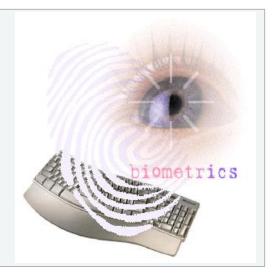


# **Continuous User Verification through Behavior Biometrics**

Dipl.-Inform Arik Messerman













#### 1. Introduction, Motivation

- 2. Focus
- 3. Deeper view to a Keystroke Dynamic Approach
- 4. Milestones, Discussion

## Introduction to Biometrics



Biometrics

#### Physiological Biometric

- Passive approach
  - Measure distinct traits that humans have
  - Do not vary over time
- Iris scans, retina scans, fingerprints, DNA, …

#### Behavioral Biometric

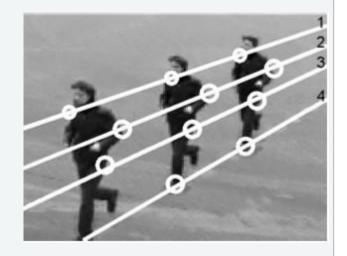
- Active approach
  - Measure performed tasks
  - Do vary over time
- Types of behavioral Biometrics
- Each subdivision has its own characteristics in terms of
  - usage, deploy ability, user acceptance, quality, ...

# Motivation (Behavioral Biometrics)



Humans can be verified traditionally by / through ...

- Knowledge (passwords, PINs, ..)
- Ownership (software/security token, ID card, ...)
- Inherence (fingerprint, voice, interaction, ...)
  - In most cases: Physiological Biometrics
- Risk for traditional solutions:
  - Object is verified, object = actor
- Additional security layer is required
  →behavioral biometrics
  - Further requirements







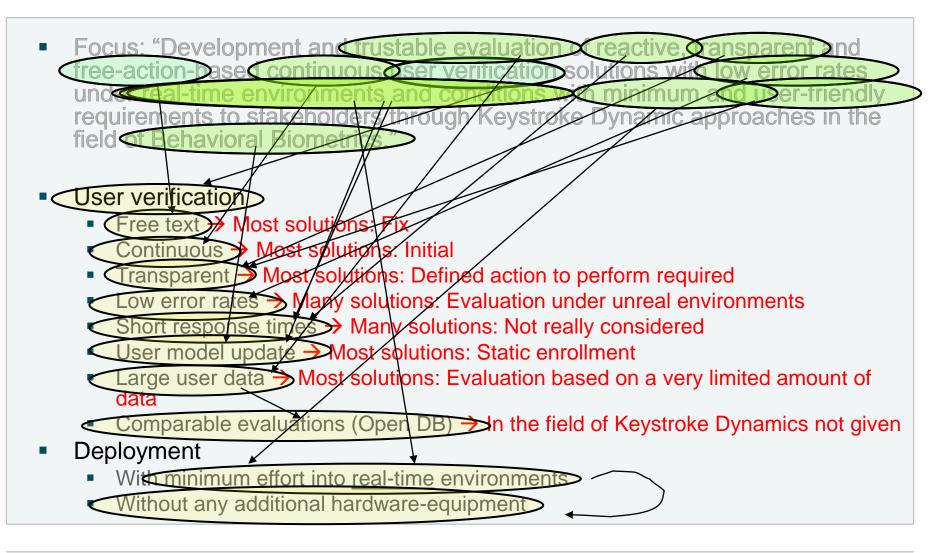
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#### Focus: Transparent Continuous Verification









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Database of user u:

ab	11	ff	се	ef	by	gk	gl	ew	kl	mn	op	ΟZ	qr	th	uv	WX	nt	УZ
10ms	11ms	14ms	15ms	16ms	17ms	19ms	20ms	20ms	23ms	24ms	27ms	27ms	28ms	30ms	33ms	35ms	36ms	41ms

New typing sample: "i will buy a new table ... open the door, ... efficient"



Database of user u:

ab	11	ff	се	ef	uy	gk	gl	ew	kl	mn	op	ΟZ	qr	th	uv	WX	nt	УZ
10ms	11ms	14ms	15ms	16ms	17ms	19ms	20ms	20ms	23ms	24ms	27ms	27ms	28ms	30ms	33ms	35ms	36ms	41ms

New typing sample: "i will buy a new table ... open the door, ... efficient"



Database of user u:

ab	11	ff	се	e	f <mark>al</mark>	o u	yl.	Lg	kf	Ēg	lbj	γe	wei	v k	:l <mark>o</mark> j	n q	ın <mark>t</mark> l	h c	pn	tс	Z	qr	th	uv	WX	nt	УZ
10ms	11ms	14ms	15ms	16	îh Sr	nls7	ih\$1	nls9	îh <b>s</b> r	1250	în Si	<b>1</b> 25	21S1	1253	দ্রীস্টা	12 <del>3</del> .	<u>_</u> B191	1257	'b161	<i>ඩ් පි</i>	ms	28ms	30ms	33ms	35ms	36ms	41ms

