

Monsters of the DiD

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2/3/26



Identity friction

UX – not a number

Re-Decentralised

“We live in two worlds...

the world into which we were born, and the otherworld that was born within us.

Both may be a blessing or a curse. We choose.” – Druid saying.

NO2ID
Stop the Database State



Very Big Id Systems of the World



मेरा आधार
मेरी पहचान



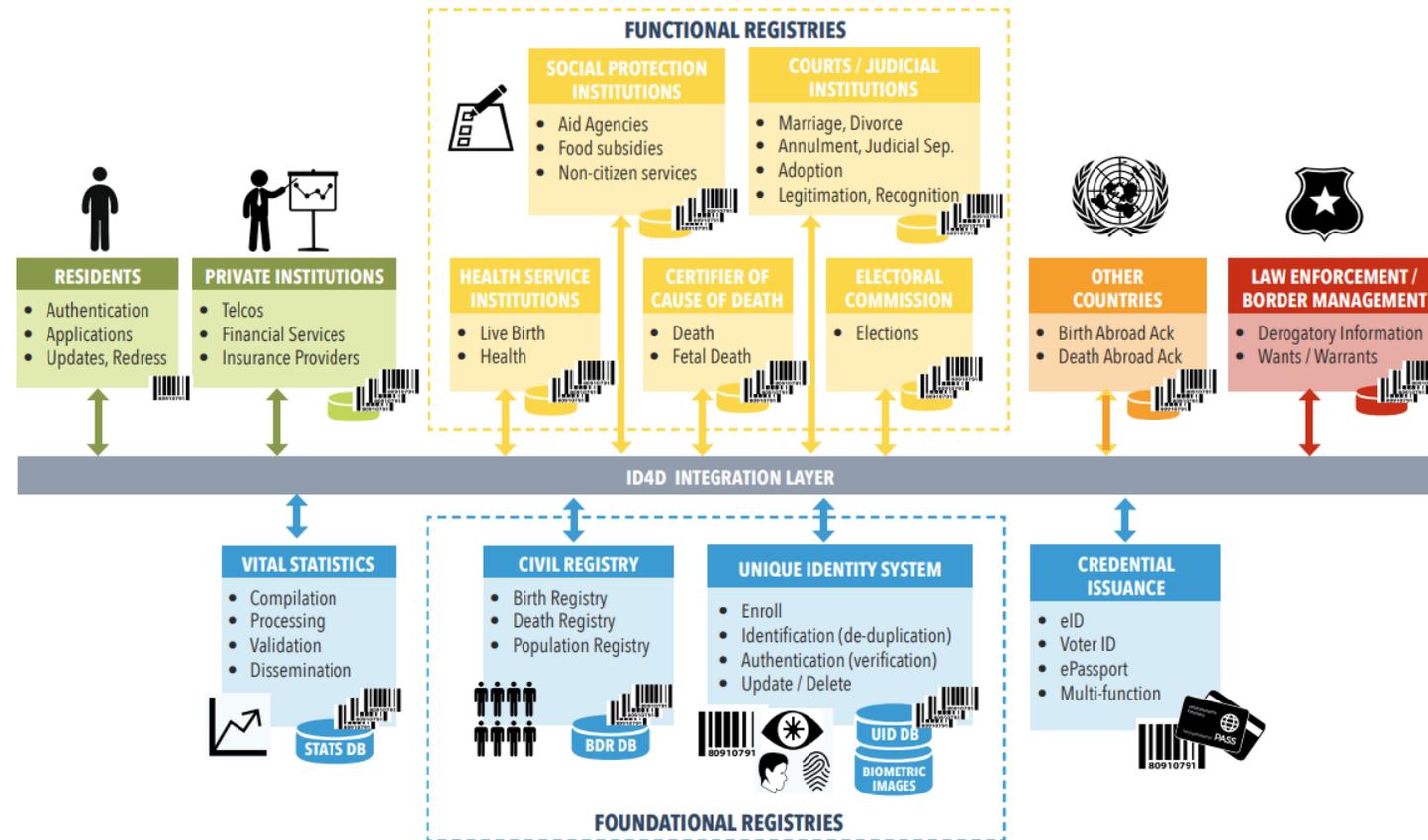
What makes you you? What is identity

- = or == ?
 - The Ship of Theseus
 - Trigger's Broom
- Heraclitus – objective context
 - You can't step in the same river twice
- Locke – cognitive context
 - Memories (solipsistic)
- **You** are who **they** say you are – personal context
 - E.g. you are always your parents child (etc) (at least til you take over care/attorney!)
 - Petal Kelly
- Crises
 - Cultural context
- Economicrisis
 - metrics
- A2D
 - Analog – Mind/Body (Sajjani) ->
 - How does being digital change anything?
 - Mind/Body/Bits/Body/Mind journey

