

Modeling Dynamic Processes with Panel Data: An Application of Continuous Time Models to Prevention Research

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Introduction

- Continuous Time Models and Prevention/Intervention Research²

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- Open Systems: Dynamic/Process Errors change your model
- Dynamical Systems: Mean differences are not enough

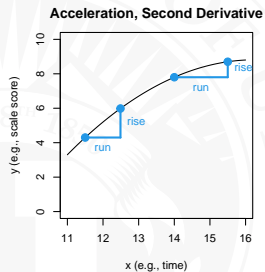
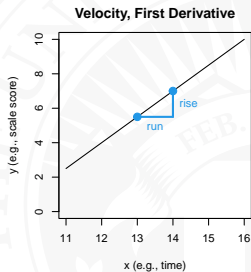
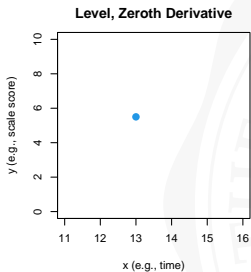
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- Dynamical Systems: Mean differences are not enough
- Substantive Example: What are the hallmark(s) of a successful intervention?

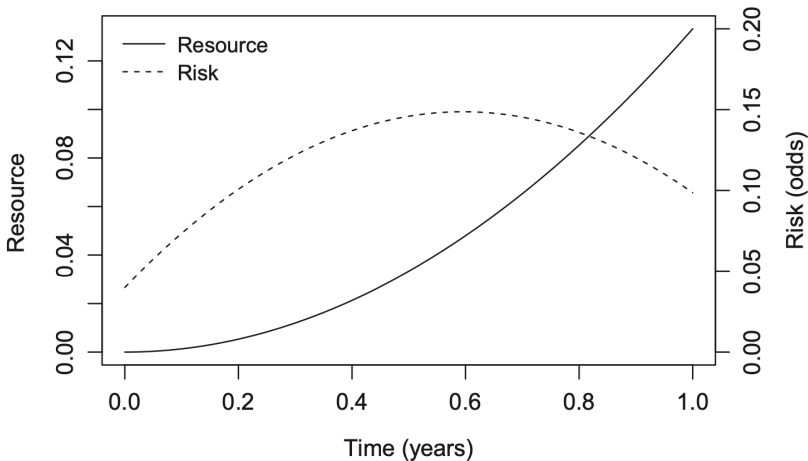
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Derivatives



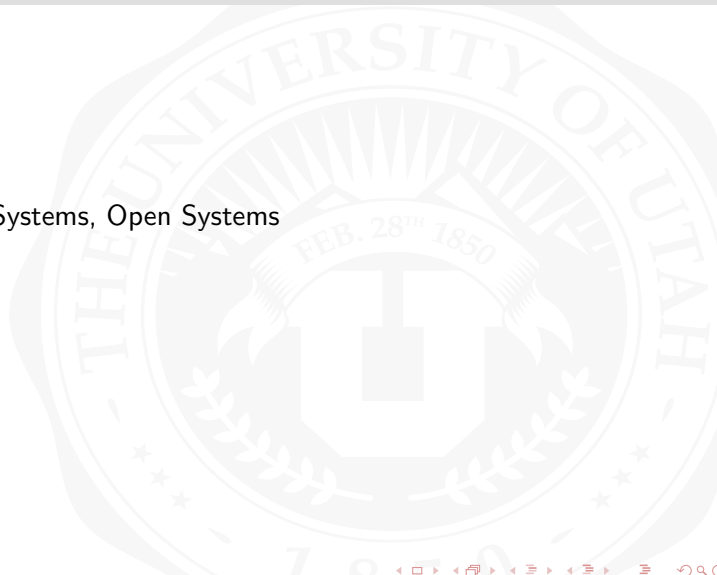
- Relations between derivatives, Differential equations

