

NAACL - SIGTYP 2022

Typological Word Order Correlations with Logistic Brownian Motion



Almotion Bavaria



Technische Hochschule
Ingolstadt

14.07.2022

Kai Hartung

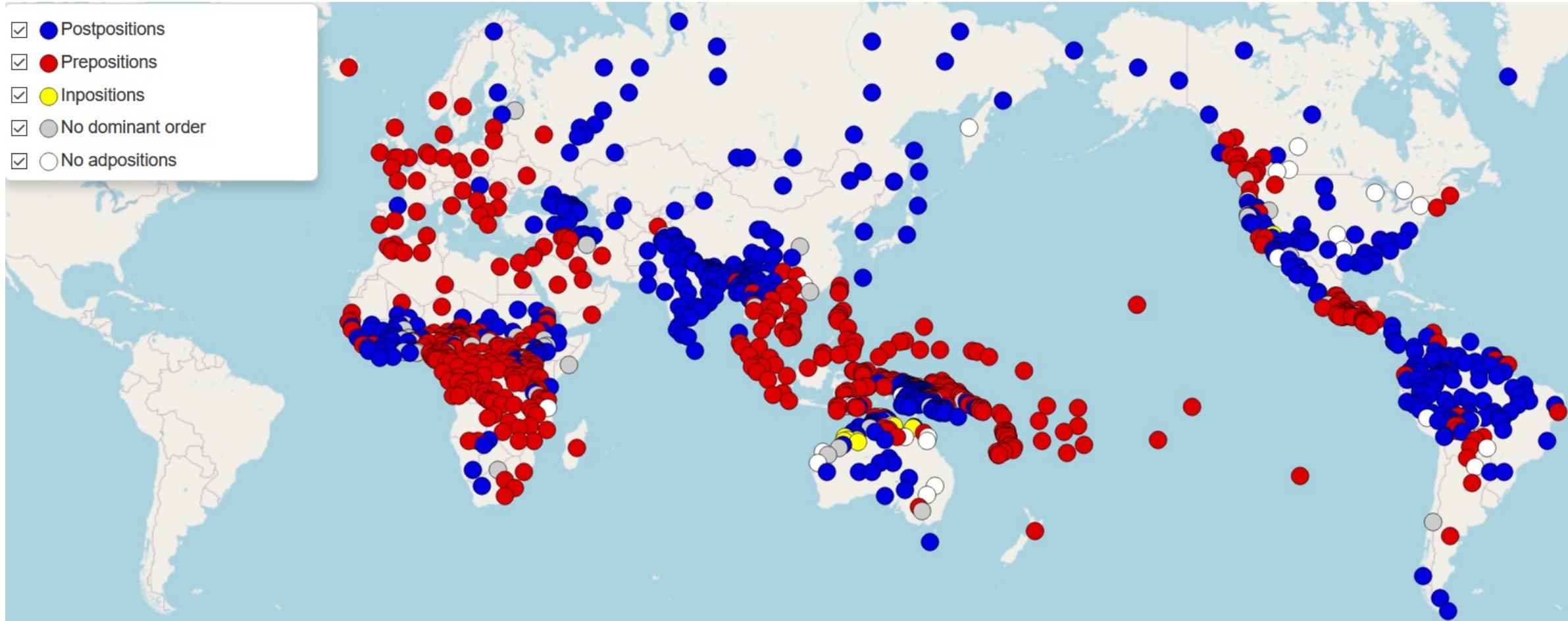
Gerhard Jäger

Sören Gröttrup

Munir Georges



Word-Order Traits in Languages



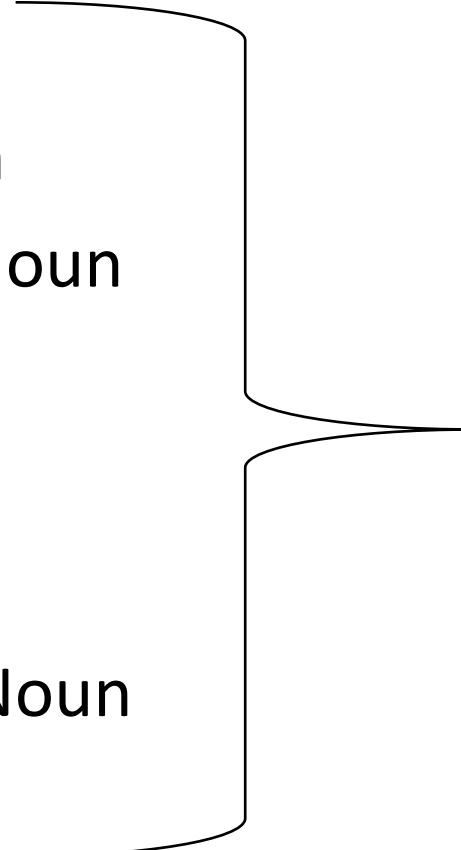
World Atlas of Languages (Dryer and Haspelmath, 2013)

Aim

- Probing for **universal correlation patterns** in the evolution of word-order traits
- Testing if **cross-family models** can capture correlation patterns not found in single-family models.