Identity & Self

“The whole of this doctrine leads us to a conclusion, which is of great importance in the present affair, viz. that all the nice and subtile questions concerning personal identity can never possibly be decided, and are to be regarded rather as grammatical than as philosophical difficulties. Identity depends on the relations of ideas; and these relations produce identity, by means of that easy transition they occasion. But as the relations, and the easiness of the transition may diminish by insensible degrees, we have no just standard, by which we can decide any dispute concerning the time, when they acquire or lose a title to the name of identity.” David Hume

DiD Basics: Functional vs Foundational



The Unique ID system and Civil Registry are integrated through the ID4D Integration layer. This architecture preserves the modular nature of the model versus customized direct interactions.

Uniqueness – Is that needed?

	Uniqueness Matters	Uniqueness is Less Critical
	Foundational ID systems (e.g. national ID, civil registry)	Loyalty programs / memberships
Functional ID Systems	Voter registration & elections	Transportation cards / event passes
	Social protection/welfare programs	E-commerce or social media accounts
	Subsidy distribution (e.g. food, fuel, cash transfers)	Education records (basic)
	Immigration/border control	Digital login/auth systems (e.g. OAuth, SSO)
	Pension/employment tracking	Short-term or anonymous surveys/census
	Criminal justice	<i>“Uniqueness of identity is essential when a system confers rights, benefits, or responsibilities to individuals — where one person must not claim entitlements meant for another.”</i>
	Refugee registration / humanitarian aid	
	National health insurance systems	

Non Biometric Models

– For Avoiding Double Spend (so often about money!)

Mechanism	Description
Cryptographic Tokens	Unique, signed tokens (e.g. QR, NFC) issued per user or claim, validated on use
Verifiable Credentials (VCs)	Issued by trusted parties; users prove eligibility or uniqueness cryptographically
Zero-Knowledge Proofs (ZKPs)	Users prove they haven't claimed before, without revealing identity
Fuzzy Demographic Deduplication	Matching individuals based on partial or noisy personal data (name, DOB, etc.)

Pros and Cons

Mechanism	Pros	Cons
Cryptographic Tokens	<ul style="list-style-type: none"> - Easy to deploy offline - Low-cost - Strong one-time use control 	<ul style="list-style-type: none"> - Requires secure issuance and tracking - Token loss = benefit loss
Verifiable Credentials	<ul style="list-style-type: none"> - Privacy-preserving - Reusable & portable - Enables selective disclosure 	<ul style="list-style-type: none"> - Requires digital wallets & trusted issuers - Not trivial to deploy at scale
Zero-Knowledge Proofs	<ul style="list-style-type: none"> - Maximum privacy - Strong uniqueness without identity disclosure 	<ul style="list-style-type: none"> - Technically complex - Harder to implement in low-tech environments
Fuzzy Demographic Matching	<ul style="list-style-type: none"> - No biometric or advanced tech needed - Scalable for existing databases 	<ul style="list-style-type: none"> - Prone to errors (false matches or misses) - Lower trust in low-data-quality contexts