Differential Equations Modeling



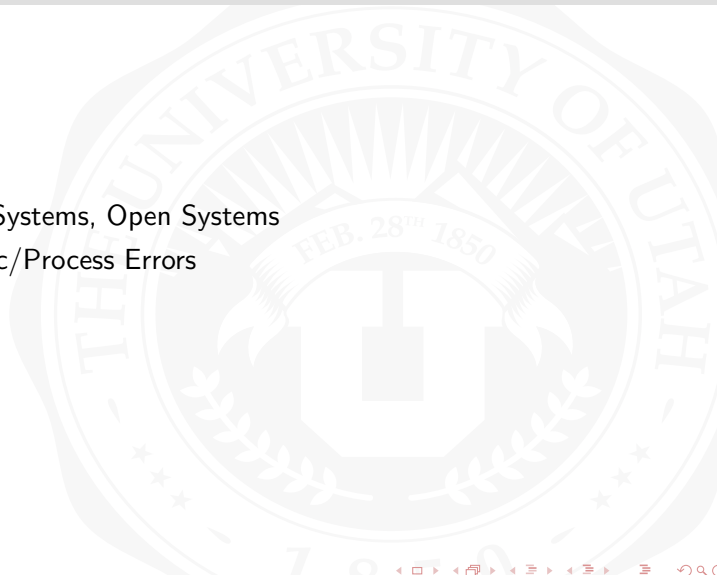
Open Systems

- Closed Systems, Open Systems



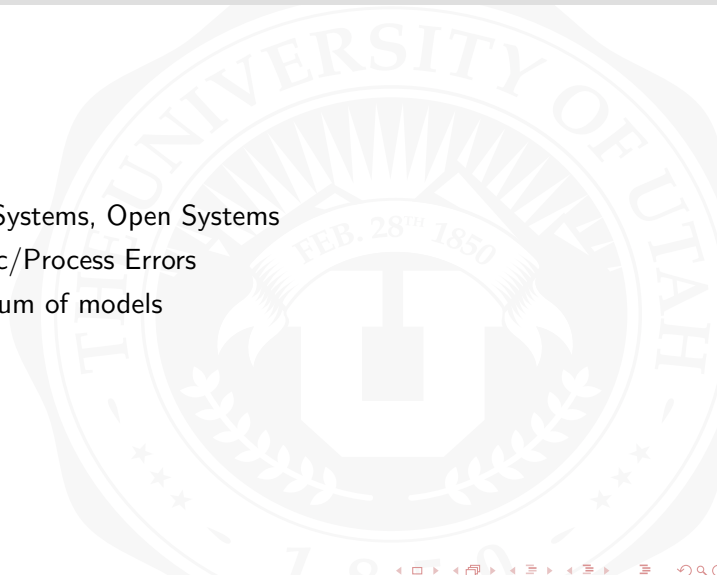
Open Systems

- Closed Systems, Open Systems
- Dynamic/Process Errors

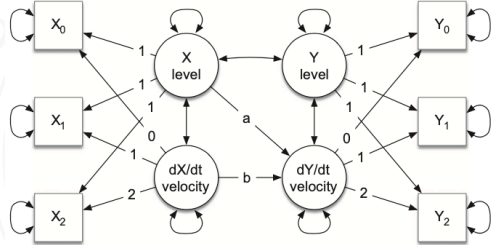
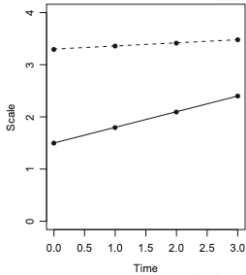


Open Systems

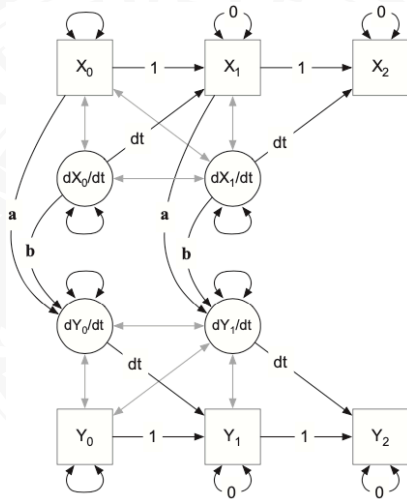
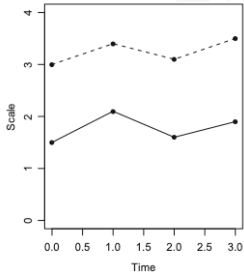
- Closed Systems, Open Systems
- Dynamic/Process Errors
- Continuum of models



