



An Introduction to Markov Chain Models: Transitions between Robustness, Frailty, and ADL-Dependence in Late Life

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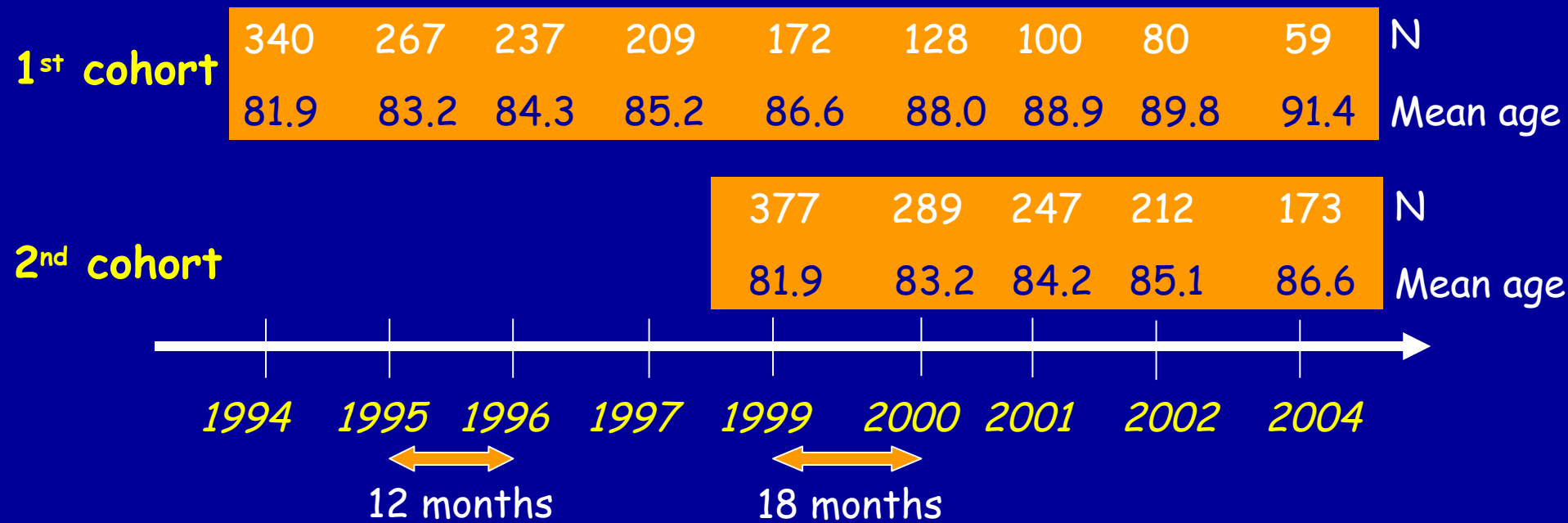
Objectives

1. To document health transitions or health trajectories with Markov Chain (MC) and Double Chain Markov Model (DCMM)
2. To compare MC models with regressions