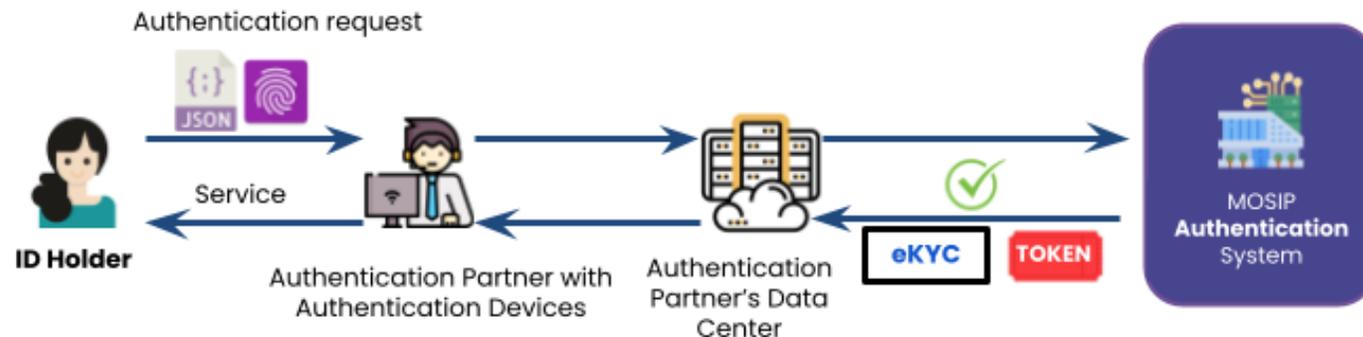
Identity Collision

- Capacity of biometrics crucial for deduplication.
- Analogous to “birthday problem” – how many people must be assembled before it is likely two share the same birthday.
- Theoretical capacity of IrisCode (245-bits, Daugman 2024) and false match rate of 10^{-20} enough to prevent identity collision of 12 billion persons.
- Biometric entropy of other modes is lower e.g. 82-bits for fingerprints (Young et al. IJCSSET 2013), and 40-bits for face ([Adler et al. 2006](#))

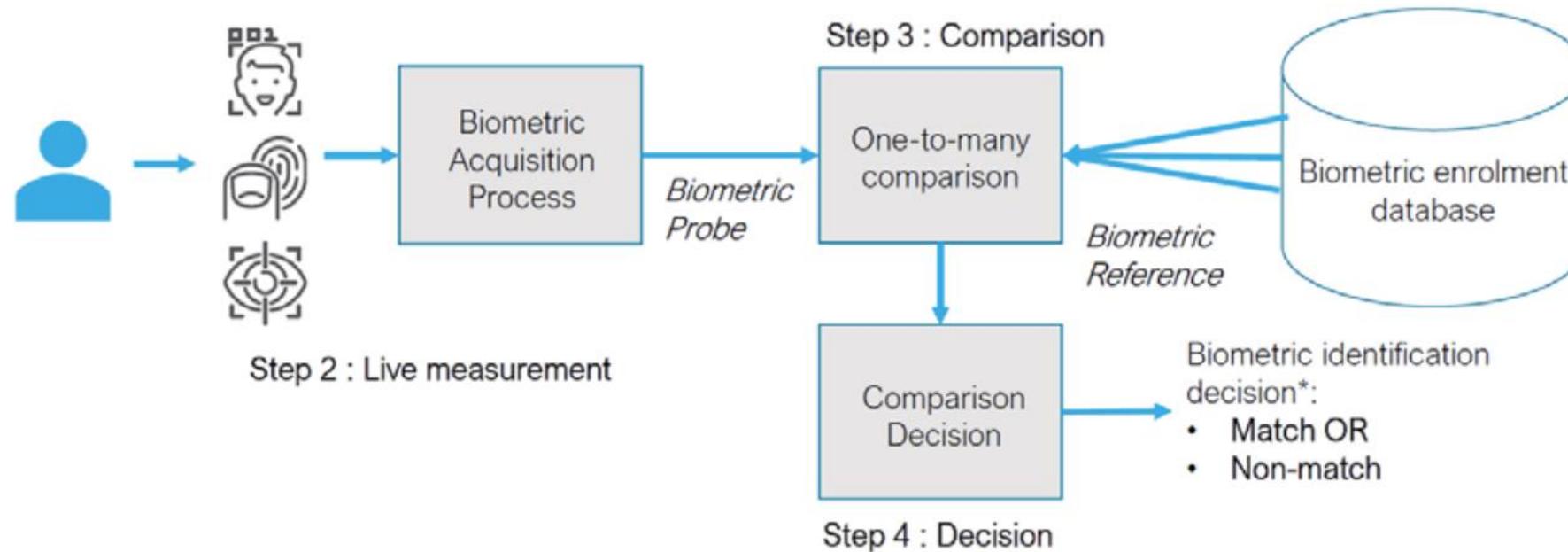


Figure 1: Representation of the IrisCodes [9], produced by four different eyes. The eight rows within each can be regarded as eight concentric rings, each encoding a $[0, 2\pi]$ traversal around the iris. (Eyelid masking is not shown.)

