



Fingerprints, forever young?

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Fingerprint (Template) Ageing

Methods

Results

Conclusion



Fingerprint (Template) Ageing



A decrease in mated similarity score,
with increasing time interval
between reference and probe image.

children

- ▶ fingerprints grow uniformly into each direction

e.g.

- ▶ Gottschlich, Carsten; Hotz, Thomas; Lorenz, Robert; Bernhardt, Stefanie; Hantschel, Michael; Munk, Axel (2011). Modeling the Growth of Fingerprints Improves Matching for Adolescents. IEEE Transactions on Information Forensics and Security
- ▶ Haraksim, Rudolf; Galbally, Javier; Beslay, Laurent (2019). Fingerprint growth model for mitigating the ageing effect on children's fingerprints matching. Pattern Recognition
- ▶ Galbally, Javier; Haraksim, Rudolf; Beslay, Laurent (2018). A Study of Age and Ageing in Fingerprint Biometrics. IEEE Transactions on Information Forensics and Security

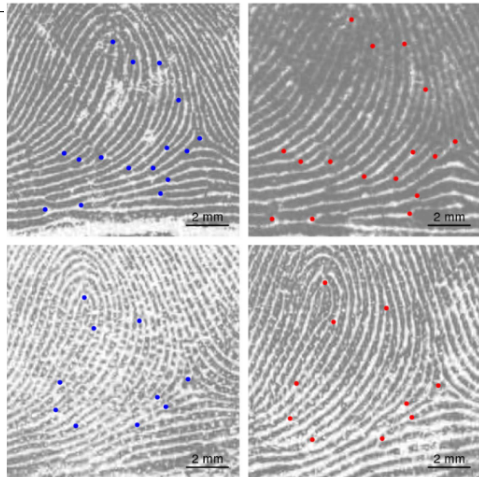


Figure adapted from: Gottschlich, T. Hotz, R. Lorenz, S. Bernhardt, M. Hantschel, and A. Munk, "Modeling the growth of fingerprints improves matching for adolescents"



adults

- ▶ mated similarity scores decrease with increasing time between samples
- ▶ decreasing image quality \rightarrow decreasing similarity score
- ▶ e.g.
 - ▶ Arnold, M.; Busch, C.; Ihmor, H. (2005). [IEEE Proceedings from the Sixth Annual IEEE Systems, Man and Cybernetics (SMC) Information Assurance Workshop, 2005
 - ▶ Yoon, Soweon, and Anil K. Jain. "Longitudinal study of fingerprint recognition." Proceedings of the National Academy of Sciences
 - ▶ Galbally, Javier; Haraksim, Rudolf; Beslay, Laurent (2018). A Study of Age and Ageing in Fingerprint Biometrics. IEEE Transactions on Information Forensics and Security
 - ▶ Kirchgasser, Simon; Uhl, Andreas (2017). [IEEE 2017 IEEE International Conference on Identity, Security and Behavior Analysis (ISBA) - New Delhi, India

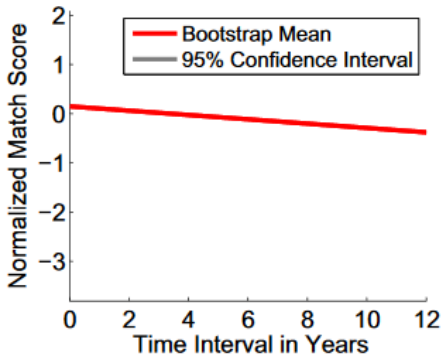


Figure adapted from: Yoon, Soweon, and Anil K. Jain. "Longitudinal study of fingerprint recognition." Proceedings of the National Academy of Sciences



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data & subjects

- ▶ fingerprint scanner
 - ▶ capacitive (UPEK TouchChip)
 - ▶ embedded in an access control framework
 - ▶ 508 ppi
 - ▶ 256×360 px
- ▶ data subjects
 - ▶ $n = 20$ (6 females)
 - ▶ 21 – 58 years (Md. 31) at enrolment
 - ▶ 1 – 4 finger instances per data subject (left/right index finger/thumb)
 - ▶ 3 – 1772 samples per finger
 - ▶ up to 12 years between samples



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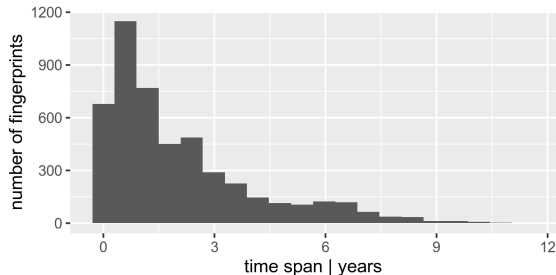




image processing

- ▶ FingerNet framework

(minutia extraction)

