) at time t ($t \leq u$)

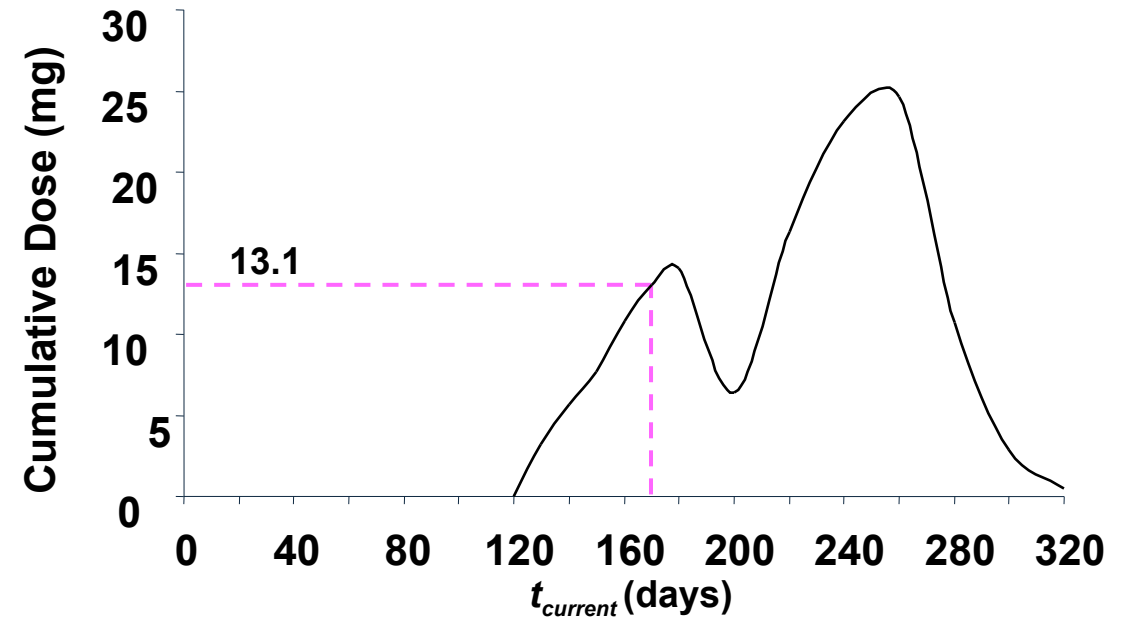
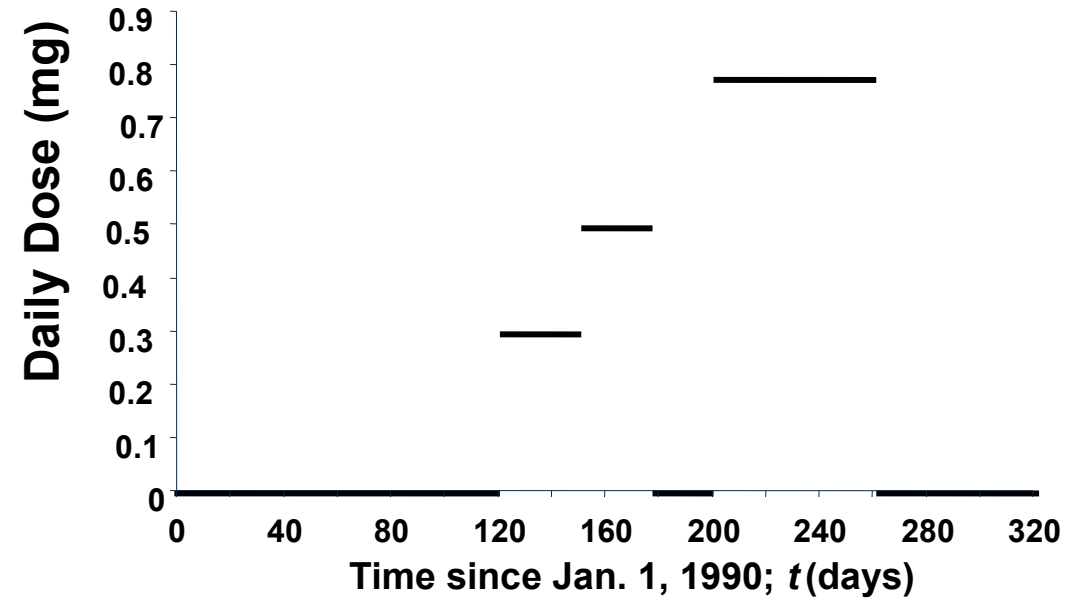
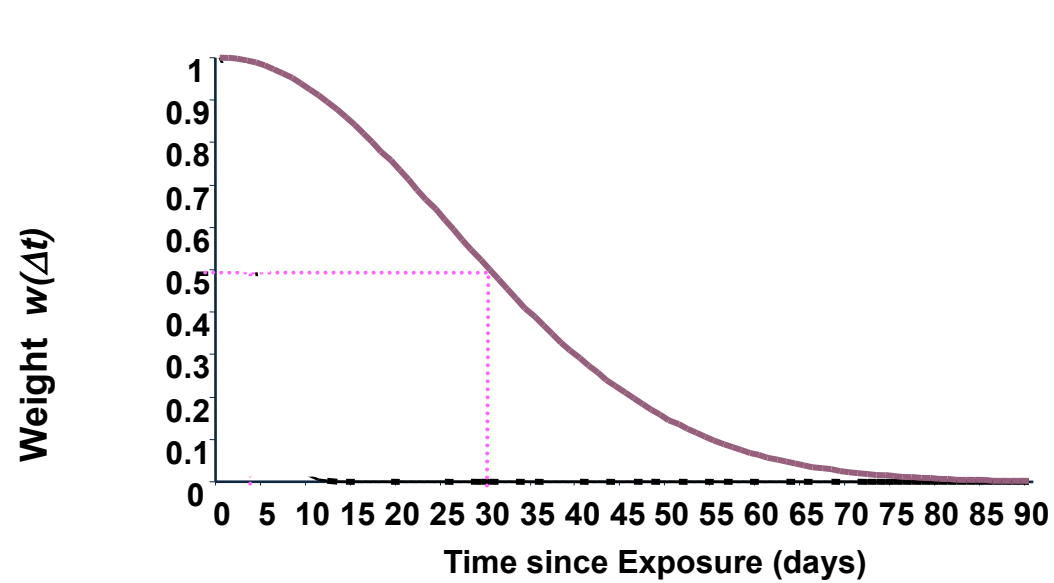
$u-t$ = **time elapsed since exposure** $X(t)$