using a Logistic Brownian Motion Model

Word-Order Traits in Languages

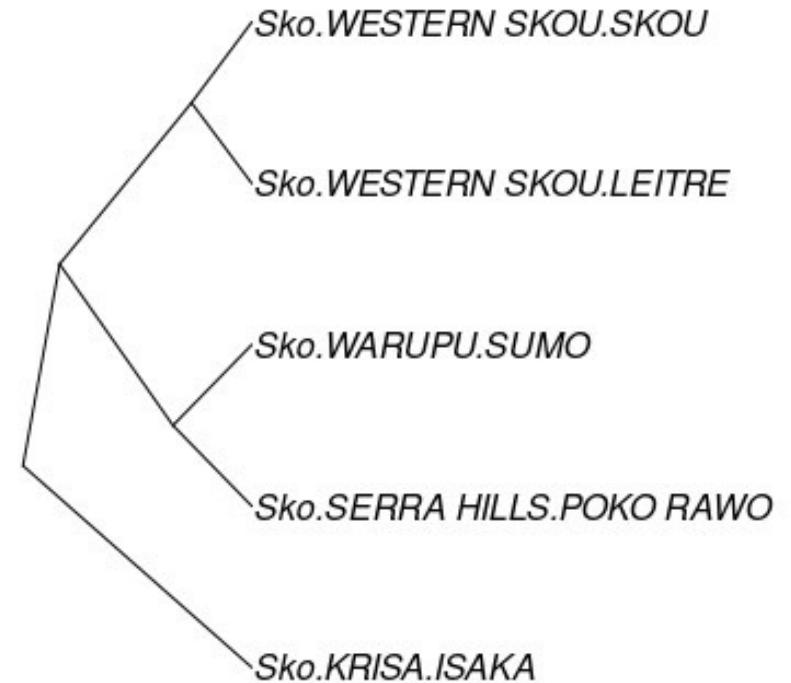
- Adjective-Noun
- Adposition-Noun
- Demonstrative-Noun
- Genitive-Noun
- Numeral-Noun
- Object-Verb
- Relative Clause-Noun
- Subject-Verb



28 **binary trait pair combinations**
derived from WALS (Dryer and Haspelmath, 2013)

Language Families

- Evolutionary history in the form of **phylogenetic trees**
- 33 language families
- 768 languages in total
- Provided by Jäger (2018)



