Phenomics

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Outline

- Factor Analysis and Measurement Invariance
- Testing Hypotheses about Gene Action: FTND
- Drug factors vs. Symptom factors
- Genomewide Structural Equation Modeling
- Occult Subtypes
- Obligate Missingness
- Age differences decay correlations
- Future Directions including GREML

Measurement Invariance: Factor Model



Usually want to know about F, the latent factor!

Indirect measurement

MNI Causes Errors of Inference

- Sum Scores & Factor Scores Depend on Model
- Item-level Differences May:
 - Invalidate Group Mean Tests (Association even)
 - Invalidate Group Variance Tests
- MI Still Rarely Tested

Invariance: Five Potential Types of Difference

- Factor Variances
- Factor Means
- Factor Loadings
- Item Variances
- Item Means



Invariance Models of Factor-Level Effects









1. No Covariates

2. Age/Sex on Factor Mean 3. Age/Sex on Factor Variance 4. Age/Sex on Factor Mean and Variance

Correlations across Substances: Add Health

	Stimulants	Tranquilizers	Marijuana
Stimulants			
Tranquilizers	0.74		
Marijuana	0.63	0.66	
Factor Loadings	0.84	0.87	0.75

Medland & Neale (2010) An integrated phenomic approach to multivariate allelic association. European Journal of Human Genetics 18:233–239

DRD2 Association Results (Add Health)

- Univariate associations
 - Stimulants: $\chi^2 = 3.88$, $\beta = -.18$, p < .05
 - Tranquilizers: $\chi^2 = 1.65$, $\beta = .13$, NS
 - Marijuana: $\chi^2 = 2.60, \beta = .11, NS$
- Factor level association
 - χ²=0.65, kF= .06, NS
- Item level association
 - χ²=13.91 (3df; p < 0.005)

 $\beta_{\text{Stimulants}} = -0.19$

 $\beta_{\text{Tranquilizers}} = 0.14$ $\beta_{\text{Marijuana}} = 0.11$





MI Application: National Survey of Drug Use in Households (NSDUH)

- Substance Abuse and Mental Health Services Administration (SAMSA) regular data collection
- ~50,000 persons per assessment
- Face-to-face Interviews(!)
- Audio-Computer-Assisted Testing

Map Items to DSM-IV Substance Abuse and Dependence Criteria

- A1 During the past 12 months, did using marijuana or hashish cause you to have serious problems like this either at home, work, or school?
- A2 During the past 12 months, did you regularly use marijuana or hashish and then do something where using marijuana or hashish might have put you in physical danger?
- A3 During the past 12 months, did using marijuana or hashish cause you to do things that repeatedly got you in trouble with the law?
- A4 Did you continue to use marijuana or hashish even though you thought it caused problems with family or friends?

DSM-IV Dependence Criteria

- D1 During the past 12 months, did you need to use more marijuana or hashish than you used to in order to get the effect you wanted?
- D3 Were you able to keep to the limits you set, or did you often use marijuana or hashish more than you intended to?
- D4 During the past 12 months, did you want to or try to cut down or stop using marijuana or hashish?
- D5 During the past 12 months, was there a month or more when you spent a lot of your time getting or using marijuana or hashish?
- D6 This question is about important activities such as working, going to school, taking care of children, doing fun things such as hobbies and sports, and spending time with friends and family.
 - During the past 12 months, did using marijuana or hashish cause you to give up or spend less time doing these types of important activities?
- D7 Did you continue to use marijuana or hashish even though you thought it was causing you to have physical problems?

Test of Factor Loading Invariance: Marijuana in NSDUH

Model	Npar	Comparison Model	Likelihood	df	AIC	diffLL	diffdf	р	Age Effect	Sex Effect
1. No Covariates	20	NA	62514	78204	-93894	NA	NA	NA	NA	NA
2. Age/Sex on Factor Mean	22	1	62009	78202	-94395	505	2	<.0001	-3.85567	-0.09112
3. Age/Sex on Factor Variance	22	1	62480	78202	-93924	33.84	2	<.0001	0.70624	0.40302
4. Age/Sex on Factor Mean and Variance	24	1	61893	78200	-94507	620.4	4	<.0001		
5. Age/Sex on Mean and Loadings	42	4	61801	78182	-94563	92.34	18	<.0001		
6. Age/Sex on Thresholds and Factor Variance	42	4	61802	78182	-94562	91.2	18	<.0001		