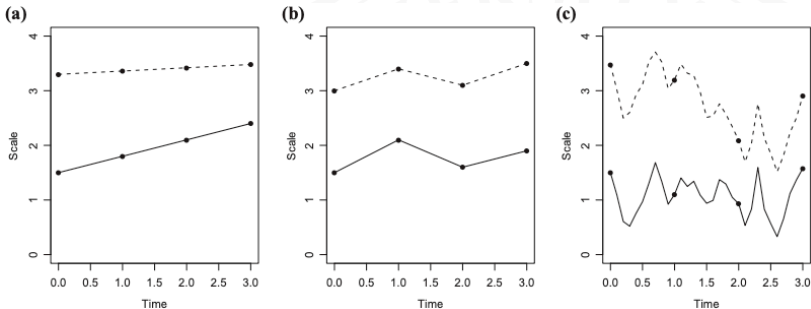
Small Dynamic Errors



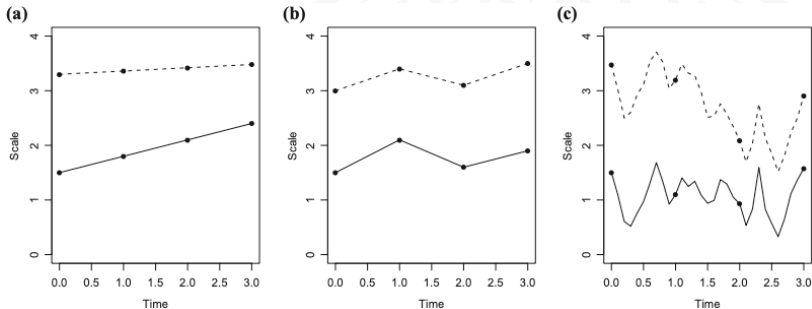
Moderate Dynamic Errors



Frequent Dynamic Errors

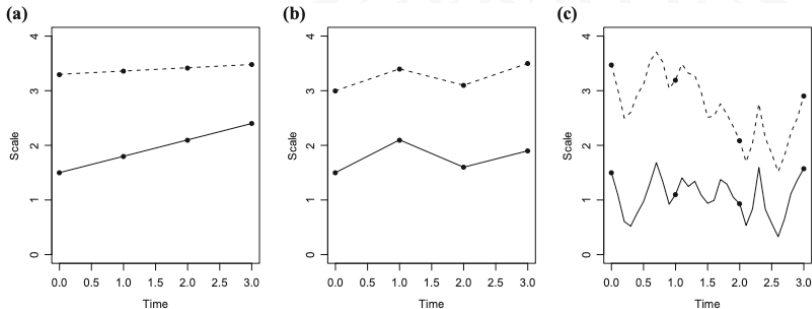


Frequent Dynamic Errors



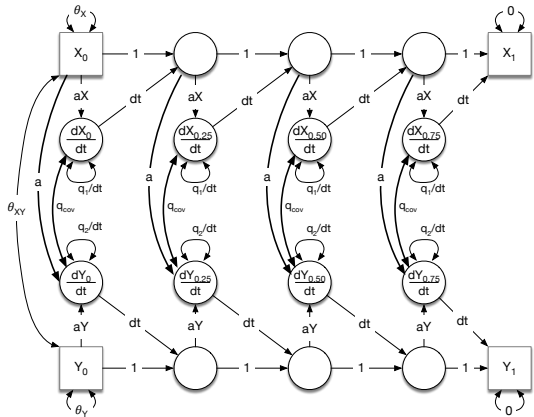
- Changes are frequent relative to sampling rate