New typing sample: "i will buy a new table ... open the door, ... efficient"

ab	11	ff	th	uy	ew	op	nt
14ms	16ms	18ms	19ms	21ms	28ms	31ms	38ms

## **Distance calculation\***



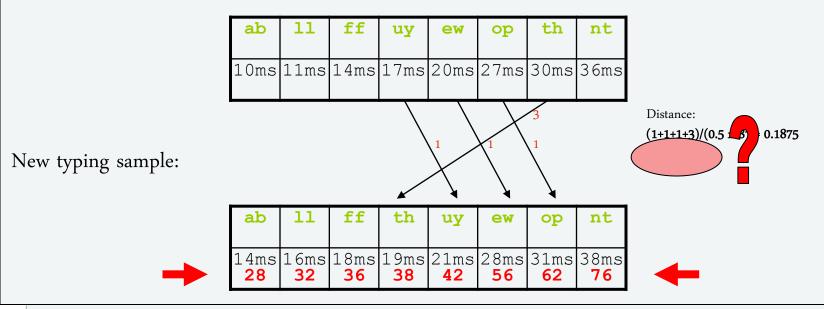
Database of user u:



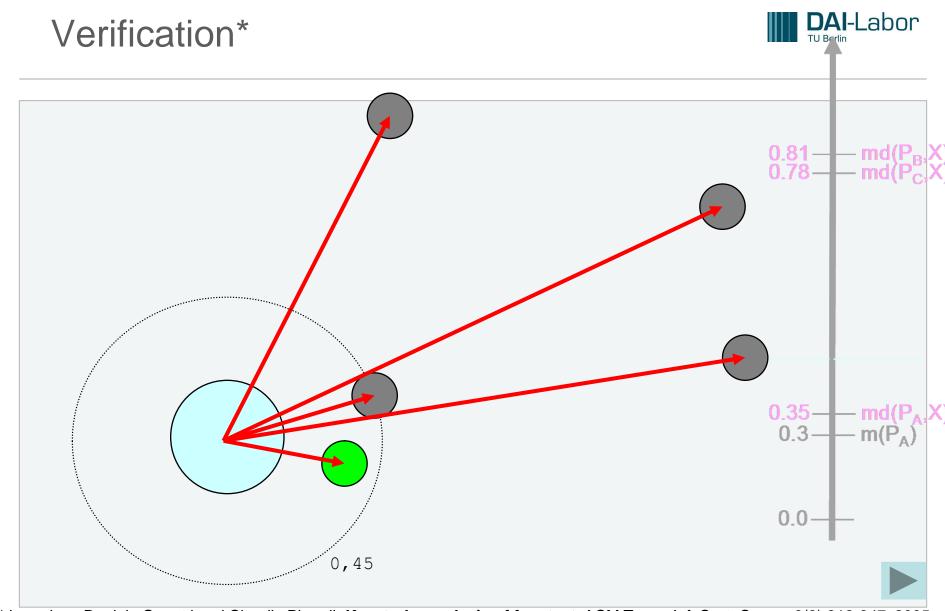
## **Distance calculation\***



Database of user u:

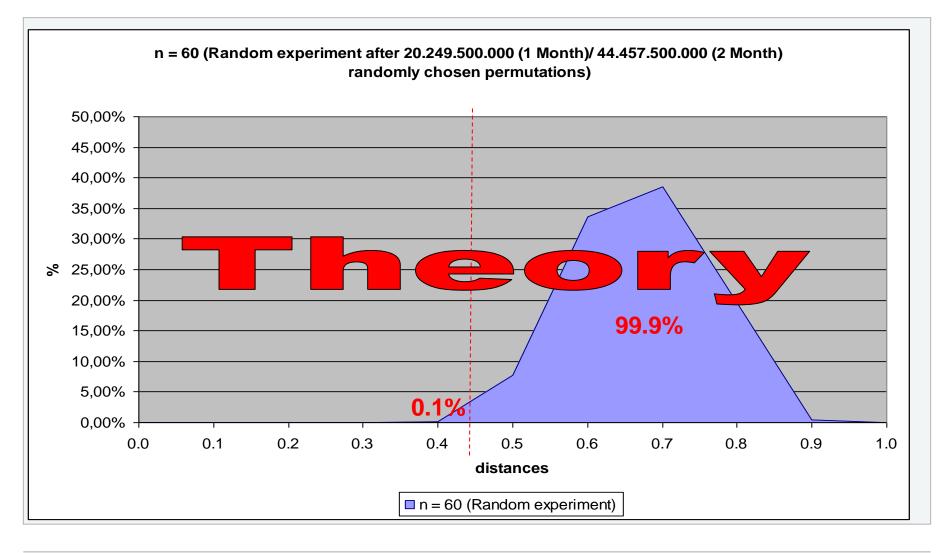


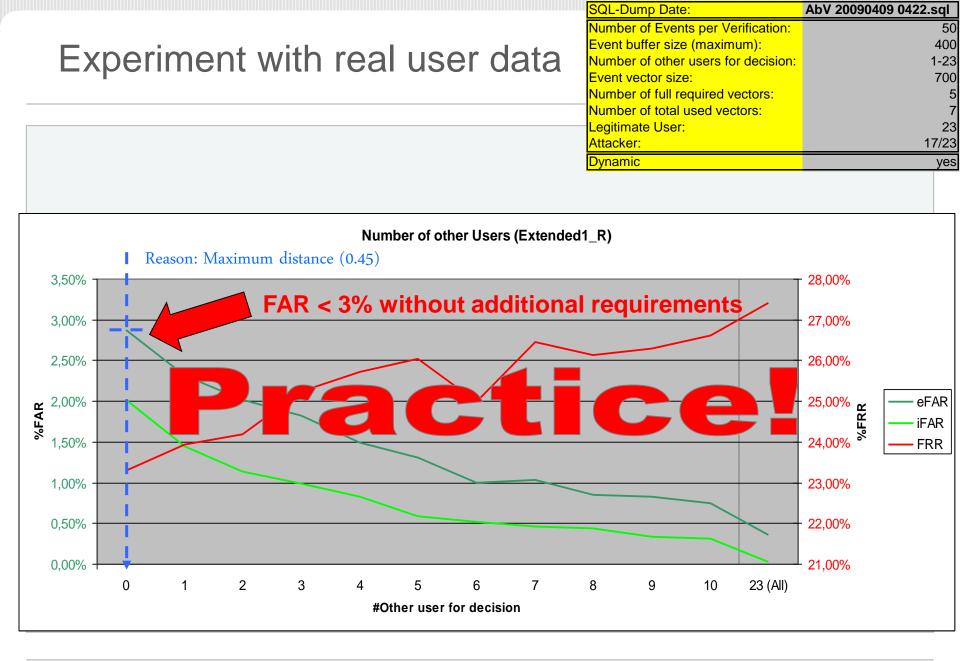
- Pattern of a user can be regarded as an <u>array</u> with values
- Calculation of the <u>distance</u> between patterns from the user data base and new one is to reduce to the calculation of the position of elements in permutations \* based on: Daniele Gunetti and Claudia Picardi. **Keystroke analysis of free text.** ACM Trans. Inf. Syst. Secur., 8(3):312-347, 2005.





 $n = 60 \rightarrow n! = 60! = 8,321 x$ 10<sup>81</sup> (after 2 Month on 8 CPU)





## **Bioinformatics approach**



- Distance calculation → Edit distance
- Evolution-Theory:
  - Combinations of amino acids are specified through sequences of nucleotides in DNA → Genes
    - Edit distances between DNA, RNA or protein strings
    - Protein: Sequence of units = amino acids
  - Example: glyceraldehyd3-phosphate dehydrogenase (GADPH) protein
    - Fly: GAKKVIISAPSAD-APM-F
    - Human: GAKRVIISAPSAD-APM-F
    - Yeast: GAKKVVSTAPSS-TPM-F
  - How closely related are two strings which represent the amino acid sequence of a particular gene between two species?
  - From a computer science perspective this issue is one of pattern matching and search

## **Bioinformatics approach**



From a computer science perspective this issue is one of pattern matching and search

- Idea:
  - Apply the <u>huge amount</u> practical and theoretical research that have been successfully developed in bioinformatics to the task of authentication/verification [/Anomaly detection]

## Other distance metrics:

- Levenshtein distance:
  - Levenshtein distance between two strings is given by the minimum number of <u>operations</u> needed to <u>transform</u> one string into the other, where an operation is an insertion, deletion, or substitution (???) of a single character
- Hamming, Euclidian, Cayley, Ulam, Spearman's Footrule, Spearman's rank correlation. Kondall's tau
- Spearman's rank correlation. Kendall's tau
  From a computer science perspective this issue is one of pattern matching and search





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#### Parallel work in project "Activity-based Verification" (until 06.2010) Work packages for DAI reflect next steps of my dissertation

Milestones (1)

- e.g. new distance metrics/approaches
- IEEE ISI 2009: "Identity Theft, Computers and Behavioral Biometrics" (Jun 09), Dallas, TX
- Survey journal: Draft version (45 pages)
  - State of the art, deeper discussions, review and novel views in the field of B.B.
- Verification & Evaluation Service Platform paper: Pre-draft version
  - Conception and Implementation of a generic platform was made

#### Several verification approaches

Focus: Continuous Free Text Verification



## Milestones (2)



- Web mail application
  - Enable transparent collection of behavioral data
  - Large dataset of user behavior
  - Currently: 52 users, ~5000 'KeyDown' events, Goal: ~100 users
- Theoretical/Scientific work paper: Paper planned (Start Sept. 09)
- Smart Senior: "Erkennung von Notsituationen im häuslichen Umfeld durch sensorbasierte Analyse von Verhaltensanomalien" Paper planend (Start Sept. 09)
  - Adapt knowledge made in the field of Behavioral Biometrics to anomaly detection

#### Bachelor/Diploma thesis

- 1. Adoption of existing (own) methods to login verification (Start 08.2009)
- 2. Generic evaluation engine of AbV verification methods (Start 07.2009)
- 3. Adoption of existing (own) methods to Smart Phone environments (Start 08.2009)

## Some References



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