

# **Risk factors for MCI & Dementia: new statistical tools**

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# Outline

- Contrasting two cohorts: BRAINS and Nuns
- Analysis of retrospective data: the Nun Study

Markov chain defines transitions between assessments  
One step transition probabilities  
Adjusting for baseline

- Analysis of prospective data:

Censored versus competing events  
Kaplan-Meier curves and Cox models  
Cumulative incidence curves

# Descriptive statistics

Factor	BRAiNS	Nuns
n	553	424
Age at entry	73.5 ± 7.6	83.1 ± 5.1
No. visits	8.0 ± 4.0	7.2 ± 2.8
Female	64.2 %	100 %
Positive family hx	37.5 %	-
APOE 4 allele	30.2 %	19.3 %
Years of Education:		
≤12	11.9 %	9.7 %
13-15	21.0 %	0.0 %
16	31.6 %	41.5 %
> 16	35.4 %	48.6 %

# Recognizing impairments retrospectively

At each annual assessment each participant is categorized into one of five states:

1. Cognitively intact
2. “Mild Cognitive Impairment”
3. Global Impairment
4. Demented
5. Dead

## Retrospective review produces a Longitudinal Record: Categorical responses

Each participant generates a “vector” of responses

$$(y_1, y_2, \dots, y_n)$$

Here  $y_j$  is the state at visit  $j$

Examples:

Subject 1 record (1, 1, 2, 3, 1, 4)

Subject 2 record (2, 1, 2, 5)

# Challenge: analyze categorical vectors of varying lengths

**Solution:** (proposed by our Core)

use a Markov chain with a shared random effect

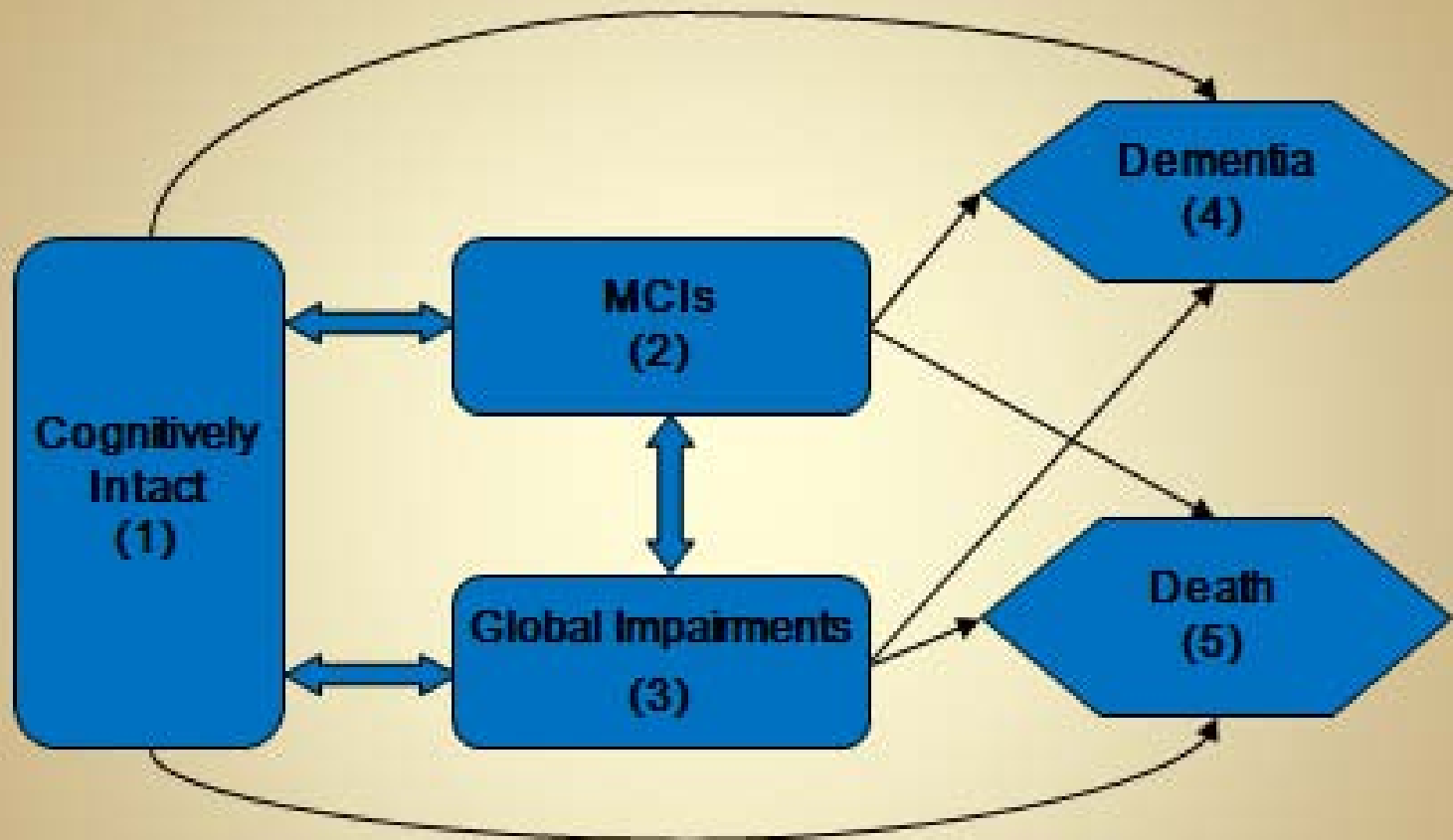
**Markov:** next cognitive assessment depends on current assessment and is independent of prior assessments