Frequent Dynamic Errors

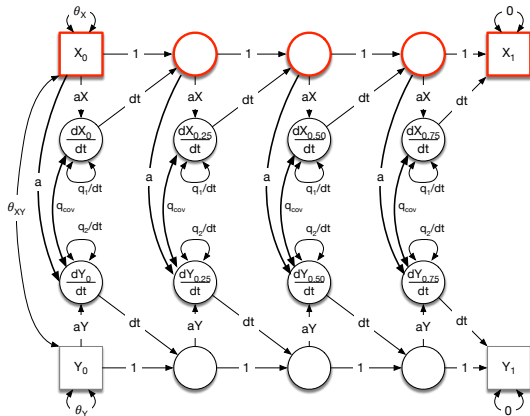


- Changes are frequent relative to sampling rate
- Ongoing interaction between constructs
- Stochastic Differential Equations

Stochastic Differential Equations

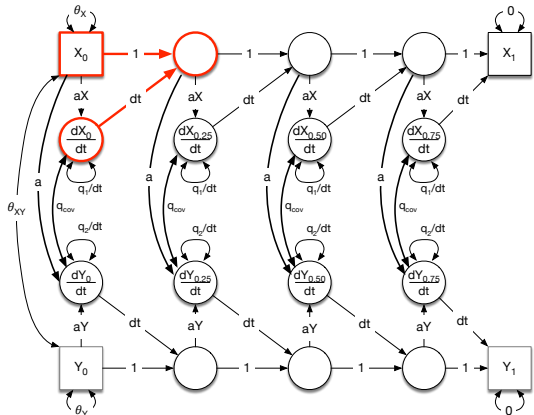


Stochastic Differential Equations



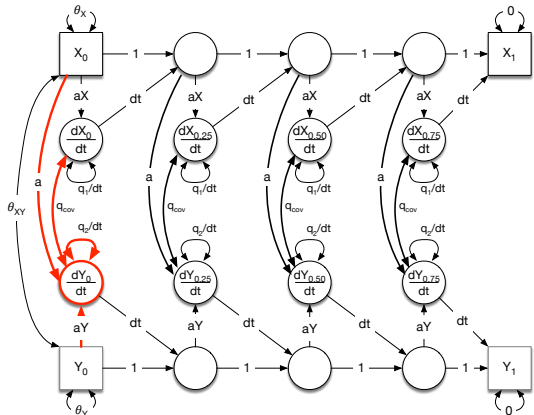
- Constructs exist between observed values

Stochastic Differential Equations



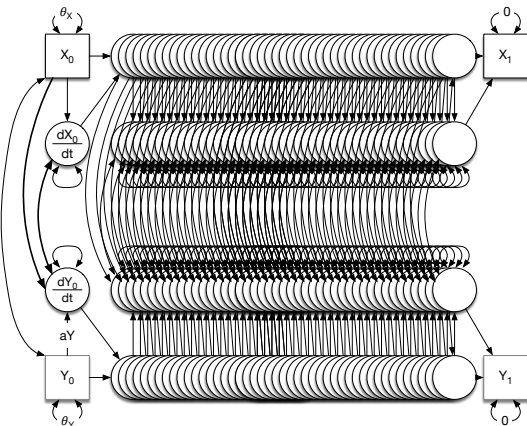
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Stochastic Differential Equations



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- Value is equal to the prior value plus change
- Change is defined (stochastic perturbations, correlated change, differential equation modeling)

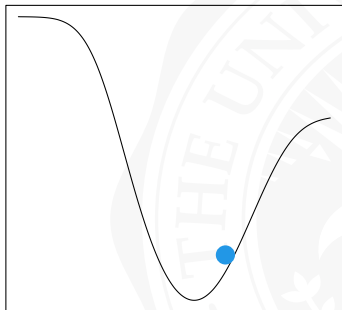
Stochastic Differential Equations



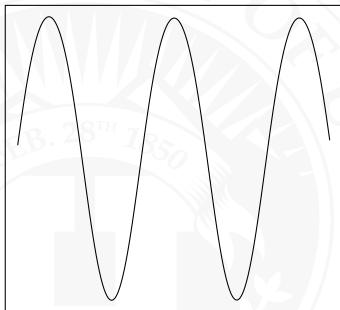
- Constructs exist between observed values
- Value is equal to the prior value plus change
- Change is defined (stochastic perturbations, correlated change, differential equation modeling)
- Number of intermediate latent occasions goes to infinity^a