Logistic Brownian Motion

$$x \sim \text{MultiNormal}(a, V),$$

trait values x ,

means a ,

Variance-Covariance matrix $V = R \otimes C$

Logistic Brownian Motion

$$x \sim \text{Binomial}(p)$$

$$\text{inv_logit}(p) \sim \text{MultiNormal}(a, V),$$

trait values x ,

value probabilities p ,

means a ,

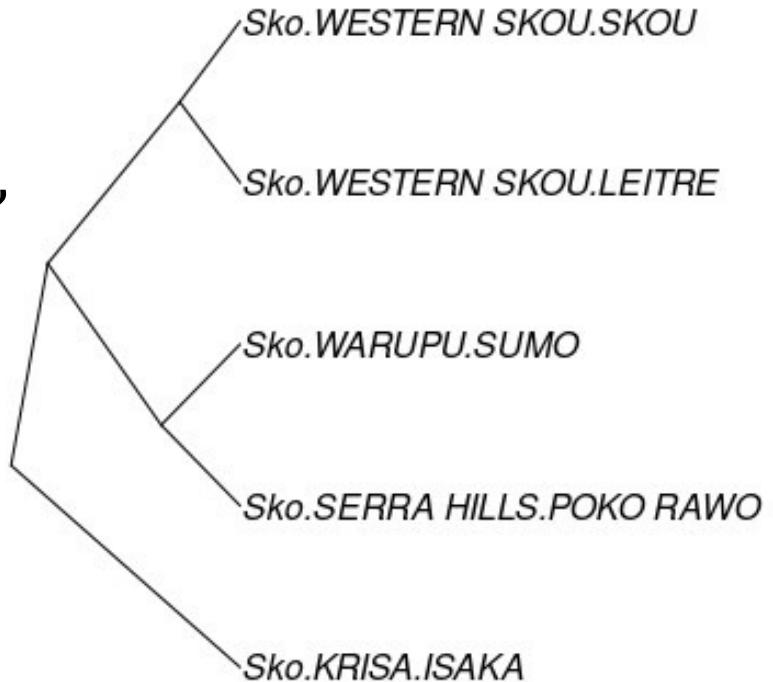
Variance-Covariance matrix $V = R \otimes C$