$w(u-t)$ = **Weight function**, estimates Relative Importance assigned to exposure $X(t)$ as a function of Time-since-Exposure ($u-t$)

**$WCE(u)$ is then included as a Time-Varying Exposure metric in, e.g.,
Cox model for time-to-event analyses**

[Sylvestre & Abrahamowicz, *Stat Med* 2009]

Weights for Cumulative Exposure



Phase “I D”: Proof of the Concept

- To demonstrate the **practical usefulness of the WCE concept**, in the 1st WCE paper we applied the weight function (shown on slide 8), selected *a priori* based on known pharmacodynamics of the drug, in a **real-world pharmaco-epi database** [Tamblyn et al, *J Am Geriatr Soc* 2005]
- The goals were to (i) re-assess the **association between recent use of Temazepam (a psychotropic drug) and hazard of fall-related injuries** (presumably partly due to drug-induced cognitive impairment), and (ii) compare the fit (AIC) of the WCE model with simpler ‘conventional’ exposure models [Abrahamowicz et al, *J Clin Epidemiol* 2006]

N = 3,798 new Temazepam users,
follow-up: max = 5 yrs, median = 2 yrs, **186 Events** (falls) (4.9%)

WCE model fit much better than any of the 5 conventional models,
with 7.5 points improvement in AIC !

Best-fitting WCE model for (weighted) Cumulative Duration, adjusted for current dose:
AIC = **2262.4 versus 2269.8** for the Best-fitting Conventional model (Un-weighted Cumulative Duration) [Abrahamowicz et al, *J Clin Epidemiol* 2006]

Phase “I E”: Refined **Estimation** (Flexible modeling)

- In most real-life applications, the relative importance of doses taken, e.g., 1 week ago vs. 3 months ago is difficult to specify *a priori* !

[Pazzagli, *Pharmacoepidemiol Drug Saf* 2018; Abrahamowicz & Tamblyn, *Enc Biost* 2005]

- Thus, in our next WCE paper we model a **flexible Weight function $w(u-t)$ using unpenalized Cubic regression Splines** [Sylvestre & Abrahamowicz, *Stat Med* 2009]:

$$w(u-t) = \sum_{j=1}^m \theta_j B_j(u-t) \quad (2)$$

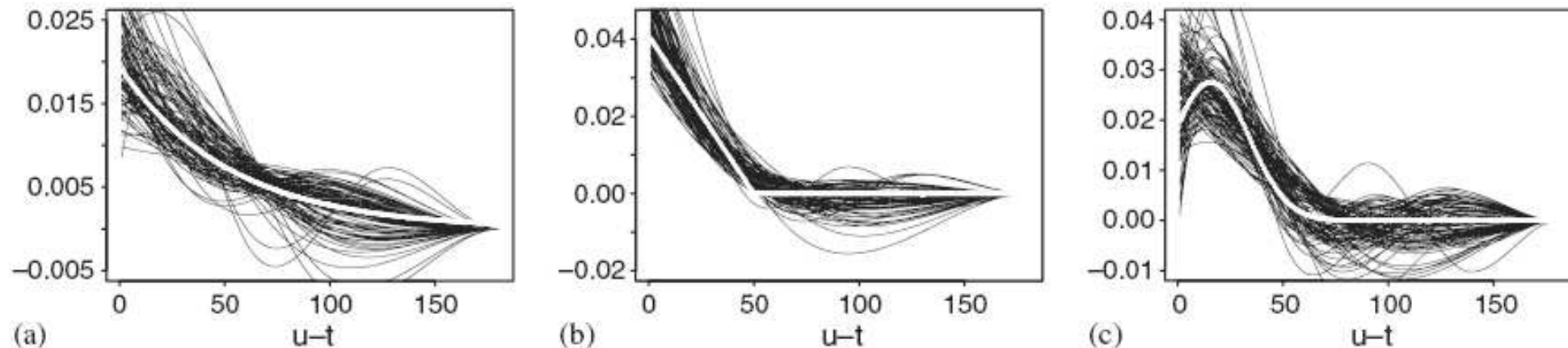
where: $B_j, j=1, \dots, m$, are the m functions in the Cubic Spline basis, and θ_j are the estimable spline coefficients **