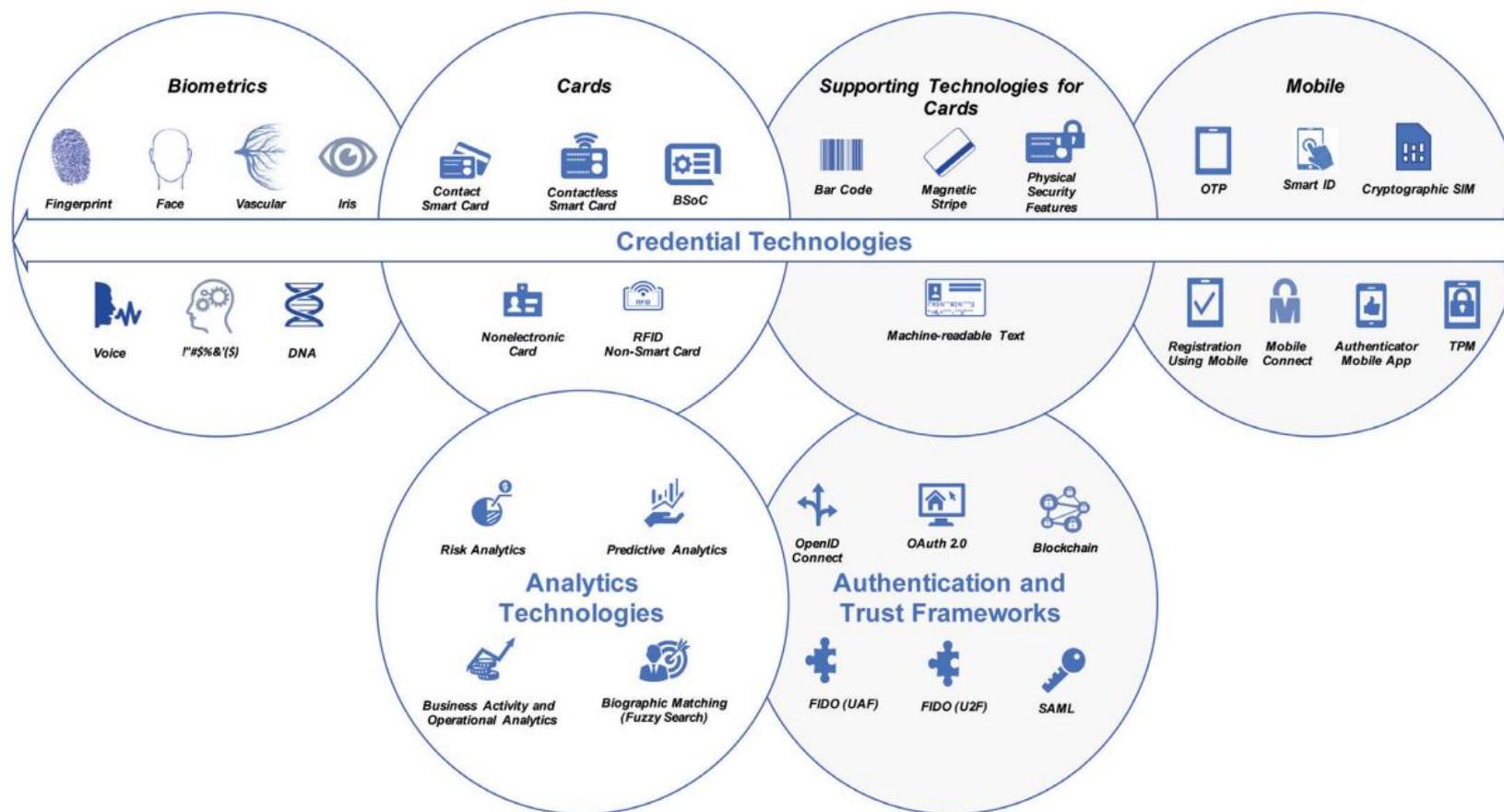
Identity Verification – typical workflow