^aInSDE; Deboeck & Boulton, 2016  

Attractor Dynamics



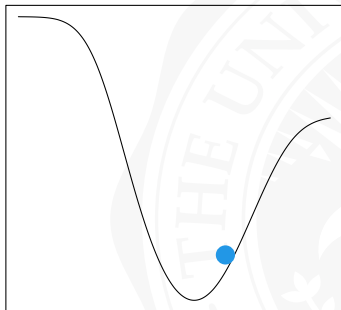
Scale Score



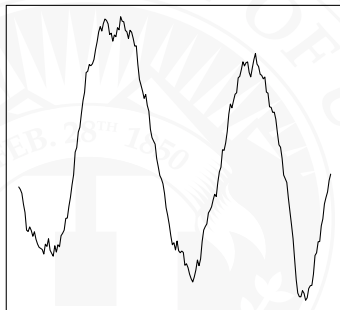
Scale Score

Time

Attractor Dynamics



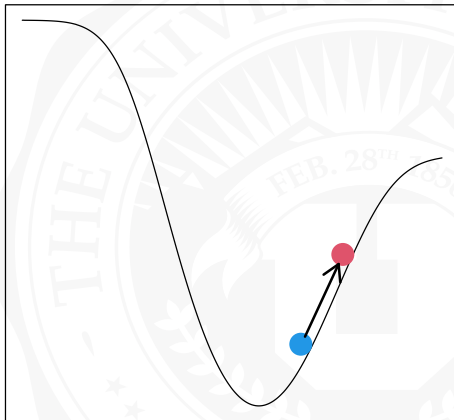
Scale Score



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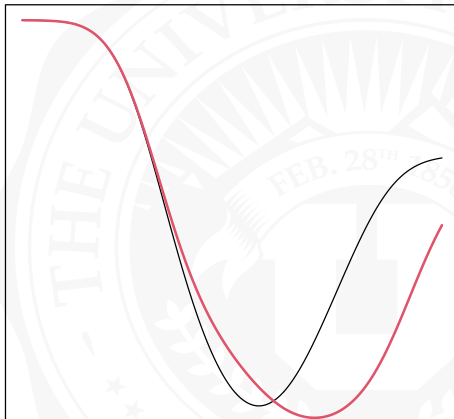
Time

Producing A Change in Mean Score



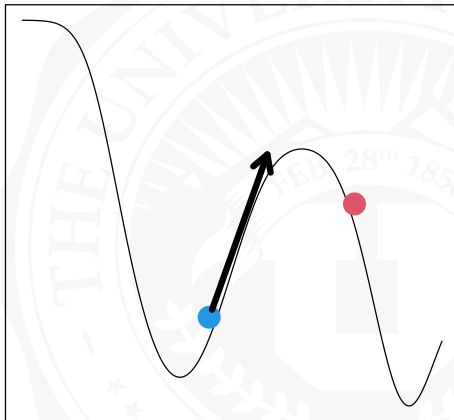
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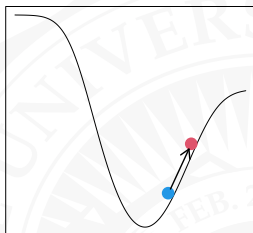
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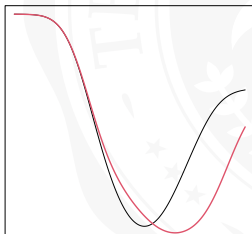


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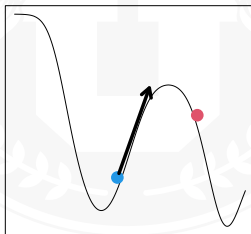
What is the effect of an intervention?