Participants to Swilsoo

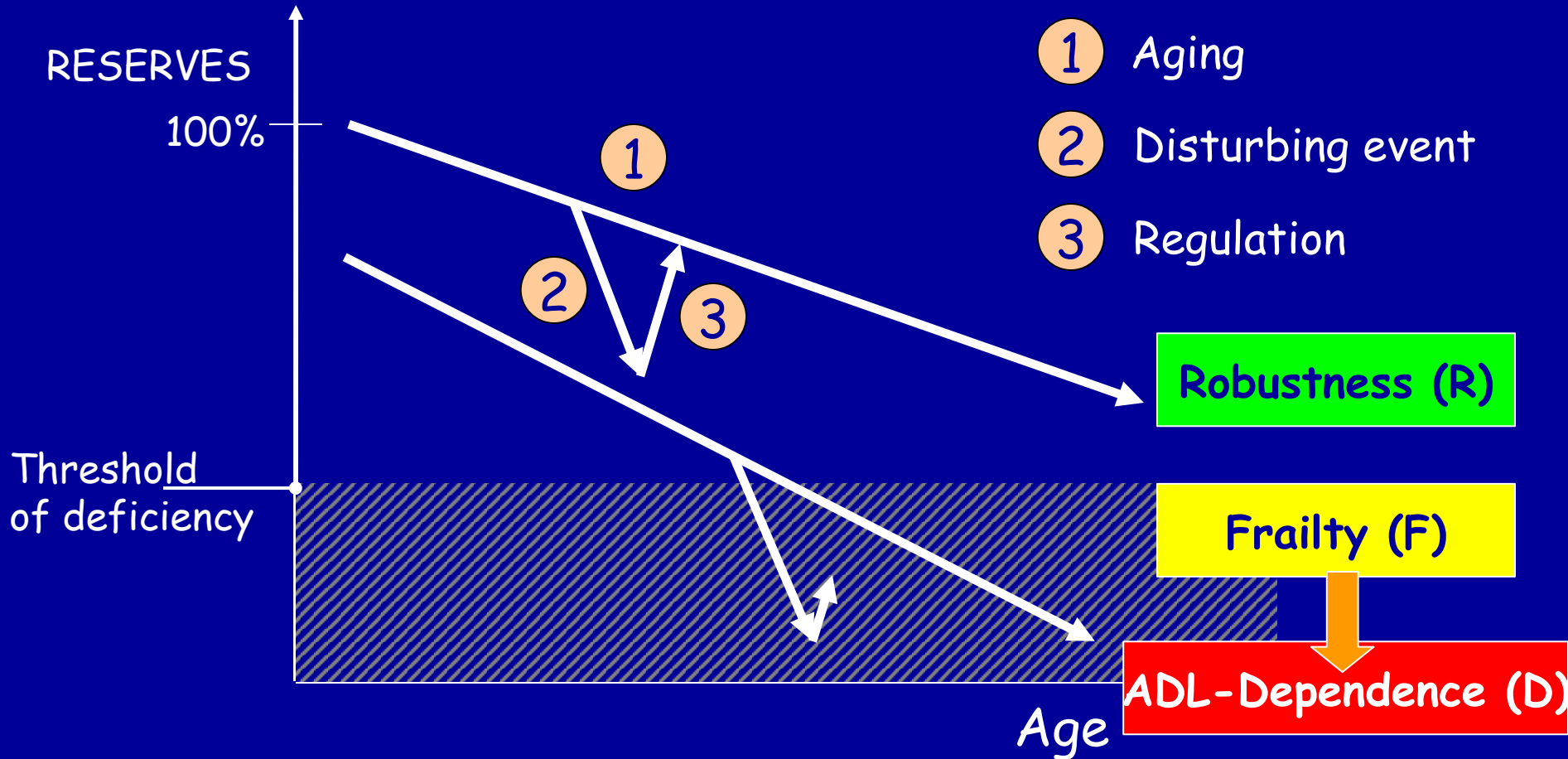
(Swiss Interdisciplinary Longitudinal Study on the Oldest-Old)

- Primary Investigator: Prof. Christian Lalive d'Epina
- Supported by the Swiss National Science Foundation