Test of Item Mean Invariance: Marijuana in NSDUH

Model	Npar	Comparison Model	Likelihood	df	AIC	diffLL	diffdf	p	Age Effect	Sex Effect
1. No Covariates (1f model)	20	NA	62514	78204	-93894	NA	NA	NA	NA	NA
2. Age/Sex on Factor Mean	22	1	62009	78202	-94395	505	2	<.0001	-3.85567	-0.09112
3. Age/Sex on Factor Variance	22	1	62480	78202	-93924	33.84	2	<.0001	0.70624	0.40302
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5. Age/Sex on Mean and Loadings	42	4	61801	78182	-94563	92.34	18	<.0001		
6. Age/Sex on Thresholds and Factor Variance	42	4	61802	78182	-94562	91.2	18	<.0001		

Test of Item Mean Invariance: Marijuana in NSDUH

- Strong evidence of MNI with respect to age and sex
- Examine individual items
- Four column heatmap for significance of effects
 - Item Means & Factor Variances
 - Sex and Age
- Compare across self-reported race

Age

Entire Sample

MRJA1 -5.334.68 10.3 -14.28MRJA2 0.04 -0.05-1.821.81 MRJA3 -20.7-5.94-2.280 MRJA4 -0.41-2.131 0 MRJD1 0.12 -1.32-11.24-11.96MRJD3 -0.72-1.5314.87 -2.74MRJD4 0.69 -0.3728.87 12.54 -11.526.18 -0.09 0 1.8 14.79 -14.2-0.09MRJD7 0.81 -1.9416.17 -9.31sex age age M sex ≥ п п

Sex

Work Danger Law Friends ΤΟ >Intend TryCut TimeGet MRJD5 TimeOther< MRJD6 PhysProb

+/- sign denotes direction

Ane

Sex

White

	3	3.46	-3.85	-2.66	MRJA1
	1.69	-0.11	-6.87	1.34	MRJA2
	-7.52	-0.12	-5.82	-2.2	MRJA3
	1.68	1.8	-1.21	-1.04	MRJA4
	0.09	-1.93	-10.72	10.29	MRJD1
0000	-0.05	-0.23	14.99	-6.57	MRJD3
	2.81	-0.1	23.1	-15.97	MRJD4
	-6.33	1.12	-0.25	0	MRJD5
	-0.04	0.9	17.45	-12.71	MRJD6
	0.02	-1.68	12.77	-12.62	MRJD7
00000	sex	sex	age	age	
	Σ		Σ		

Work Danger Law Friends Tol >Intend TryCut TimeGet TimeOther< PhysProb

African American

	Sex		ge		
2.52	6.42	-1.67	-2.87	MRJA1	
-0.04	-0.26	0.31	-0.21	MRJA2	
-15.34	-11.7	0.28	0.18	MRJA3	
-1.07	-2.63	0.58	-0.13	MRJA4	
0.58	-1.23	-0.15	0.04	MRJD1	
0	-2.01	-0.06	0.84	MRJD3	
-0.33	-0.64	1.35	0	MRJD4	
-5.1	4.49	-0.01	0.01	MRJD5	
10.34	6.78	-0.07	0.42	MRJD6	
0.11	0	0.17	-0.37	MRJD7	
M_sex	'L_sex	M_age	L_age		

Work Danger aw -riends ΤΟ >Intend TryCut TimeGet TimeOther< PhysProb

Aae

Sex

Hispanic

Wor	MRJA1	4.74	-12.16	-1.36	8.9
Dan	MRJA2	2.43	0.01	-0.06	-0.52
Law	MRJA3	-0.29	-1.36	1.87	-11.54
Frie	MRJA4	-0.12	-0.31	-2.95	0.37
Tol	MRJD1	1.92	-0.91	0.81	-0.28
>Int	MRJD3	0	2.77	0.19	-1.56
TryC	MRJD4	0.24	3.32	-0.22	0.13
Time	MRJD5	-0.19	0.57	1.26	-1.14
Time	MRJD6	-4.03	3.49	-1.26	0.5
Phy	MRJD7	-0.1	4.3	-0.07	1.27
		ge	ge	sex	sex
			Σ Β		Σ

ger nds end Cut eGet eOther< sProb

Estimating Factor Scores



ML Estimation of Factor Scores



Factor Score

Factor Score * Likelihood of items conditional on factor score

Items independent conditional on factor score: Means and variances change according to size of factor loadings





Drug vs Symptom Factors

- DSM III-R/IV drug abuse and dependence symptoms for cannabis, sedatives, stimulants, cocaine, opioids and hallucinogens
- 13 misuse symptoms measured across six illicit substance categories (78 items)
- 4179 males born 1940–1970 from the population-based Virginia Adult Twin Study of Psychiatric and Substance Use Disorders
- Confirmatory factor analyses tested specific hypotheses regarding the latent structure of substance misuse

Drug vs Symptom Factors



Clark, S. L., Gillespie, N. A., Adkins, D. E., Kendler, K. S., and Neale, M.