Scale Score



Scale Score



Scale Score

Reading Intervention

- Frequent Dynamic Errors (SDEs)
- Changing/Different Attractors (Differential Equations)

³Compas, B. E. et al., 2009, 2010, 2011, 2015

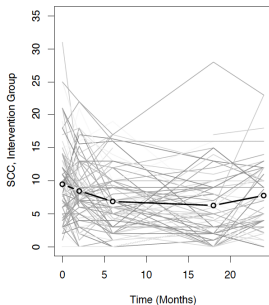
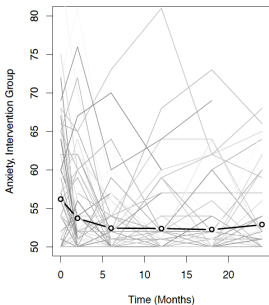
Reading Intervention

- Frequent Dynamic Errors (SDEs)
- Changing/Different Attractors (Differential Equations)
- Efficacy of a cognitive-behavioral preventive intervention for children of parents with a history of depression³
- Active intervention consisted of a 12-session program (8 weekly sessions, 4 month booster) teaching children skills to cope with stress related to their parents' depression
- $n = 122$ active intervention, $n = 120$ control group

³Compas, B. E. et al., 2009, 2010, 2011, 2015

Reading Intervention

- Data were collected at 6 times: baseline (before the intervention) and 2, 6, 12, 18 and 24 months after baseline.
- Children's reports of secondary control coping (SCC)⁴, and anxiety/depression⁵



⁴Compas, Connor, Osowiecki, & Welch, 1997; Compas, Connor, Saltzman, Thomsen, & Wadsworth, 1999

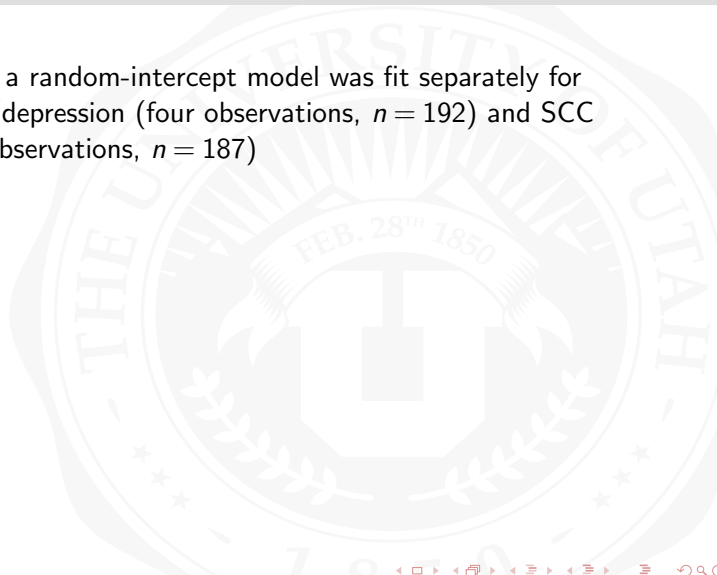
⁵Achenbach, 1991

Random Intercept, Months 6 to 24



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- Initially, a random-intercept model was fit separately for anxiety/depression (four observations, $n = 192$) and SCC (three observations, $n = 187$)



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- The random intercept-only models fit the data relatively well and seem to be a reasonable approximation of the trajectories between 6 and 24 months ($CFI_{ANX} = .978$, $RMSEA_{ANX} = .062$; $CFI_{SCC} = .962$, $RMSEA_{SCC} = .079$)