Subj. 1: (1, 1, 2, 3, 1, 4) provides 5 data points

Subj. 2: (2, 1, 2, 5) provides 3 data points

one step transitions  $2 \rightarrow 1$ ,  $1 \rightarrow 2$ , and  $2 \rightarrow 5$

**Shared random effect:** a latent (unobservable variable) used to correlate the transitions for a given subject



# Markov chain (ignores risk factors)

## Nun Study (n = 424)

### Current Assessment

Prior Assessment	Cognitively Intact	MCI	Global Impairment	Dementia	Death
Intact	537 (65.8)	183 (22.4)	53 (6.5)	5 (0.6)	38 (4.7)
MCI	163 (15.0)	644 (59.2)	123 (11.3)	81 (7.4)	77 (7.1)
Glob. Imp.	15 (4.1)	36 (9.9)	163 (44.9)	68 (18.7)	81(22.3)



# Regression analysis

Consider a one step transition and let

$$P_{sk}(\theta | \mathbf{z}_i) = P [ i^{\text{th}} \text{ subject next visits state } k \text{ given state } s ]$$

Assume: **Polytomous logistic regression model**

$$\text{Log} [ P_{sk}(\theta | \mathbf{z}_i) / P_{ak}(\theta | \mathbf{z}_i) ] = \alpha_{sk} + \beta_{sk} \mathbf{z}_i + \gamma$$

Here

- a = baseline state,
- $\mathbf{z}_i$  = vector of fixed effects (risk factors) for  $i^{\text{th}}$  person
- $\theta$  = vector of unknown parameters ( $\alpha, \beta$ )
- $\gamma$  = unobservable shared random effect

**Definition:** Likelihood function for the unknown vector  $\theta$  is the **product of  $P_{sk}(\theta | \mathbf{z}_i)$  over all transitions and subjects** with the shared random effect integrated out

To evaluate the likelihood must solve a numerical integral

## Results of the regression analysis

### Significant risk factors for a single transition

Factor	BRAiNS	Nun Study
Age	All four states	All four states
APOE 4	Amnestic MCI Dementia	All four states
Education: ≤ 12 yrs.	Amnestic MCI non Amnestic MCI	MCI Global Impairment Dementia
	Salazar et al Stat. Med, 2007	Tyas et al Am J Epi, 2007

## Event rates in BRAiNS versus Nuns

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Cohort	n	Baseline Dementias	Follow-up Dementias	Deaths*	Percent Events
BRAiNS	553	0	55	144	36.0%
Nuns	501	77	153	184	82.6%

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\*died before dementia

Conclude: do ADCs introduce a selection bias in their recruiting protocols ?

## Adjusting for baseline

In a recent simulation study we showed  
(Yu et al, Comp Stat Data Analysis, 2009)

If some subjects are demented at baseline ( so called left truncated events),  
then ignoring baseline attentuates the effects of the risk factors

## Dependent variable = time to MCI

Problems of interest:

1. Identify an appropriate statistical method for determining the probability of conversion to MCI after  $t$  years of follow-up
2. Adjust 1 for risk factors: education, family history of dementia, APOE 4 status, gender