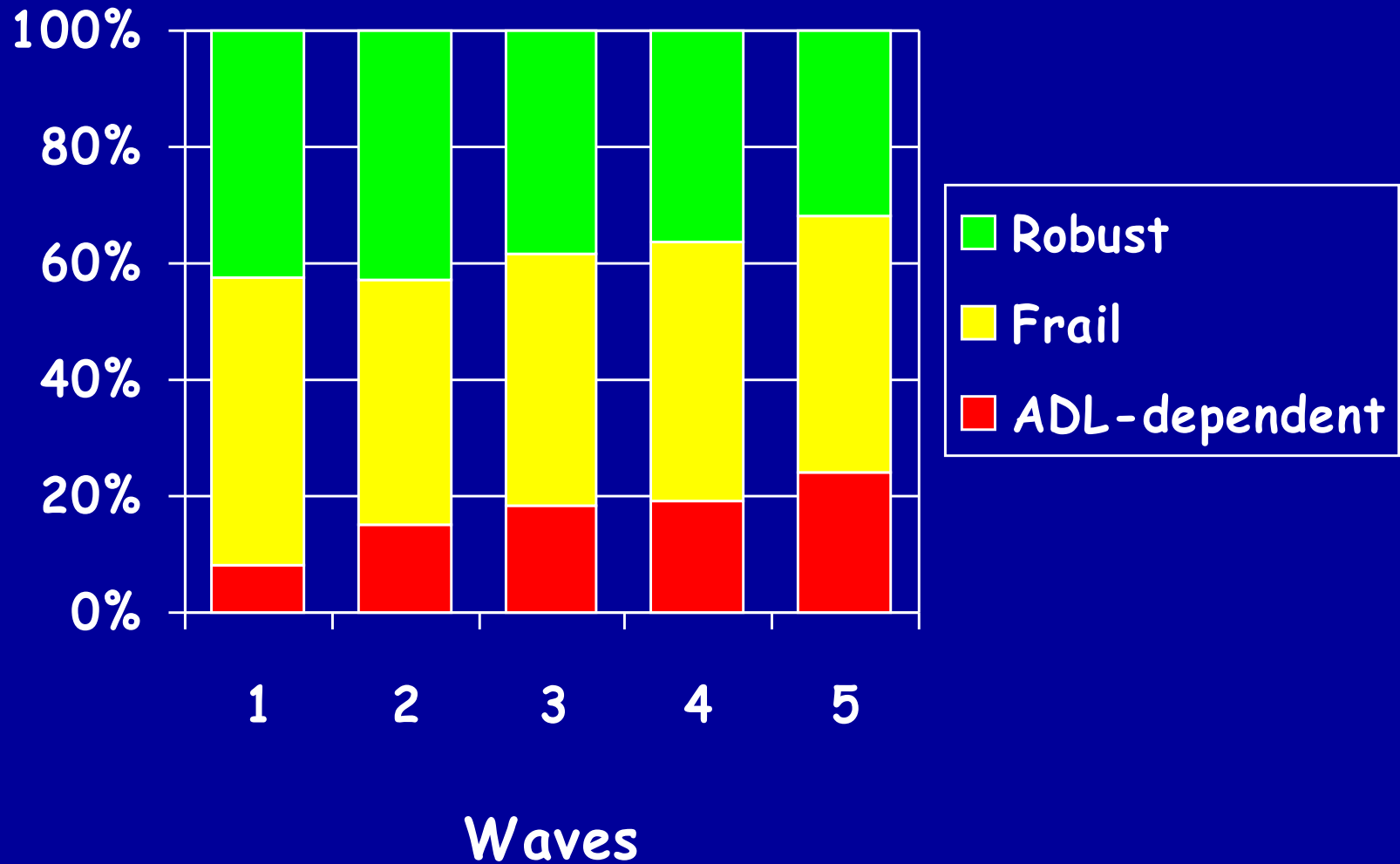


- Residing at home at study inception
- Sample stratified by gender (~50% F) and geographical area (~50% urban and ~50% semi-rural)

Health states



5-wave evolution of health status for the two cohorts



Contingency-table

Homogeneous 1st order MC

... F - F - **F - D** ...
 ... R - R - R - F - F - F ...
F - D - D - †

4 destination states

12 or 18 months

		T					
		T-1		R	F	D	†
3 states of origin	R	558	257	43	41	899	
	F	195	639	142	112	1088	
	D	14	49	248	109	420	
						2407	

Transition matrix

1st order Markov Chain

12 or 18 months

T-1 \ T	R	F	D	+	Total
R	0.62	0.29	0.05	0.04	1
F	0.18	0.59	0.13	0.10	1
D	0.03	0.12	0.59	0.26	1

Transition matrix

1st order Markov Chain

		4 destination states				Precision
		R	F	D	+	
3 states of origin	12 or 18 months T-1 \ T	0.62	0.29	0.05	0.04	± 0.04
	R	0.18	0.59	0.13	0.10	± 0.03
	F	0.03	0.12	0.59	0.26	± 0.06

Transition matrix

on 3 successive waves:
2nd order Markov Chain

4 destination states

		4 destination states				Total	
T-2	T-1 \ T	R	F	D	+		
9 trajectories of origin	R	R	0.69	0.23	0.03	0.04	1
	R	F	0.29	0.51	0.10	0.10	1
	R	D	0.17	0.14	0.53	0.17	1
	F	R	0.37	0.52	0.05	0.06	1
	.	.	Continued...				
	.	.					
	.	.					