** The model can be fit using **standard R functions for time-dependent Cox model** using **Artificial Time-varying Covariates** [Sylvestre & Abrahamowicz, *Stat Med* 2009]

Phase "II A" (Initial Evaluation in Simulations)

Clean Data, 250 events, True model = WCE

[Sylvestre & Abrahamowicz, *Stat Med* 2009]

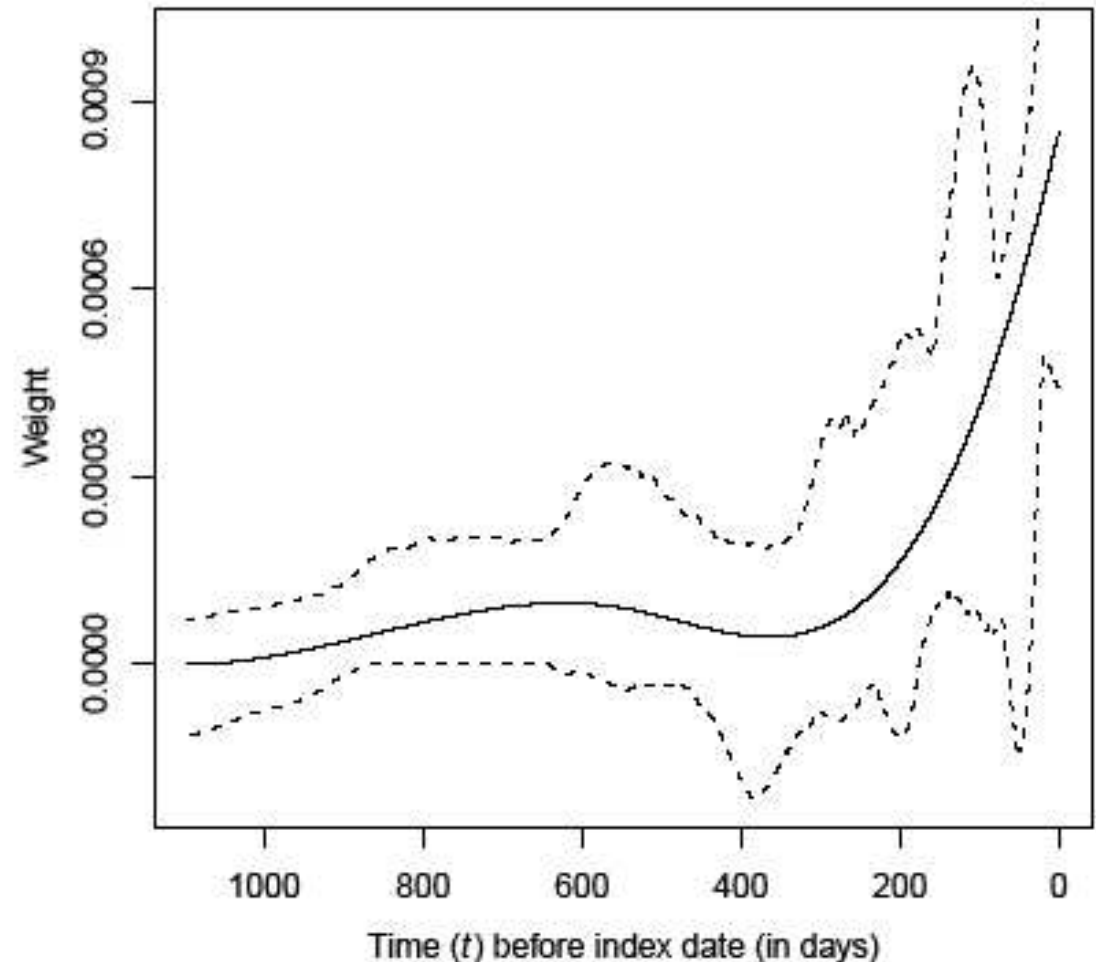


- **Initial Simulation results:**

- **Reasonably Unbiased & Stable Weight function Estimates** (only 250 events)
- Selecting *Time Window that is Too Long* has little impact (Middle & Left panels)
- Constraining $w(u-t)$ to fall to 0 at the right end of time window stabilizes estimates

Phase “II B” (Initial “ideal” Application: Large N (>1800 events), strong association)