Logistic Brownian Motion

$x \sim Binomial(p)$

$inv_logit(p) \sim MultiNormal(a, V),$

$$V = R \otimes C$$

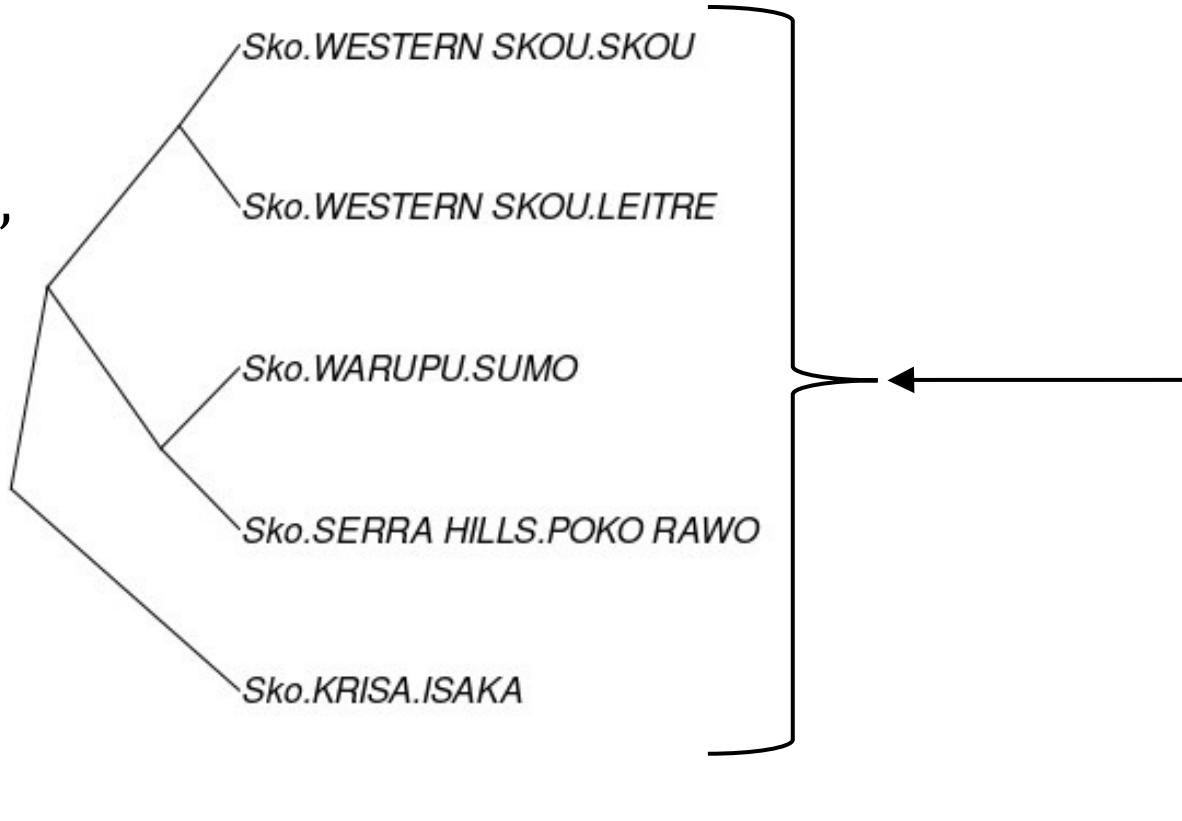


Logistic Brownian Motion

$$\mathbf{x} \sim \text{Binomial}(\mathbf{p})$$

$$\text{inv_logit}(\mathbf{p}) \sim \text{MultiNormal}(\mathbf{a}, \mathbf{V}),$$