Neale, M. C. (2016). Across six Higit substance:

indicates novel dimensions of mis

32–40, PMCID: PMC4679450.



Drug vs Symptom Factors

χ^2	DF	<i>p</i> -Value	CFI	RMSEA
4175	2910	< 0.001	0.78	0.017
3647	2847	< 0.001	0.86	0.013
2966	2754	< 0.001	0.96	0.007
4598	2925	< 0.001	0.71	0.019
1209	156	< 0.001		
681	93	<0.001		
	χ ² 4175 3647 2966 4598 1209 681	χ^2 DF41752910364728472966275445982925120915668193	χ^2 DF <i>p</i> -Value41752910<0.001	χ^2 DF <i>p</i> -ValueCFI41752910<0.001

 Adding symptom factors dramatically improves fit

 Majority of variance in many Sx due to symptom not drug factor

Factor Score Notes

- Factor scores do not all have same error variance
- Factor scores of A, C & E components may correlate highly
- Latent trait may be non-normal (Schmitt et al 2006 Multiv Behav Res)
- Factor loadings (precision) may vary across the distribution and give spurious GxE results
- Variation may be discrete not continuous



AFQT

100 Items

Subscales 1 Arithmetic Reasoning 2 Mathematics Knowledge 3 Word Knowledge 4 Paragraph Comprehens.



AFQT: Overall Test Information Curve



More information at left

By design

Consequences for GxE?

Genome-wide SEM

Avoid problems with factor scores

Fit factor or growth curve models to ordinal data

Include effect of SNP on factor or items

Repeat for the other 8m-1 SNPs

Manhattan plot results

<u>http://goo.gl/f44UmD</u>



Verhulst, B, Maes, H, & Neale, M (In Press) GW-SEM: A Statistical Package to Conduct Genome-Wide Structural Equation Modeling. *Behavior Genetics*

Testing Hypotheses about Gene Action: FTND

Table 1: Percentage of Variance Accounted for by the SNP rs16969968 in Latent FTND and Measured CPD

Sample	Ν	FTND	Total CPD	Indirect Effect
Sage	$2,\!461$	0.46	1.70	0.08
Smoking Cessation (SC)	574	0.48	1.76	0.08
CIDR	296	0.50	1.85	0.08
COPD	$2,\!042$	0.45	1.67	0.08

Note: The Direct Effects of FTND and the Total and Indirect Effects on CPD are taken from the best fitting model (H_{1c}) .

 rs16969968 Neuronal acetylcholine receptor subunit α-5 CHRNA5 associated with both ND and CPD

- What is the mechanism of action?
- CPD mere symptom of FTND
- Increases CPD increases addiction?
- Feedback loop between CPD and addiction?

H₀ No Association



(a) H_0 : Path diagram showing the Null Model where rs16969968 is unrelated to latent FTND factor and CPD.

H_{1a} Factor Only



(b) H_{1a} : Path diagram for regression of the latent FTND factor on the SNP.

H_{1b} CPD Only



(c) H_{1b} : Path diagram for regression of CPD on the SNP

H_{1c} Factor & CPD



(d) H_{1c} : Path diagram for regression of the latent FTND factor and CPD on the SNP

H_{2a} CPD Only & CPD causes Factor



(e) H_{2a} : Path diagram for sequential effect of the SNP causing CPD, which causes Nicotine Dependence.



H_{2b} SNP to CPD & Reciprocal Factor



Smoke

when Ill 🛴

D_{COVj}

(f) H_{2b} : Path diagram for the the SNP causing CPD, which reciprocally causes Nicotine Dependence.

Results

H_{1c} CPD & Factor 270 45217.89 -18718.11	
H_{1a} Factor Only 269 45225.07 -18712.93 7.18 1 7.38e-	03^{+}
H_{1b} CPD Only 269 45238.02 -18699.98 20.13 1 7.25e-	06†
H_{2b} Reciprocal 270 45238.01 -18697.99	
H_{2a} Sequential 269 45238.02 -18699.98 0.01 1 0.93	+
H_0 No SNP Effect 268 45251.58 -18688.42 33.69 2 4.84e-	08†
13.57 2 1.13e-	03‡

No support for sequential or reciprocal hypotheses

 Rapid habit development & later assessment may obscure relationship of CPD to addiction

Verhulst, B, Neale, M, Chen, J & Chen, S (In Revision) Disentangling genetic influences on Nicotine Dependence and Quantity of Use

Factor Model Alternative: Mutualism



Identified with data from relatives MZ & DZ Twins Or adoptees needed for A/C resolution

What if Variation is Discrete?