Double Chain Markov Model (DCMM)

MC



A unique transition matrix used for the whole sample
(Homogeneous)

DCMM



Different transition matrices C_1, C_2, \dots
can be used to model different portions of the observed data
(Non-homogeneous)

Double Chain Markov Model (DCMM) with SWILSO-O

Two transition matrices
 C_1 and C_2

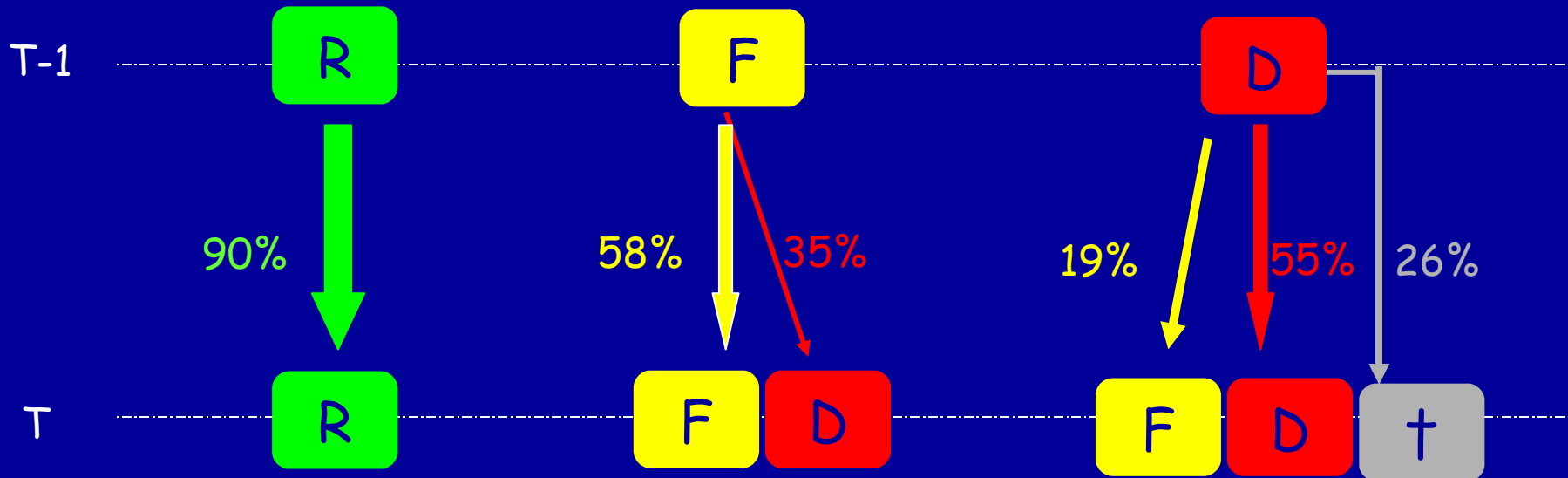
 This DCMM helps finding typologies of trajectories

Double Chain Markov Model (DCMM)

Typologies of trajectory



$T-1 \backslash T$	R	F	D	†
R	0.90	0.06	0.00	0.04
F	0.00	0.58	0.35	0.07
D	0.00	0.19	0.55	0.26



Double Chain Markov Model (DCMM)

Typologies of trajectory

 C_2	$T-1 \backslash T$	R	F	D	+
	R	0.50	0.38	0.07	0.05
F	0.23	0.59	0.07	0.11	
D	0.06	0.07	0.62	0.26	

MC and DCMM can be computed using
MARCH available at
<http://www.andreberchtold.com>