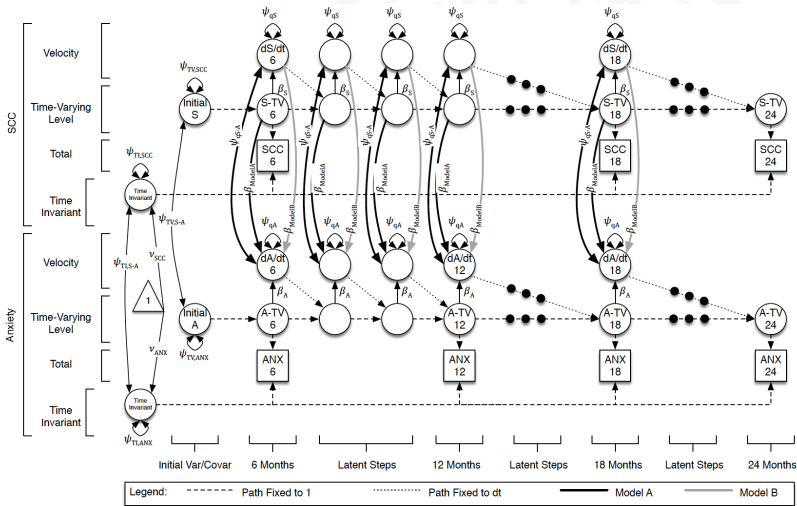
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- Does not describe how individuals vary around their individual intercepts
- Random mean model, SDE, two-group SEM

Substantive Models



Results

- AIC/BIC did not show a clear preference for one model over the other
- Random mean anxiety/depression lower in intervention group, and less variance (both models)

Results

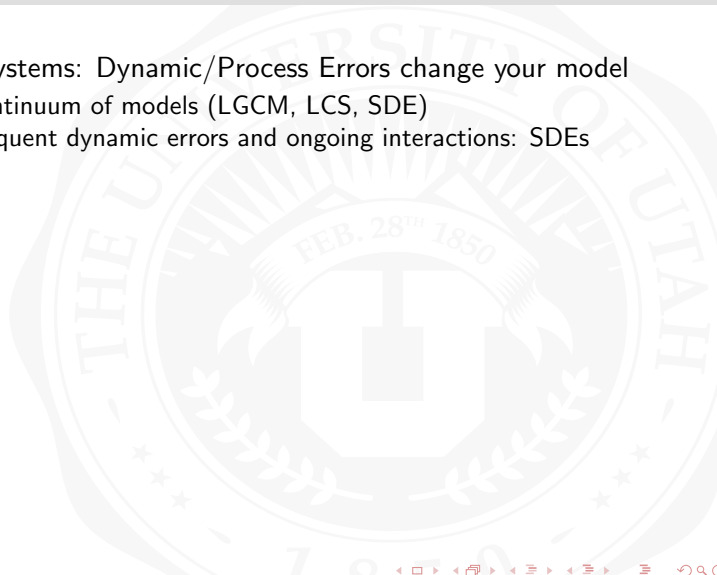
Parameter	Model A (level-velocity)			
	Readings	Group Diff	S.E. Diff	p Value
Random effects (time-invariant)				
Mean anxiety/depression (ν_{ANX})	55.50	-2.82	0.78	.000
Mean SCC (ν_{SCC})	7.84	-0.89	0.71	.208
Variance anxiety/depression ($\psi_{TI,ANX}$)	33.08	-22.29	7.60	.003
Variance SCC ($\psi_{TI,SCC}$)	20.30	-10.85	6.58	.099
Covariance anxiety, SCC ($\psi_{TI,S-A}$)	14.96	-8.86	5.11	.083
TV initial covariance				
Variance anxiety/depression ($\psi_{TV,ANX}$)	15.39	-5.50	5.59	.326
Variance SCC ($\psi_{TV,SCC}$)	-0.09	1.89	3.95	.632
Covariance anxiety/depression, SCC ($\psi_{TV,S-A}$)	12.54	4.59	6.54	.483
Derivative relations				
Anxiety/depression level-velocity (β_A)	-1.43	0.60	0.63	.342
SCC level-velocity (β_S)	-1.69	1.03	0.89	.244
SCC level to anxiety/depression velocity (β_{ModelA})	-0.82	1.17	0.90	.195
SCC velocity to anxiety/depression velocity (β_{ModelB})	—	—	—	—
Stochastic perturbations				
Variance anxiety/depression (ψ_{qA})	78.02	-53.55	3.22	.021
Variance SCC (ψ_{qS})	39.59	-10.36	7.34	.550
Covariance anxiety/depression, SCC (ψ_{qS-A})	36.12	-31.39	5.35	.041