- Dixon et al [*Ann Rheum Dis* 2012]
(**395 citations on Google Scholar**)
- Glucocorticoids vs Infection-
caused hospitalizations in
Rheumatoid Arthritis
- **WCE model had AIC 28 points
better than any of 9
conventional models**



Phase "II B": Initial "ideal" Application (cont-d):

New (*plausible/interpretable*) Insights from WCE analyses

- Current Infection risk is affected by Cumulative effects of GC exposures in past 2-3 years** (*contrary to previous beliefs that the latency does not exceed 3-6 months...*)

**** AIC for best fitting 3-yrs WCE model improved by 34.3 or 8.6 relative to WCE models restricted to 3 m. or 1 yr.** [Dixon et al, *Ann Rheum Dis* 2012]

- **Possible Biological Interpretation of the "Bi-modal effect":**
 - GC act on both (i) Innate & (ii) Adaptive Immune Systems
 - (i) Short-Term effect of doses from last 3-6m. on Innate system was known
 - (ii) Long-Term effect on Adaptive system may be Indirect, involving T-lymphocyte apoptosis & failure to generate pathogen-specific adaptive immune responses [McMaster & Ray, *Nat Clin Pract Endocrinol Metab* 2008], or prolonged adrenal suppression

Phase "II/III" (*Re-purposing*): WCE model's Extensions:

1) Non-Drug Exposures:

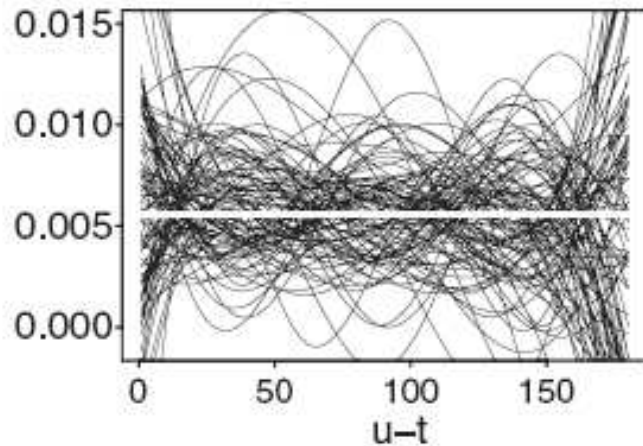
- 1a): $X(t)$ = Radiation doses [Danieli et al, *Am J Epi* 2019];
- 1b): $X(t)$ = Air pollution [Biel et al, *Scientific Reports* 2020]
- 1c): $X(t)$ = Physical activity's intensity [Wang et al, *Paediatr Perinat Epidemiol* 2025]

2) More Complex Models (beyond single-endpoint survival):

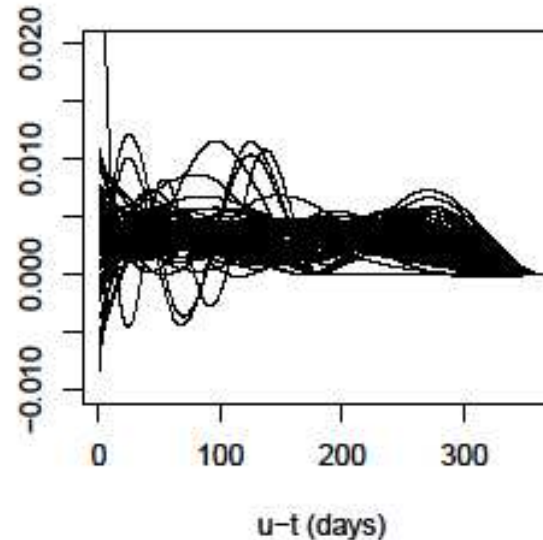
MODEL	Statistical Reference	Methods	Real-world APPLICATION	Applied REFERENCE
Competing Risks	Danieli et al, <i>SMMR</i> 2019	Data Augmentation (Separate Weights for Com events)	Radiation vs Cancer OR Other-Causes Mortality	Danieli et al, <i>Am J Epi (AJE)</i> 2019
Marginal Structural Cox Models (MSM)	Xiao et al, <i>JASA</i> 2014	IPT weights for time-varying confounders	Didanosine vs CVD risks in HIV	Young et al, <i>J AIDS</i> 2015
Mixed Effects Linear models	Danieli et al, <i>SMMR</i> 2020	Changes in Longitudinal	Opioids vs Changes in the	Bhondoeckhan et al

Phase “III”: Further Simulations: Comparing with Simpler Models + regression Diagnostics