Latent Class and Latent Profile Models

Factor Mixture Models

Latent Growth Curve Mixture Models

Regime Switching

Mixture Distributions

Pearson, K. (1894). Contributions to the mathematical theory of evolution. II. skew variation in homogeneous material. Philosophical Transactions of the Royal Society of London A, 186, 343-414.



 Skewness in a set of measurements of the ratio of forehead to body length of crabs

Two species or one?

Latent Class (Subgroup)

Class I probability þ



Class 2 probability (I-p)

μ₄IC2 ^μ5IC2 μ₆IC2 μ₇IC2 ^μ3IC2 μ1IC2 μ₂IC2 Dp Abu Abu Dp Dp Dp Abu 1 2 3 4 5 6 7 r2IC2 r3IC2 r_{1IC2} r4IC2 ^r5IC2 ^r7lC2 ^r6lC2 R Ŕ R R R R R D6 A2 A3 D4 D5 D7 10 1.0 1.0 1.0 1.0 1.0 1.0

Conditionally Independent?!

Expensive!

Published in final edited form as: Int J Methods Psychiatr Res. 2010 June ; 19(2): 63–73. doi:10.1002/mpr.301

Searching For Valid Psychiatric Phenotypes: Discrete Latent Variable Models

Jeannie-Marie S. Leoutsakos, PhD, MHS 1 , Peter P. Zandi, PhD, MHS 2 , Karen Bandeen-Roche, PhD 3 , and Constantine G. Lyketsos, MD, MHS 1,2

Factor Mixture Model

Class I probability þ

Class 2 probability (I-p)



Growth Curve Mixture

Class I probability þ

Class 2 probability (I-p)



Regime Switching Model



Year

Occult Heterogeneity

- Suppose >1 uncorrelated heritable phenotypes
- Generate data, equal proportions
- Estimate correlations in mixture sample
- Vary threshold
- Induces VD



Occult Heterogeneity

- VA underestimated
- VD overestimated
- More effect the rarer the disorder
- Effect attenuates
 with genetic
 correlation between
 subtypes



Additive & Dominance Variance Components

Obligate Missingness

- Estimating correlation between Stem and Probe
 - 3+ categories of Stem and at least 2 lead to probe
 - 2 binary Stem items and endorsing either or both = probe
 - Binary Stem but collected from relatives who correlate < 1</p>

Do not mark missing probes as zero! Usually causes inflated item correlations

Obligate Missingness

- Stem: Have you ever used cocaine? 0/1/2
- Probe: Was it difficult to cut down or quit?
- Probe items are MAR conditional on Stem being 1 or 2
- WLS but not ML drastically attenuate correlation estimate
- Must code probes as missing!



Figure 2: Attenuation of the estimated correlation using WLS based on the level of MAR missingness.

Genetic Heterogeneity with Age/Cohort

 Neuroticism within-person .6 correlation over 10 years

Twin studies show rG < 1 over time</p>

 Expressed genetic factors change during development

Substance Use

Different age, different genes?

The Decay in the Correlation over Time





Age-Related Decay of Correlation



Verhulst, B., Eaves, L. J., and Neale, M. C. (Jul 2014). Moderating the covariance between family member's substance use behavior. Behav Genet, 44(4):337–46.

 $\mathbf{Cov} = \mathbf{Acov}^* \mathbf{e}^{-|\Delta age|^* \alpha_a} + \mathbf{Ccov}^* \mathbf{e}^{-|\Delta age|^* \alpha_c} + \mathbf{Tcov}$

Application

Virginia 30,000 Data on Smoking

Twins, their parents, spouses, sibs and children Twins only here, N=14,763

Crude smoking measure (1980s)

(1) never smoked, (2)
used to smoke but gave it
up, (3) smoked on and off,
(4) smoked most of his/her
life.

Strong evidence of decay with age difference

Decay in the Correlation between First Degree Relatives as a Function of Age Difference



Age Difference (in Years)

Future Directions

- Use "GCTA" Genetic relatedness matrices in place of close family relatives
 - Technical challenges, invert 20k x 20k matrices or larger
- Extend tests for direction of causation with combined twin and Mendelian Randomization model
- Dynamical models for high density repeated measures

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