Tang, F. Gao, J. Feng, and Y. Liu, "FingerNet: A unified deep network for fingerprint minutiae extraction"

- ▶ Minutia Cylinder Code

(→ mated similarity scores)

Cappelli, M. Ferrara, and D. Maltoni, "Minutia cylinder-code: A new representation and matching technique for fingerprint recognition"

- ▶ NFIQ2.0

(→ fingerprint quality)

<https://www.nist.gov/publications/fingerprint-image-quality>

- ▶ filtering procedure

- ▶ only impressions with quality score > 10
- ▶ only impressions with > 16 Minutia
- ▶ keep only instances with a time interval > 1 year



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Hierarchical Linear Modeling (HLM)

- ▶ research question: *Can we find support for fingerprint template ageing?*
- ▶ operationalization: mated similarity scores should decrease with increasing time interval
- ▶ consider: fingerprint quality und sociodemographic factors
- ▶ modeling:

$$y_{ij} = \underbrace{\beta_0 + \beta_1 \cdot x_{ij} + \beta_2 \cdot q_{ij} + \alpha_i + \gamma_i}_{\text{fixed effects}} + \underbrace{b_{0i} + b_{1i} \cdot x_{ij} + b_{2i} \cdot q_{ij} + e_{ij}}_{\text{random effects}}$$

fixed effects \longleftrightarrow random effects

- ▶ model reduction

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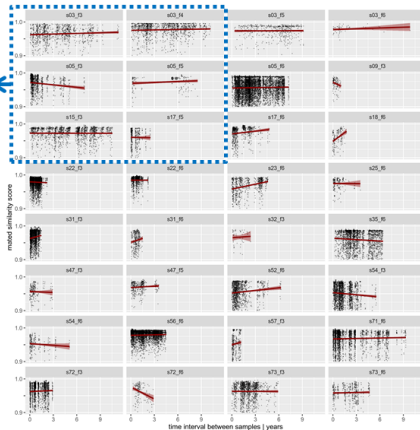
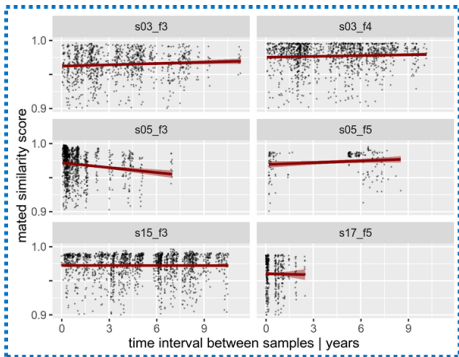
Model backwards regression: what do eliminated terms tell us about the data?

$$y_{ij} = \beta_0 + \beta_1 \cdot x_{ij} + \beta_2 \cdot q_{ij} + \alpha_i + \gamma_j + \underbrace{b_{0i} + b_{1i} \cdot x_{ij} + b_{2i} \cdot q_{ij} + e_{ij}}$$

Effects on **mated similarity score**:

- ▶ **global** (i.e. fixed) intercept is high (high similarity scores between samples acquired closely in time)
- ▶ no **global** effect of increasing **time interval**
- ▶ no **global** effect of **image quality**
- ▶ no **global** effect of **age and gender**
- ▶ random intercept: significant, subject-specific deviations in intercept
- ▶ **within-subject** (i.e. random) effect of increasing **time interval**
- ▶ **within-subject** effect of **image quality**

mated similarity scores behave differently within different finger instances



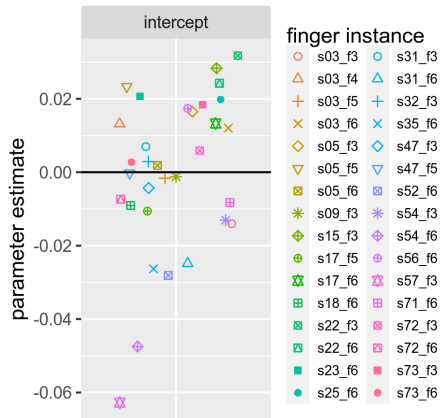
How much variability in the data is explained by each random effects term?

$$y_{ij} = \beta_0 + \beta_1 \cdot x_{ij} + \beta_2 \cdot q_{ij} + \alpha_i + \gamma_j + \underbrace{b_{0i} + b_{1i} \cdot x_{ij} + b_{2i} \cdot q_{ij}} + e_{ij}$$

Effects on **mated similarity score**:

- ▶ random intercept: 38%
- ▶ **time interval**: 0.5%
- ▶ **image quality**: 0.05%
- ▶ random error: 61%

subjects score differently well





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Summary of findings, limitations & possible implications

- ▶ no general **fingerprint template ageing**
- ▶ no general effect of image quality
- ▶ high inter-individual variability of mated similarity scores



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Yes - at least during long periods of professional life!

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