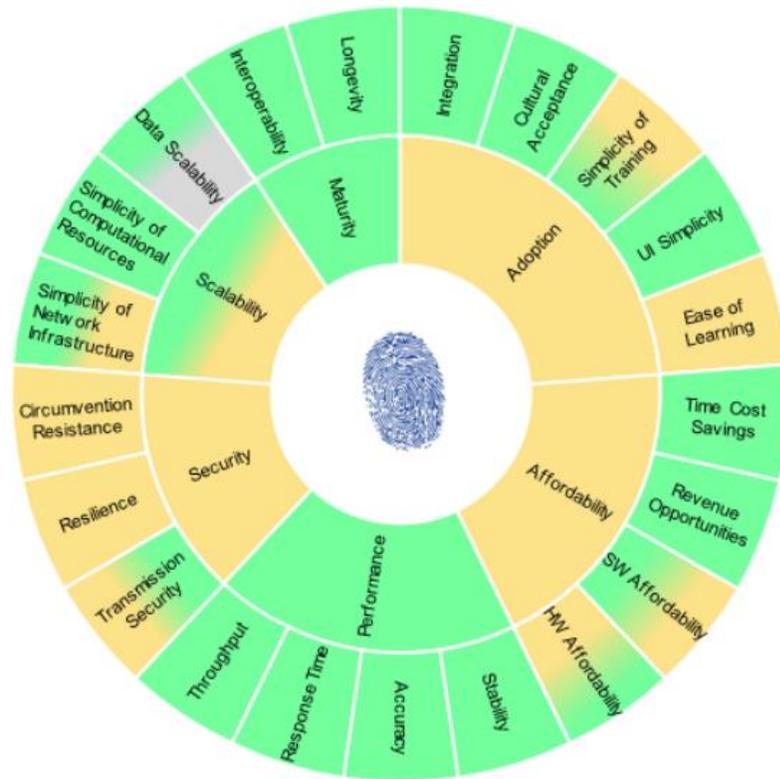
Deduplication/ Uniqueness



Technology Landscape

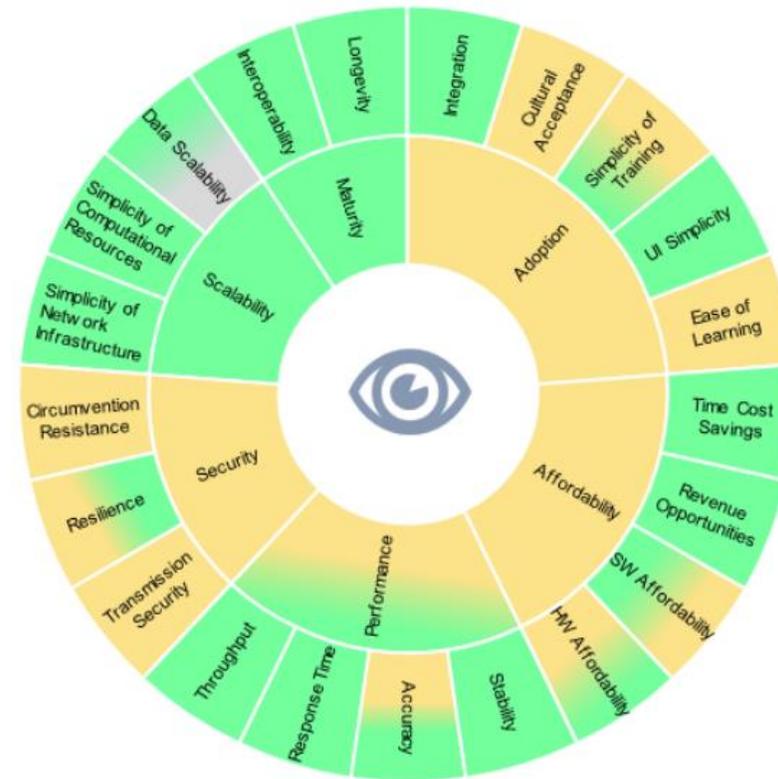


Fingerprint & Iris



High Medium Low N/A

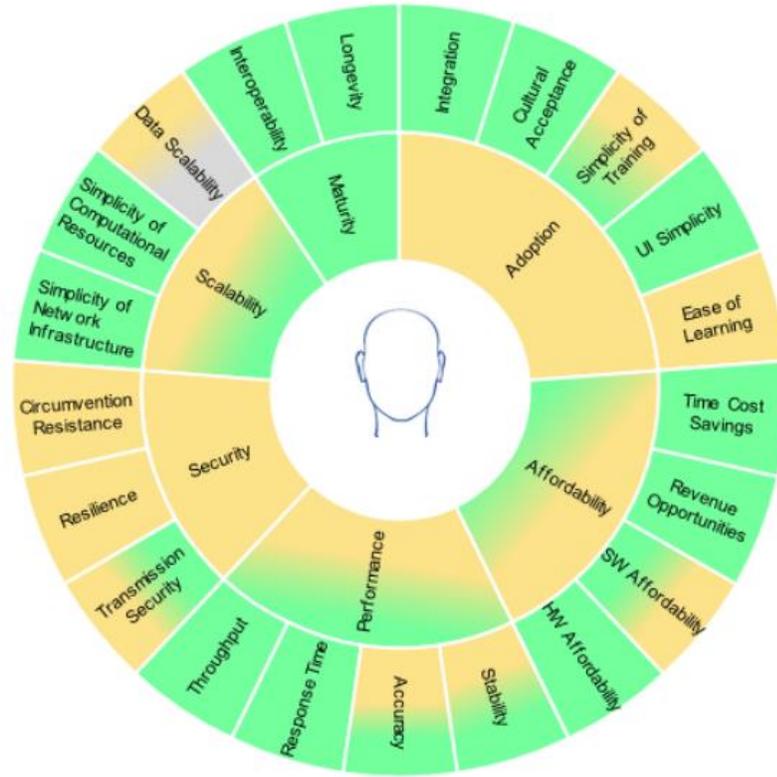
Fingerprint Capture and Matching



High Medium Low N/A

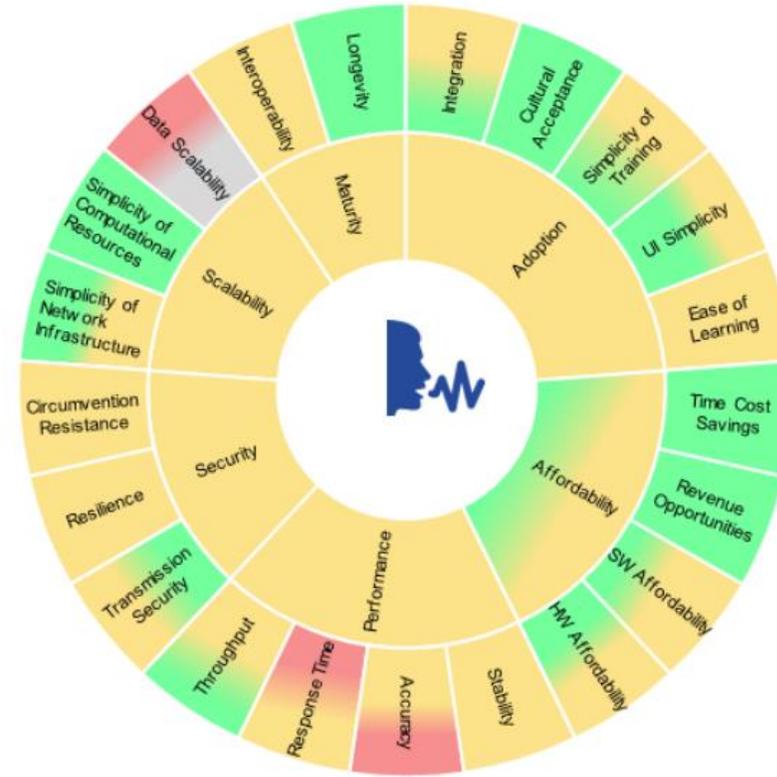
Iris Capture and Matching

Face & Voice



High Medium Low N/A

Face Capture and Matching



High Medium Low N/A

Voice Capture and Matching



Finger

Iris

Face

USE

Number available

1-10

1-2

1

Ease of capture

Easy to medium

Medium to hard

Easy

Adjudication

Medium—requires trained fingerprint examiner

Impossible with naked eye

Easy—any person can compare two faces

Accuracy for deduplication (1:N) assuming quality capture

Very high depending on number of fingers used and population size

Very high with 2 irises

Low to medium, but improving over time

COST

Capture device cost

1-print (US\$5-40), 2-print (US\$200-250), 10-print (US\$500-750)

US\$ 500-1000

Varies from cheap webcam-type devices to more expensive smartphones/tablets

Computing for duplicate enrollment check

Medium to high—more complicated algorithms require high-end computer cluster with large memory

Low to medium—iris matching algorithms are the most efficient as templates are stored in binary code

Medium to high—more complicated algorithms require high-end computer cluster with large memory

Tech Comparison



INCLUSION

Failure to capture (FTC)**Children**

<6 years: may not be viable

<1 year: may not be viable

All ages with updates needed over time (accuracy improves at older ages because the face stabilizes)

>6 years to adult: usable with software that accommodates for aging

1-5 years: challenging, requires parental assistance

Other groups with difficulties

Manual laborers, persons with disabilities, people with cuts on their fingers, people with diabetes