Results

Parameter	Model B (velocity–velocity)			
	Readings	Group Diff	S.E. Diff	p Value
Random effects (time-invariant)				
Mean anxiety/depression (ν_{ANX})	55.49	–2.83	0.79	.000
Mean SCC (ν_{SCC})	7.80	–0.79	0.71	.267
Variance anxiety/depression ($\psi_{TI,ANX}$)	33.32	–22.88	8.15	.005
Variance SCC ($\psi_{TI,SCC}$)	20.94	–8.29	6.56	.206
Covariance anxiety, SCC ($\psi_{TI,S-A}$)	14.93	–5.64	4.83	.243
TV initial covariance				
Variance anxiety/depression ($\psi_{TV,ANX}$)	25.30	–12.68	9.17	.166
Variance SCC ($\psi_{TV,SCC}$)	–11.50	9.09	6.26	.146
Covariance anxiety/depression, SCC ($\psi_{TV,S-A}$)	13.03	2.40	5.22	.646
Derivative relations				
Anxiety/depression level–velocity (β_A)	–1.55	0.84	0.69	.225
SCC level–velocity (β_S)	–9.30	8.27	19.45	.671
SCC level to anxiety/depression velocity (β_{ModelA})	—	—	—	—
SCC velocity to anxiety/depression velocity (β_{ModelB})	0.90	–0.72	0.28	.010
Stochastic perturbations				
Variance anxiety/depression (ψ_{qA})	45.29	–21.75	15.29	.155
Variance SCC (ψ_{qS})	182.52	–147.28	356.9	.680
Covariance anxiety/depression, SCC (ψ_{qS-A})	—	—	—	—

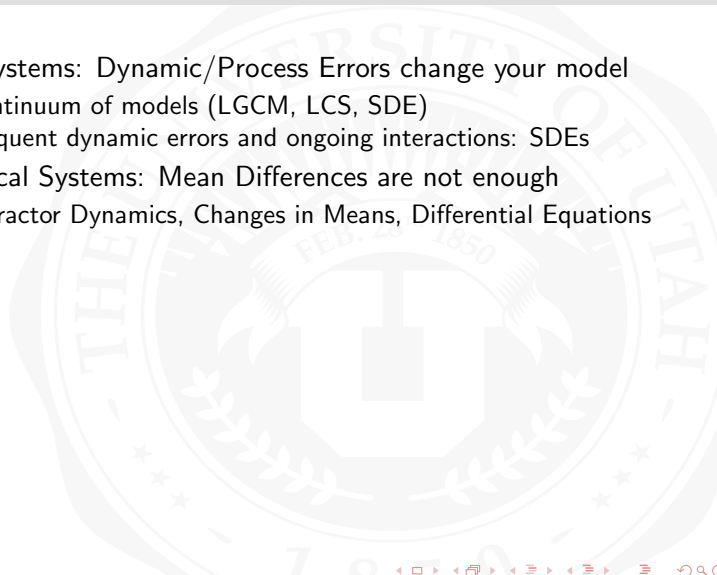
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 - Different underlying dynamic processes with limited data

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 - Different underlying dynamic processes with limited data
 - Intervention may have affected more than the mean and variance of individual means
 - Relations within constructs, stochastic perturbations, different coupled relations between variables



- Paper: Deboeck, P. R., Cole, D. A., Preacher, K. J., Forehand, R. & Compas, B. E. (2021). Modeling Dynamic Processes with Panel Data: An Application of Continuous Time Models to Prevention Research. *International Journal of Behavioral Development*, 45(1), 28–39.
- Derivatives Language: Deboeck, P. R., Nicholson, J. S., Kouros, C., Little, T. D., & Garber, J. (2015). Integrating developmental theory and methodology: Using derivatives to articulate change theories, models, and inferences. *Applied Developmental Science*, 19, 217–231.
- SDEs: Deboeck, P. R., & Boulton, A. J. (2016). Integration of stochastic differential equations using structural equation modeling: A method to facilitate model fitting and pedagogy. *Structural Equation Modeling*, 23, 888–903.