

# The Breslow Estimator, the Breslow-Crowley Theory, and Their Modern Impacts

## A Personal History

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University of Washington

August 4 / In honor of Norman E. Breslow

# Outline

- 1 Introduction: the facts
  - The Breslow estimator
  - The Breslow - Crowley theory
- 2 The Impacts, Part I
  - Citations
  - Developments
- 3 The Impacts, Part II

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# The Breslow estimator.

of the baseline hazard, Cox model

- $X = (\Delta, T, Z)$ ,  
 $T = \min\{\tilde{T}, C\}$ ,  $\Delta = 1\{\tilde{T} \leq C\}$ ,  $Z =$  vector of covariates.
- $\theta =$  finite-dimensional parameter;
- three infinite - dimensional parameters;  
 $\Lambda = \int_0^{\cdot} \lambda(s) ds =$  baseline hazard function,  
 $G(t|Z) = P(C \leq t|Z) =$  conditional distribution of  
censoring time given covariates  $Z$   
 $G_Z$ , the marginal distribution of the covariate vector  $Z$ .
- $\Lambda(t|Z) = e^{\theta'Z} \Lambda(t) \equiv$  “the Cox model”

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David R. Cox (1975)



Sir David R. Cox (March, 2006)

- $N(t) = \Delta 1_{[T \leq t]}$  = event counting process;  
 $Y(t) = 1_{[T \geq t]}$  = at risk process.
- Density of  $X = (\Delta, T, Z)$  at  $x = (\delta, t, z)$  is given by

$$\exp(-e^{\theta'z} \Lambda(t)) [e^{\theta'z} \lambda(t) (1 - G(t|z))]^\delta [g(t|z)]^{1-\delta} g_Z(z).$$

- Suppose  $X_1, \dots, X_n$  i.i.d. as  $X = (\Delta, T, Z)$ .
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$$\bar{N}_n(t) = \mathbb{P}_n \Delta 1\{T \leq t\} = n^{-1} \sum_{i=1}^n N_i(t),$$

$$\hat{S}_n(t; \theta) = \mathbb{P}_n e^{\theta'Z} Y(t) = n^{-1} \sum_{i=1}^n e^{\theta'Z_i} Y_i(t).$$

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# The Breslow estimator

- The **Breslow estimator** of  $\Lambda =$  baseline hazard function is:

$$\hat{\Lambda}_n(t) = \int_0^t \frac{1}{S_n^0(s; \hat{\theta})} d\bar{N}_n(s).$$

- where  $\hat{\theta} =$  Cox partial likelihood estimator of  $\theta$ .
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# Breslow - Crowley theory

limit theory for the Kaplan - Meier estimator

## Kaplan-Meier estimator

- With no covariates, let

$$\begin{aligned}\bar{Y}_n(t) &= \mathbb{P}_n Y(t) = \mathbb{P}_n \mathbf{1}\{T \geq t\} \rightarrow_{a.s.} P \mathbf{1}\{T \geq t\} \\ &= (1 - F(t-))(1 - G(t-)) \equiv 1 - H(t-).\end{aligned}$$

- $\hat{\Lambda}_n(t) = \int_0^t \frac{1}{\bar{Y}_n(s)} d\bar{N}_n(s)$  estimates  $\Lambda(t) = \int_0^t \frac{1}{1-F(s-)} dF(s)$ .
- The survival function  $1 - F$  is given in terms of  $\Lambda$  by

$$\begin{aligned}1 - F(t) &= \exp(-\Lambda_c(t)) \prod_{s \leq t} (1 - \Delta\Lambda(s)) \\ &= \begin{cases} \exp(-\Lambda(t)), & \Lambda \text{ continuous} \\ \prod_{s \leq t} (1 - \Delta\Lambda(s)), & \Lambda \text{ discrete.} \end{cases}\end{aligned}$$

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# Paul Meier photo



Paul Meier

- Kaplan - Meier (1958) estimator  $\hat{F}_n$  of  $F$  is given by

$$1 - \hat{F}_n(t) = \prod_{s \leq t} (1 - \Delta \hat{\Lambda}_n(s)) = \prod_{s \leq t} \left( 1 - \frac{\Delta \bar{N}_n(s)}{\bar{Y}_n(s)} \right).$$

- Breslow & Crowley (1974) proved:

### Theorem

For each  $\tau < \inf\{t : H(t) = 1\}$

$$\sqrt{n}(\hat{F}_n - F) \Rightarrow (1 - F)\mathbb{B}(C) \quad \text{in } D[0, \tau]$$

where  $\mathbb{B}$  is standard Brownian motion on  $[0, \infty)$  and

$$C(t) = \int_0^t \frac{1}{(1 - H)} d\Lambda.$$

- Kaplan - Meier (1958) estimator  $\widehat{F}_n$  of  $F$  is given by

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- Cites of 1974 Breslow - Crowley: 354
- Context:
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# Impacts & consequences

- **Martingale theory:**  
Le Cam  $\rightarrow$  Aalen  $\rightarrow$  Gill  $\rightarrow$  Anderson & Gill ...
- **Martingale approach to survival analysis:** two books, Andersen, Borgan, Gill, and Keiding; Fleming and Harrington
- **Dutch school of semiparametrics**  
Groeneboom  $\rightarrow$  Gill  $\rightarrow$  van Zwet, Klaassen, van der Vaart
- **Compact differentiability:**  
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- **Empirical process approaches to Breslow-Crowley:**  
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Odd Aalen



Richard Gill



Per Kragh Andersen

- **Information bounds:**

Efron (1977) → Begun, Hall, Huang, Wellner (1983) →  
BKRW (1993)

- **Confidence bands:**

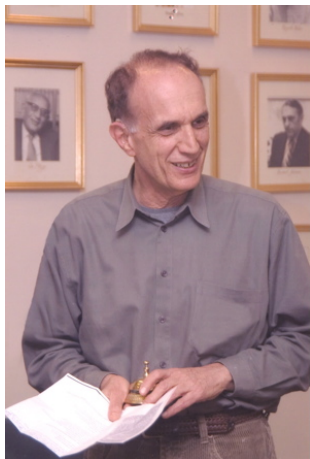
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Brad Efron

# Impacts - personal I

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*Dudley, R. M. (1966). Weak convergence of probabilities on non-separable metric spaces and empirical measures on Euclidean spaces. Illinois Journal of Math. 10, 109-126.*

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Dick Dudley (1971)



# Impacts - personal II

- Discussions with Crowley about work with Norm
- Rochester years: 1975 - 1983.
- 1979 - 1983: **efficiency theory for semiparametric models**
- one point of origin: trying to understand Efron, B. (1977). The efficiency of Cox's likelihood function for censored data. *J. Amer. Statist. Assoc.* **72**, 557 - 565.
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- **Need and motivation** for the more general empirical process theory being developed by Dudley, Giné, Zinn, Pollard, and others.

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  - getting: out of the office and into the mountains
  - three treks in Nepal (1989), (1996), (1999)
  - skiing the the Coast Range and Cascades

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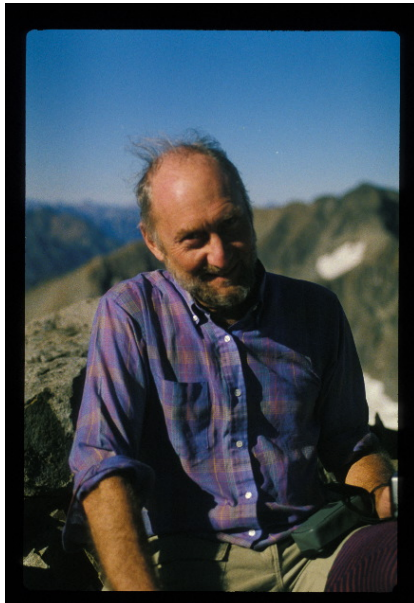
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Jon A. Wellner











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Breslow & Breslow-Crowley Impacts



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## Impacts - personal IV

*In his (1972) discussion of D. R. Cox's famous paper on the Cox proportional hazards model, **Norman Breslow** showed how the baseline cumulative hazard function  $\Lambda$  could be estimated in a natural way in parallel with the estimators of the regression parameters proposed by Cox. The resulting "Breslow estimator", together with Breslow's (1974) joint paper with John Crowley on the asymptotic behavior of the Kaplan-Meier estimator, set the stage for an extraordinary collection of developments in survival analysis, semiparametric models, and counting processes. These developments continue to have significant and profound consequences for present day statistical research and practice.*

## Impacts - personal IV

Norm has also had a profound effect on my own research directly (and indirectly) over a period of more than 30 years.

THANK YOU NORM