$$\mathbf{V} = \mathbf{R} \otimes \mathbf{C}$$



probabilities \mathbf{p}
/trait values \mathbf{x}

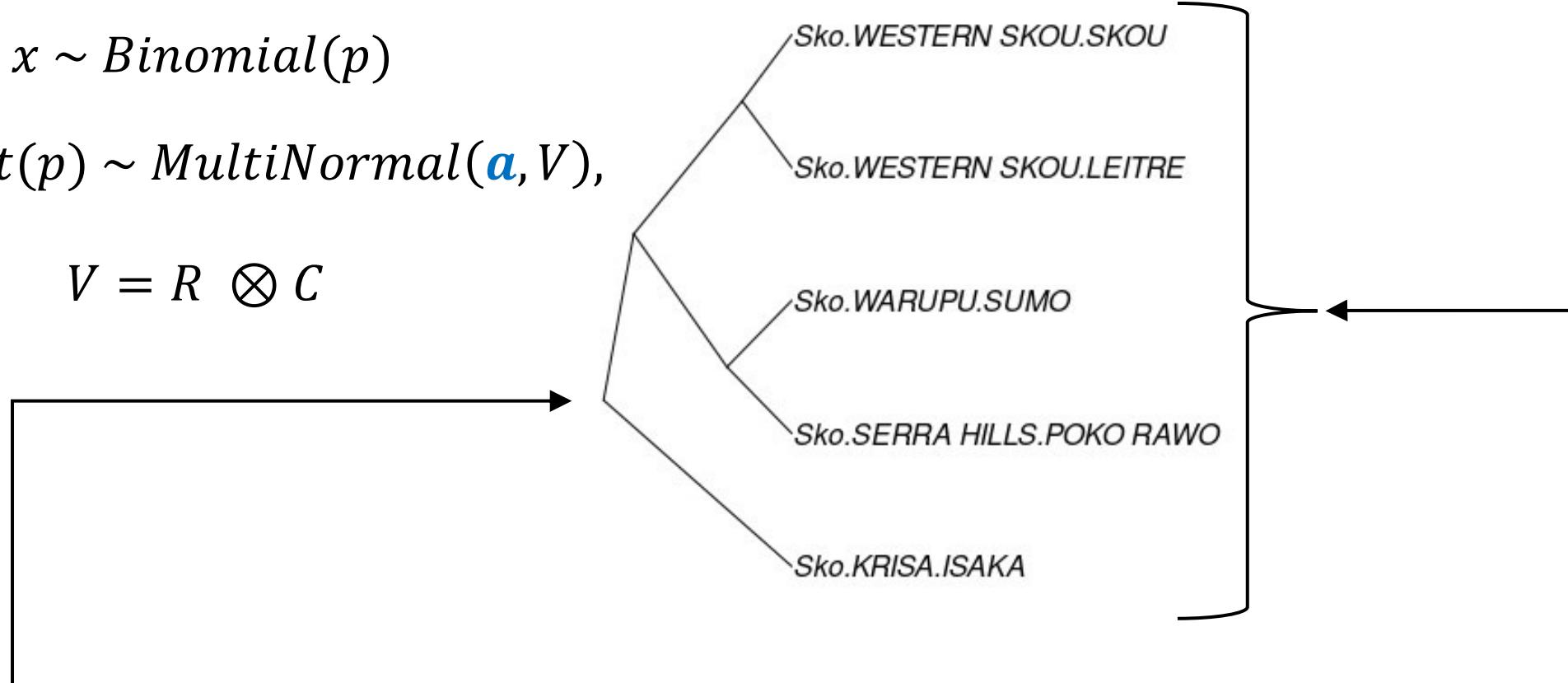
Logistic Brownian Motion

$$x \sim \text{Binomial}(p)$$

$$\text{inv_logit}(p) \sim \text{MultiNormal}(\alpha, V),$$

$$V = R \otimes C$$

means α
 /root values



probabilities p
 /trait values x

Logistic Brownian Motion