Markov Chains versus Regressions: differences

Markov Chains	Regressions
Non-parametric; No assumption	Parametric; Assumptions should be verified a posteriori
No error terms	Estimation of the probability of Y_t
Description of every possible transitions	Limited number of states for Y_t (multivariate, ordinal)

Markov Chains versus Regressions

Similarities

	Markov Chains	Regressions
Temporal homogeneity or non-homogeneity	Comparison between homogeneous and non-homogeneous MCs with BIC	Interaction between the probability of transition and Time $Y_t = f(Y_{t-1}, \text{wave}, Y_{t-1} * \text{wave})$
Time dependence of Y_t	Comparison between first-order and second-order homogeneous MC	Lag as covariables $Y_t = f(Y_{t-1}, Y_{t-2})$
Covariables	Yes - Mixture Transition Distribution	Yes

Conclusions

- ❖ MC: alternative when the assumptions of regressions are not acceptable
- ❖ MC: describes all possible transitions for exploratory analyses
- ❖ DCMM: helps finding typologies of trajectories

Thank You

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This presentation can be found at:
<http://www.unige.ch/centres/cig/>

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References

Berchtold A., Sackett, G. (2002) Markovian Models for the Developmental Study of Social Behavior, Am. J. of Primatology 58:149-167

Brémaud (1999) Markov Chains, Springer

Ross (2002) Introduction to Probability Models, 8th edition, Academic Press

Double Chain Markov Model (DCMM)

Matrix A

	T	C ₁	C ₂
T-1			
C ₁		1	0
C ₂		0	1

$$\text{BIC} = -2 \text{LL} + p \times \log(n)$$

DCMM:

computed by maximization of the log-likelihood using an EM algorithm. It reestimates successively the A and C transition matrices of the model until convergence, with the highest probability of generating the data.

Test of covariables

1st order MC

Men

T-1	T	R	F	D	†
R		0.63	0.28	0.04	0.05
F		0.21	0.53	0.16	0.10
D		0.05	0.11	0.53	0.32

Women

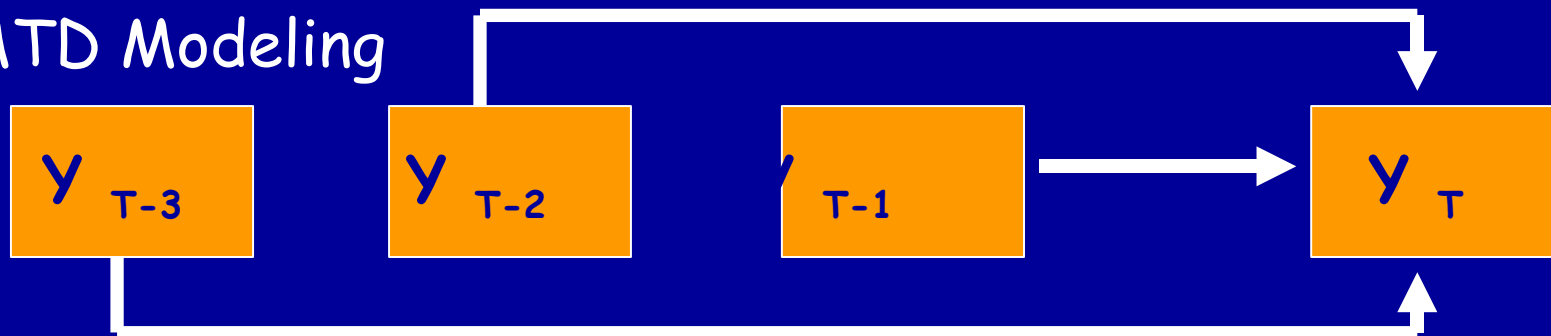
T-1	T	R	F	D	†
R		0.61	0.34	0.03	0.02
F		0.13	0.64	0.16	0.08
D		0.01	0.16	0.66	0.16

Mixture Transition Distribution model

3rd order Markov Chain



MTD Modeling



Using theoretically defined models: Hypothesis testing

$$\text{BIC} = -2 \text{ LL} + p \times \log(n)$$

BIC (1) with the « real » data

BIC (2) with a theoretical model