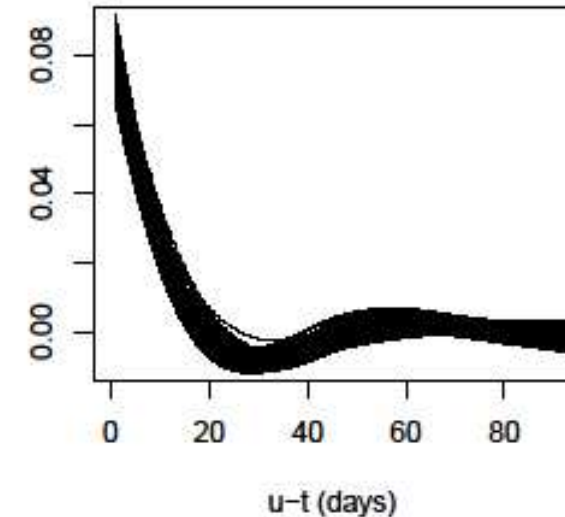
250 events



500 events



500 events



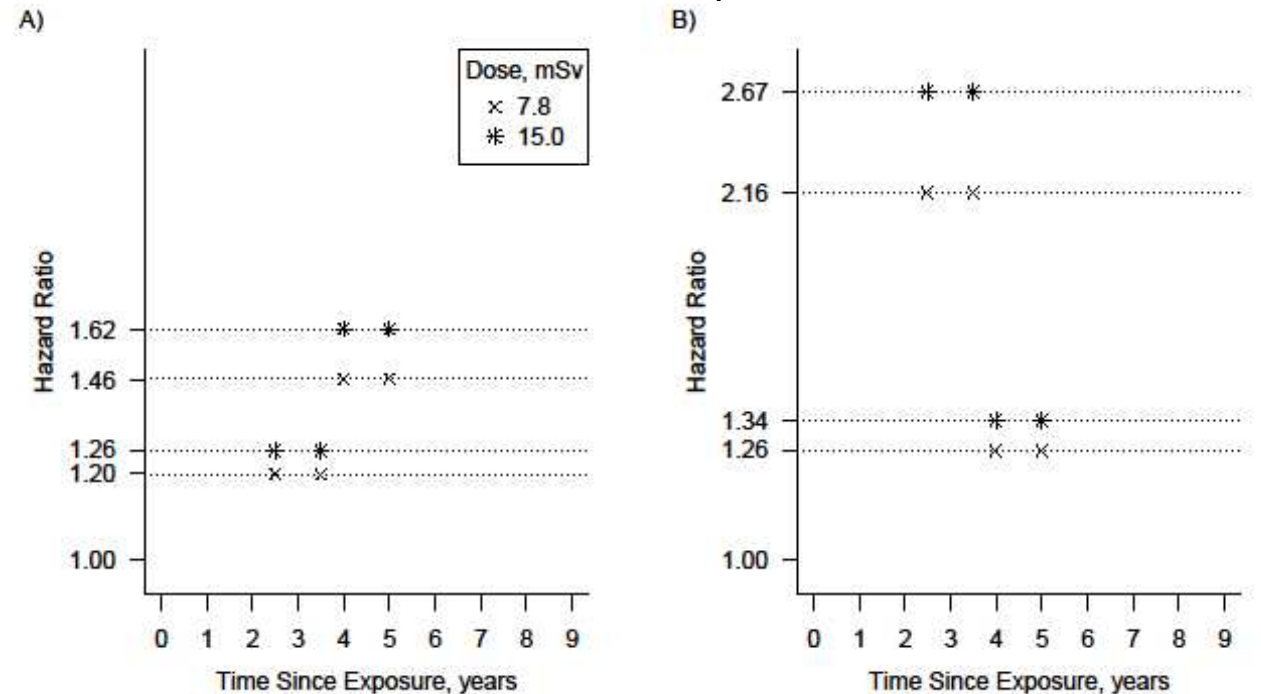
- Left & Middle graphs: **Over-fit Bias**** when True model is Simple (UN-weighted Cumulative Dose)
 - **** improving if more events** [Sylvestre & Abrahamowicz, *Stat Med* 2009]
 - Diagnostics: 3-df LRT's $p > 0.05$ for 94% of samples: **WCE does NOT improve model's fit to data vs. Un-weighted Sum of Past Doses**
- Right graph: True model = Current Dose (WCE estimate suggests Very Short-term impact);
 - Diagnostics: **AIC favors Current Dose over WCE in All samples** [Abrahamowicz et al, *Stat Med* 2012]

Phase "III" (Advanced Applications): User-friendly presentation of complex WCE results

WCE-based
Infection ORs for various clinically
relevant patterns of GC use
[Dixon et al, *Ann Rheum Dis* 2012]