May be more invasive than fingerprints, stigma in some cultures; difficult for persons with visual impairments or albinism

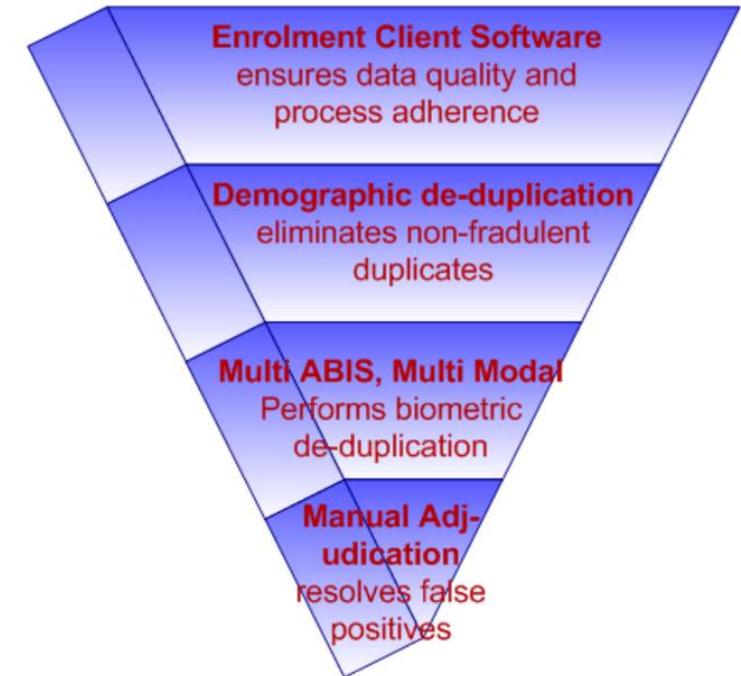
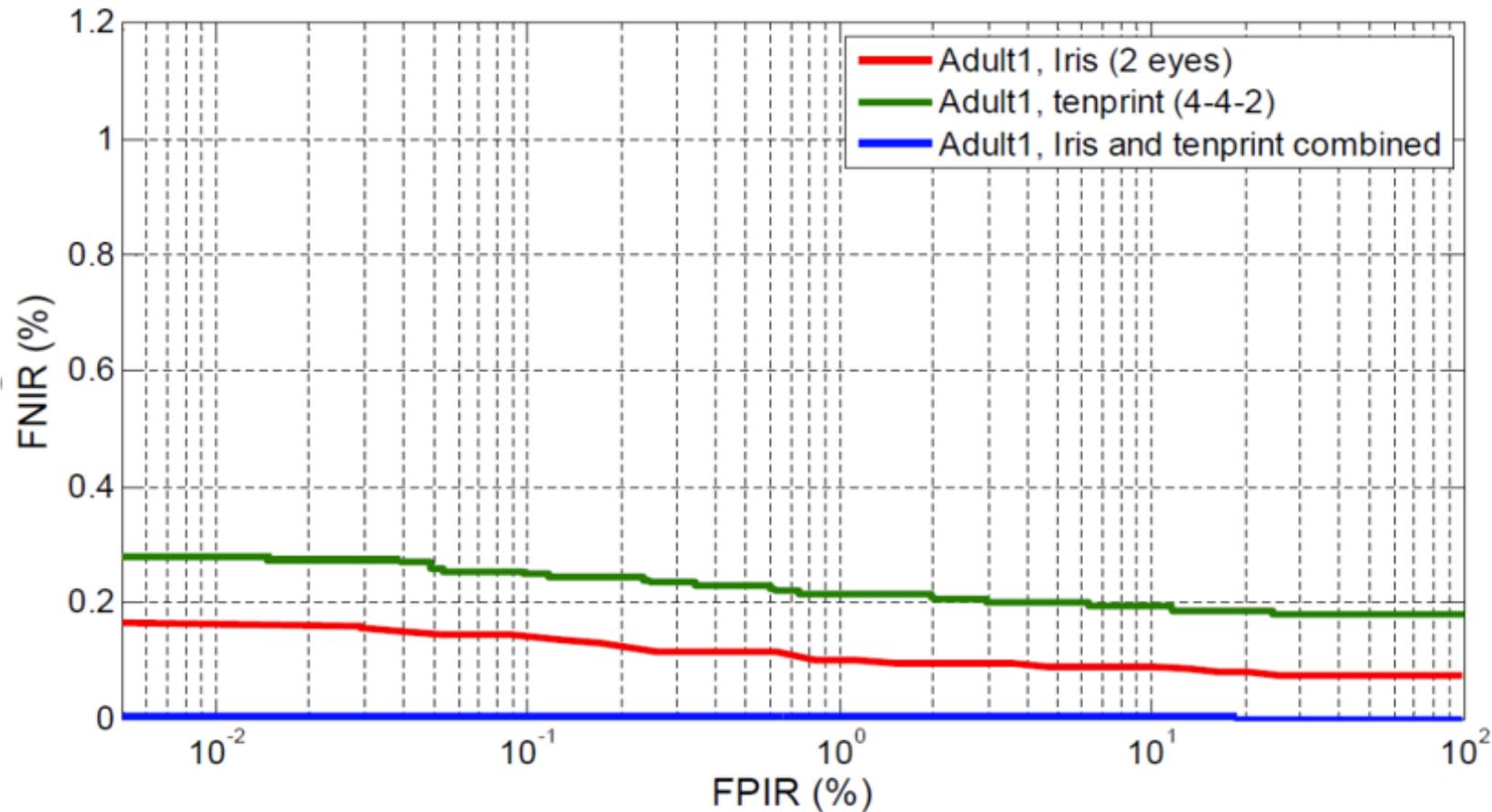
Not always optimized for recognition of darker skin tones, some algorithms have difficulty for persons with albinism