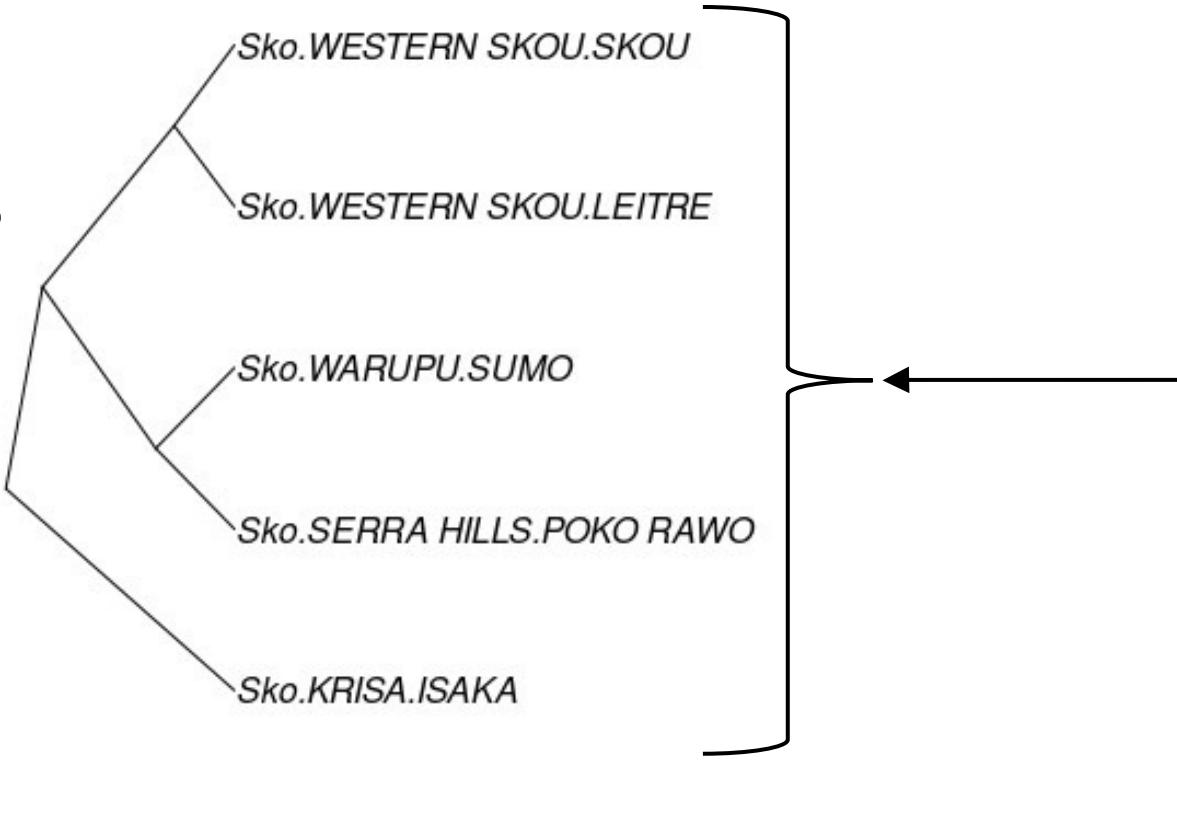
$$x \sim \text{Binomial}(p)$$

$$\text{inv_logit}(p) \sim \text{MultiNormal}(a, V),$$

$$V = R \otimes C,$$

\otimes : Kronecker product

means a
 /root values



probabilities p
 /trait values x

Logistic Brownian Motion

$$x \sim \text{Binomial}(p)$$

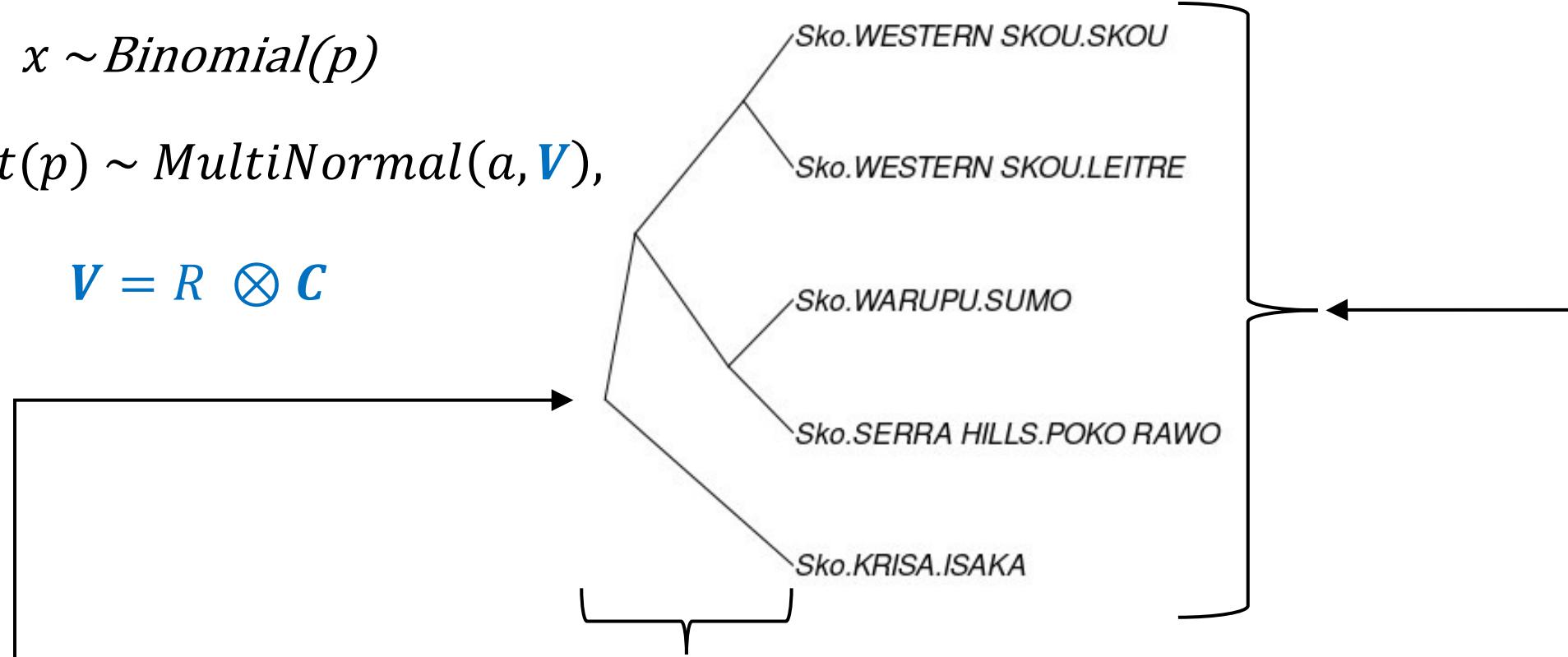
$$\text{inv_logit}(p) \sim \text{MultiNormal}(a, \mathbf{V}),$$