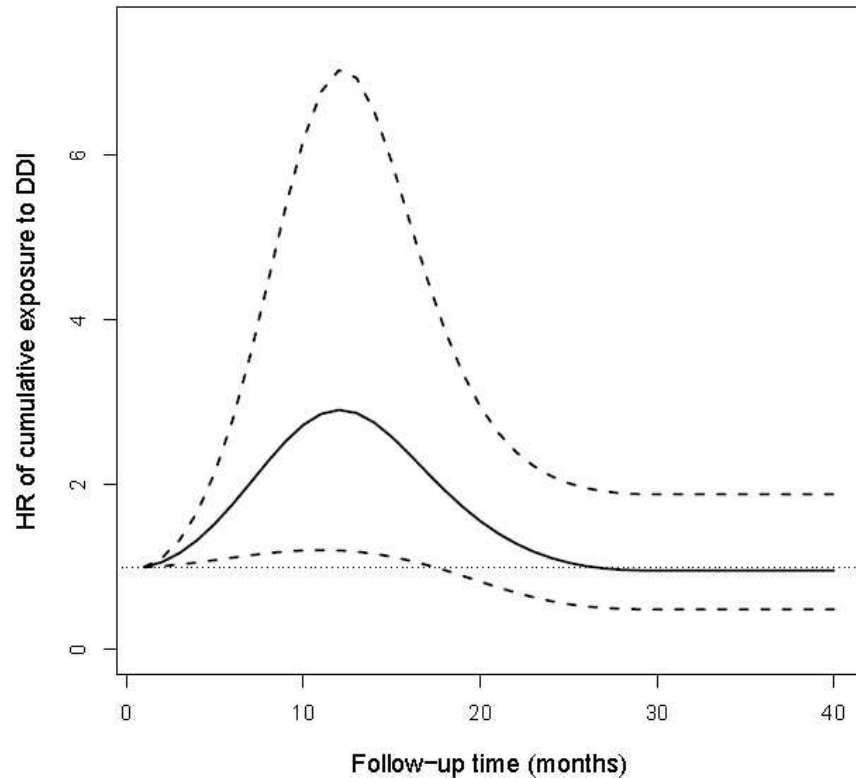
Pattern of use	Reference	OR
Current user, 5mg, for last 28 days	Non-user	1.11 (1.08, 1.26)
Current user, 5mg, for last 3 months	Non-user	1.30 (1.21, 1.45)
Current user, 30mg, for last 28 days	Non-user	1.84 (1.58, 4.00)
Current user, 30mg, for last 3 months	Non-user	4.82 (3.12, 9.29)

Competing-risks WCE HRs for Cancer
incidence
for selected real-world patterns of past doses
of low-dose ionizing radiation (LDIR)
for Men (Left) vs. Women (Right)
[Danieli et al, *Am J Epidemiol* 2019]



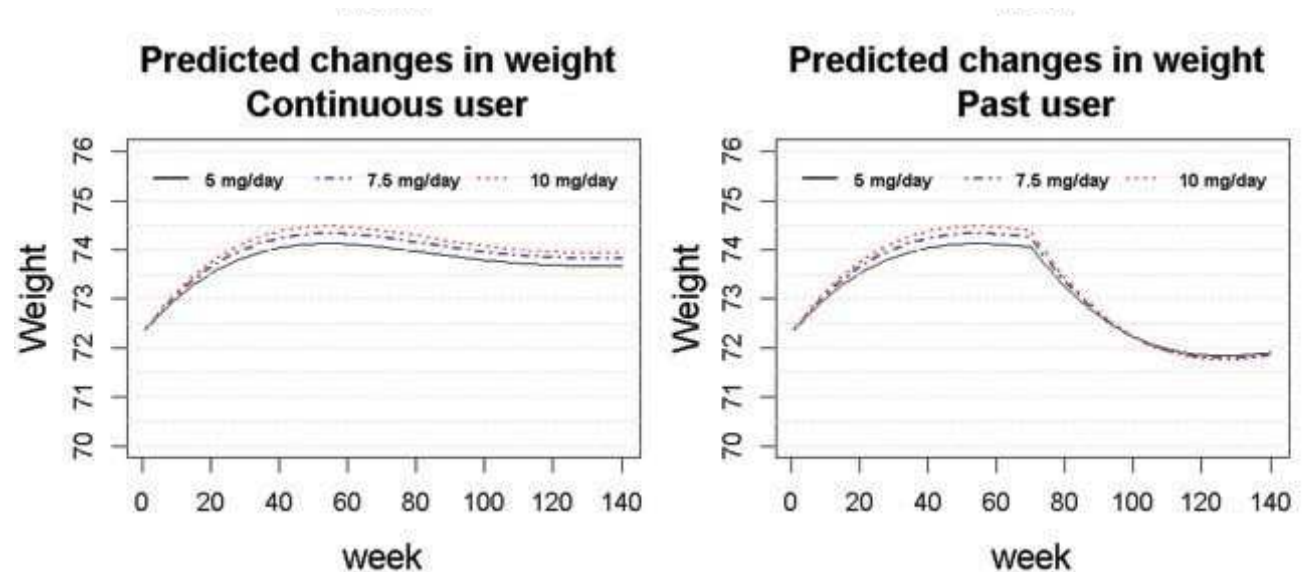
Phase “II and/or III” (Advanced Applications): User-friendly presentation of complex WCE results

Changes in CVD hazard with Increasing Duration of continuous Didanosine Tx in HIV [Xiao et al, *JASA* 2014]