# Events and non events

805  
BRAiNS subjects

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graph TD; A[805 BRAiNS subjects] --> B[298 still at risk for MCI (37.8%)]; A --> C[149 MCI conversions (true events) (18.5%)]; A --> D[269 died w/o conversion to MCI (33.4%)]; A --> E[89 withdrew w/o conversion (11.0%)];
```

298 still at risk  
for MCI  
(37.8%)

149 MCI conversions  
(**true events**)  
(18.5%)

269 died w/o  
conversion to  
MCI  
(33.4%)

89 withdrew w/o  
conversion  
(11.0%)

# Competing Risks

Gooley, Leisenring, Crowley, & Storer (2002)

Definition: a competing risk is an event whose occurrence either

**precludes** the occurrence of another event under examination or

**fundamentally alters the probability** of occurrence of this other event

Clearly: death before conversion to MCI is a competing risk and not a right censored event

withdrawals could be competing events if they are informative

# Adjustment for competing risks

Simplest cure: use **Incidence curves** instead

Two facts:

1. If there are no competing events:

$$\text{Incidence} = 1 - \text{Kaplan Meier}$$

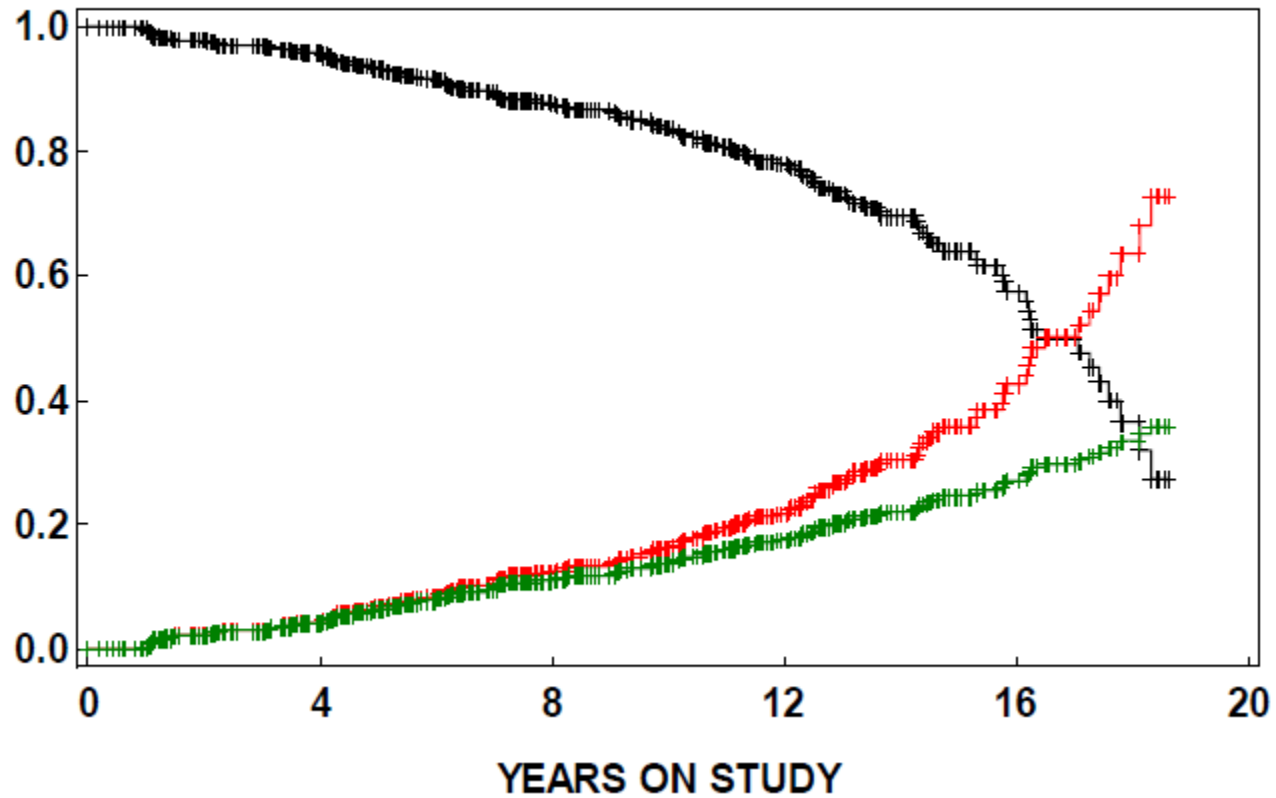
2. If there are competing events

a. calculate Kaplan-Meier for combined risks

b. calculate Incidence for a specific cause by adjusting the hazard of an event age  $a$  for the risk of that event occurring at that age



## Kaplan-Meier Curve and Incidence Curves



## Cox model, marginalized Cox model, Gompertz models event: conversion to MCI; competing risk = death

Factor	Cox model HR (P value)	marginal Cox HR (P value)	Gompertz * HR (P value)
Age entry	1.045 (0.0009)	1.089 (0.0001)	1.087 (0.005)
Apoe 4	1.44 (0.045)	1.20 (0.09)	1.20 (0.21)
Family Hx	1.09 (0.66)	0.91 (0.39)	1.92 (0.015)
Female	0.83 (0.31)	0.70 (0.0003)	0.49 (0.025)

# Conclusions:

1. Analyze cohort data retrospectively means examining transitions into and out of impaired states before absorption into dementia and/or death

Need new statistical tools for analyzing longitudinal data with categorical responses: Markov model with shared random effect

Extend this to delineate risk factors for different forms of dementia

2. Analyze prospective data: examine age at which clinical MCI first occurs

Standard tools: Kaplan-Meier curves and Cox models may not be applicable in the presence of a competing event such as death.