An Introduction to Longitudinal Biomarker Evaluations

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What is longitudinal biomarker data?

- Biomarker measured in serial samples from each individual.
- CA-125 measured annually for ovarian cancer screening.
- CEA measured monthly for colon cancer recurrence after initial treatment.

Key Idea: Change in biomarker over time may inform about disease onset.

• **Example**: a large change from relatively stable levels.



Key Idea: Change in biomarker over time may inform about disease onset.

• **Example**: consistent trend away from baseline.



Not talking about markers

- with undetectable levels in controls
- germline gene variants or other lifelong stable measures of risk

Talking about markers

- present in people without cancer (controls)
- controls tend to have their own individual set-point

Talking about

- CA-125 for ovarian cancer screening (Skates).
- PSA for prostate cancer screening (Zheng).
- AFP, DCP for liver cancer screening (Tayob).

Intra-class Correlation Coefficient (ICC)

$$ICC = \frac{variation between subjects}{var(between) + var(within)}$$



Illustration of potential added value from longitudinal biomarker

- Stored annual samples
- Could we have detected cases that occurred after start of year 3 using markers measured up to year 3?
- Single time point marker: M₃
- Longitudinal marker: $\Delta_3 = M_3 ave(M_1, M_2)$



Low within-subject variation in controls (ICC=0.90)



Moderate within-subject variation (ICC=0.70)



High within-subject variation (ICC=0.30)



Recommendations

- Markers that show mild/moderate performance at a single time point
- Consider if there is potential for better performance with longitudinal ascertainment.
- Statistical considerations.
 - Do a preliminary study using serial samples *from controls* to assess ICC
 - This is cheap compared with larger study to assess biomarker performance for detecting disease in cases.
 - Assay variability.
- Feasibility considerations and complications
 - Will people return to provide serial samples?
 - Algorithms to accommodate missing time points.

Algorithms for using longitudinal biomarker data with real data in real applications

- Ovarian cancer Steven Skates
- Prostate cancer
- Liver cancer

Yingye Zheng

Mabihah Tayob

Appendix

$$AUC(M_3) = 0.60 = \Phi(a/\sqrt{2})$$
$$\Rightarrow \quad ROC(t) = \Phi(a + b\Phi^{-1}(t))$$

To generate data

 $M_{ij} = I[i = 1]D_j + \mu_j + \varepsilon_{ij}$

$$i = \text{time} \quad j = \text{subject}$$

$$M_{ij} = I[i = 3]D_j + \mu_j + \varepsilon_{ij}$$

$$\varepsilon_{ij} = N(0, \sigma_{\varepsilon}^2) \quad \sigma_{\varepsilon}^2 = 1$$

$$\cdot \mu_j = N(0, \sigma_{\mu}^2) \quad \sigma_{\mu}^2 = \frac{\text{ICC}}{1 - \text{Icc}}$$

$$D_j = I_M \times \mu_D$$

$$I_M = \text{binary} \begin{cases} \text{prob} = 100\% & \text{marker in } 100\% \text{ cases} \\ = 50\% & \text{marker in } 50\% \text{ cases} \end{cases}$$

$$\mu_D = a \times \sqrt{(\sigma_{\mu}^2 + \sigma_{\varepsilon}^2)}$$