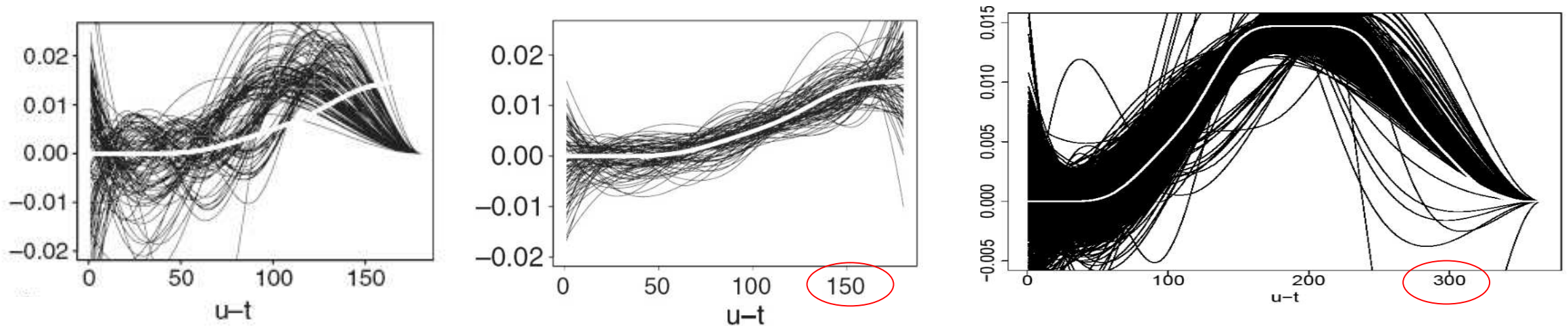


Changes in the Body Weight for a woman with baseline weight of 72.4 kg, according to different patterns of Glucocorticosteroids use

[Danieli et al, *SMMR* 2019]



Phase "III/IV": Advanced Simulations: Diagnostics to Identify **Potential Problems in Applications** [Sylvestre & Abrahamowicz, *Stat Med* 2009]



3 $w(u-t)$ Estimates for True = "Hat" function over 360 days (White curve in Right panel):

1. **Left: Wrong support window (Only 180 days) with $w(u-t)$ Constrained to 0 at the end**
2. **Middle: 180 d. window But $w(u-t)$ UN-Constrained (LRT's $p < 0.05$ in 87% samples) ****
3. **Right: Correct window (360 d.) Best Fit (Minimum AIC in All samples) *****

**** 2-df LRT of Constrained vs. Unconstrained model = DIAGNOSTIC test for Time Window**

***** > 4 points AIC Difference validated in simulations as model choice's criterion**

Phase “IV”: Independent Recommendations & Narrative Reviews

- Review of time-varying drug exposures [Pazzagli et al. PDS 2018]:
identifies Cumulative Effects of past Drug Use among 1 the 4 most important challenges, & recommend WCE methodology for such analyses
- A dedicated Narrative Review [Kelly et al. PDS 2024] of WCE modeling in 17 real-world pharmaco-epi studies concludes:
 - *“The WCE method is an important tool for exploring the effect of time-varying exposures on an outcome, including the dose, duration, and timing of past exposures, ... and allows additional insights into their effects.”*
 - *“... WCE is a powerful addition to conventional methods of classifying exposure...”*
 - *9 of the 10 papers that compared alternative exposure models reported Best Fit of WCE*

Missing elements of Phase III: Neutral Comparisons

- Still TO DO:

Neutral Simulation-based Comparisons with
Alternative flexible Models ** proposed to assess
Cumulative Effects of Time-Varying Exposures
(as recommended for Phase III by [Heinze et al. *Biom J* 2024]):

** 1/ **Distributed Nonlinear Lags Models (DNLMS)**

[Gasparini, *Stat Med* 2014; Gasparini et al. *Biometrics* 2017]

2/ **Exposure-time-response models** [Berhane et al. *Stat Med* 2008]

CHALLENGE: both 1/ & 2/ use 3D Tensor Product splines

WCE Software (*only in R*)

- **CRAN Webpage** (Comprehensive R Archive Network)
<https://CRAN.R-project.org/package=WCE>
[Sylvestre, Beauchamp, Kyle, Abrahamowicz, *R package* 2024]
- **Detailed Example illustrating package implementation in real-world analyses, based on a dataset included in the package**
<https://cran.r-project.org/web/packages/WCE/vignettes/WCE.pdf>
(Temporarily unavailable, see here for now
<https://github.com/mebeauchamp/WCE-R-package>)
- **DOWNLOADS since 2015: >32,000**
- Code for extensions available upon request

Summary & Conclusions

(Still on-going...) the process of Development, Evaluation, Applications & Extensions of the WCE methodology has followed Most of the Phases identified by Heinze et al [Biom J 2024]

- Yet, Phase I included several distinct sub-phases and some of the elements of Phases II-IV were done in a different order
- In our experience, **Essential were the Inter-connections of Methodological Developments with Real-world Applications:**
 - i. **Real-world analyses stimulated new methodological developments**, necessary to **address new analytical challenges** (and identified some limitations of the method **)
 - ii. **WCE estimates provided new insights into, and generated new hypotheses about, the underlying biological processes** linking time-varying exposures with the outcomes