$$\mathbf{V} = \mathbf{R} \otimes \mathbf{C}$$

means a
 /root values

Variance-Covariance matrix \mathbf{C}
 /shared history

probabilities p
 /trait values x



Logistic Brownian Motion

$$x \sim \text{Binomial}(p)$$

$$\text{inv_logit}(p) \sim \text{MultiNormal}(a, \mathbf{V}),$$

$$\mathbf{V} = \mathbf{R} \otimes \mathbf{C}$$

$$\mathbf{R} = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Evolutionary rate

Logistic Brownian Motion

$$x \sim Binomial(p)$$

$$inv_logit(p) \sim MultiNormal(a, \mathbf{V}),$$

$$\mathbf{V} = \mathbf{R} \otimes \mathbf{C}$$

$$\mathbf{R} = \begin{bmatrix} \sigma^2_{11} & \boldsymbol{\sigma}_{21} \\ \boldsymbol{\sigma}_{12} & \sigma^2_{22} \end{bmatrix}$$

Trait correlation

Setup

1. Models for each single family:

Correlated:

$$R = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Independent:

$$R = \begin{bmatrix} \sigma^2_{11} & 0 \\ 0 & \sigma^2_{22} \end{bmatrix}$$

Setup

1. Models for each single family:

Correlated:

$$R = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Independent:

$$R = \begin{bmatrix} \sigma^2_{11} & 0 \\ 0 & \sigma^2_{22} \end{bmatrix}$$

2. Models across **all families**:

Lineage-specific correlation:

$$R_f = \begin{bmatrix} \sigma^2_{11f} & \sigma_{21f} \\ \sigma_{12f} & \sigma^2_{22f} \end{bmatrix},$$

for each family $f \in F$

Universal correlation:

$$R = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Universal independence:

$$R = \begin{bmatrix} \sigma^2_{11} & 0 \\ 0 & \sigma^2_{22} \end{bmatrix}$$

Setup

1. Models for each single family:

Correlated:

$$R = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Independent:

$$R = \begin{bmatrix} \sigma^2_{11} & 0 \\ 0 & \sigma^2_{22} \end{bmatrix}$$

2. Models across **all families**:

Lineage-specific correlation:

$$R_f = \begin{bmatrix} \sigma^2_{11_f} & \sigma_{21_f} \\ \sigma_{12_f} & \sigma^2_{22_f} \end{bmatrix},$$

for each family $f \in F$

Universal correlation:

$$R = \begin{bmatrix} \sigma^2_{11} & \sigma_{21} \\ \sigma_{12} & \sigma^2_{22} \end{bmatrix}$$

Universal independence:

$$R = \begin{bmatrix} \sigma^2_{11} & 0 \\ 0 & \sigma^2_{22} \end{bmatrix}$$

Results

- Single Family Models:
 - No trait pairs correlated consistently across language families
 - Observation consistent for Bayes Factors and Information Criteria (WAIC, LOOIC)

Results

- Single Family Models:
 - No trait pairs correlated consistently across language families
 - Observation consistent for Bayes Factors and Information Criteria (WAIC, LOOIC)
- Cross-Family Models:
 - Bayes Factors:
Lineage-specific correlations valued much higher than universal models
 - Information Criteria:
Contrarily, *universal models* valued much higher than lineage-specific

Conclusions

- Single-Family models and Bayes Factors for Universal models are in favour of **only lineage-specific** correlations
 - However, Information Criteria for Universal models are in favour of **universal** correlations
- ⇒ **No clear evidence in favour of any universal trait correlations**
- ⇒ **No clear evidence to support that cross-family models can capture universal correlations better**

References

- Matthew S. Dryer and Martin Haspelmath, editors. 2013. WALS Online. Max Planck Institute for Evolutionary Anthropology, Leipzig.
- Gerhard Jäger. 2018. A bayesian test of the lineage-specificity of word order correlations. In 12th International Conference on Language Evolution (Evolang XII), Torun