** E.g. in 1 drug application WCE has Not improved fit over conventional current use model: the reason was that Drug Doses remained very stable over time for most subjects

[Bally et al. *Pharmacoepidemiol Drug Saf* 2018] > **Phase III Conclusion: WCE should be applied ONLY IF Individual Exposures show substantial Variation over Time!**

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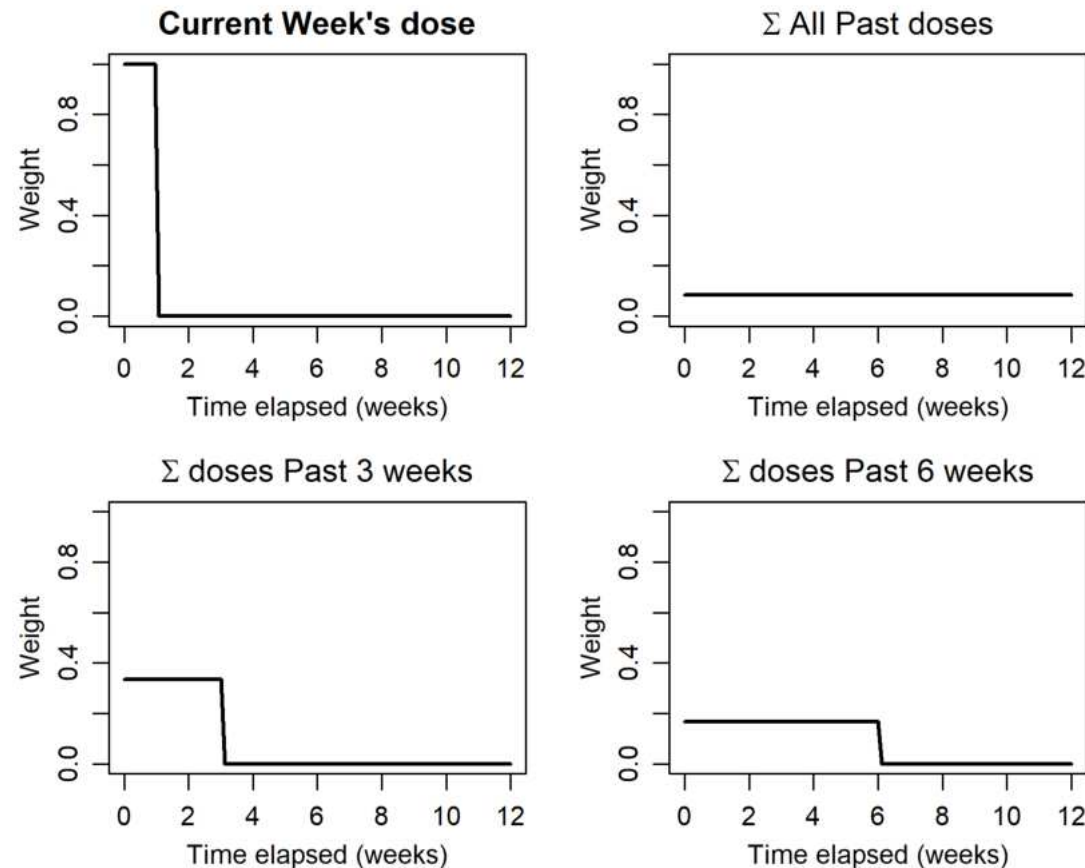
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Phase "I C": (*cont-d*)

New Model: Concept & Formulation

- The proposed **WCE model** includes **several conventional exposure metrics as its special cases**, each with a different Weight function





Flexible WCE Model

[Sylvestre & Abrahamowicz (2009)]

- **WCE in (2) is then modeled as a Time-Varying Covariate in Cox's model:**

(3)

where:

$h_0(u)$ is the baseline hazard

$\mathbf{X}(u) = \{X(t), 0 \leq t \leq u\}$ represents the time-vector of the past exposures

$Z_s(u), s=1, \dots, q$, are the values of the fixed-in-time or time-dependent covariates relevant at time u

ESTIMATION of the Flexible WCE Model through Artificial Time-varying Covariates

From equations (1), (2) & (3), the effect of WCE is modeled as:

$$WCE(u) = \sum_{t \leq u} w(u-t) * X(t) = \beta \sum_t \sum_j \theta_j B_j(u-t) * X(t)$$

where BOTH β & θ_j need to be estimated.

To Avoid Identifiability Problems, we define:

$$\gamma_j = \beta \theta_j \quad (4)$$

& construct Artificial Time-varying Covariates:

$$D_j(u) = \sum_t^u B_j(u-t) X(t) \quad (5)$$

for $j=1, \dots, m$

ESTIMATION of the Flexible WCE Model through Artificial Time-varying Covariates

Given (4) & (5), the Cox's model in (3) becomes:

(6)

Once $D_j(u)$, $j=1,...m$, are calculated for each u = uncensored event time, the model in (6) can be implemented using standard software for Cox's model with time-dependent covariates