Single vs Multi Modal vs Population

Population Size	Recommended Modalities	Can One Modality (Face) Be Enough?	Evidence / Notes
Small* (<5 million)	Face or Fingerprint	Yes , if conditions are controlled	ID4D finds face-only viable in low-scale, low-risk cases when capture quality is high and manual review is available
Medium* (5–50 million)	Fingerprint + Face	Possible , but risk of false duplicates increases	ID4D classifies face deduplication as low-medium accuracy , with fingerprint adding stronger identity uniqueness.
Large* (50+ million)	Fingerprint + Iris + Face	No , face-only is not reliable	NIST FRVT shows false-positive identification rates (FPIR) for face grow nearly linearly with database size (FPIR \approx N \times FMR). Best algorithms still struggle with scale.

**Non standard definition*

Multimodal Deduplication Accuracy



*UIDAI POC for Enrolment (135000 records)
(2008, but still relevant)*

Accuracy Levels – NIST Evaluations

Mechanism	Modality	Accuracy
1:1 Comparison (Authentication) FAR* @ 0.001%	Fingerprint	TAR* = 99.56% (Verifinger V12.3)
	Iris	TAR = 99.43% (NIST IREX IX)
	Face	TAR = 99.83 % (NIST FRVT 2022)

1:N Comparison (Identification) FPIR* = 0.001	Fingerprint (10 Fingers)	FNIR* = 0.001 (5M Gallery)
	Fingerprint (1 Finger)	FNIR = 0.019 (100K Gallery)
	Iris (Both Eyes)	FNIR = 0.0035 (500K Gallery)
	Face	FNIR = 0.03 (12M Gallery)

*

FAR = False Acceptance Rate
TAR = True Acceptance Rate

**

FPIR = False Positive Identification Rate
FNIR = False Negative Identification Rate

- NIST FRVT 1:N Identification: <https://pages.nist.gov/frvt/html/frvt1N.html>;
- NIST FpVTE: <https://nvlpubs.nist.gov/nistpubs/ir/2014/NIST.IR.8034.pdf>
- NIST IREX 10 Identification Track: <https://pages.nist.gov/IREX10/>
- <https://biometrics.cse.msu.edu/Presentations/Israeli%20School%20on%20Biometrics%20April%2021-2025-FINAL.pdf>

Face – Why not for Large Population?

Particular	Details
Scaling Effect	According to NIST, in a 1:N search, the FPIR approximately equals $FMR \times N$. More manual deduplication requirement.
NIST FRVT evidence	On the 1.6 million-record Visa–Border dataset, NIST fixed FPIR at 0.3%. At that level, top algorithms showed FNIRs of ~0.16% to ~0.72%, missing 1 to 7 duplicates per 1,000.
Demographic biases	Rate varies by age, gender, race:

https://pages.nist.gov/frvt/reports/demographics/implications_for_1N.pdf
<https://www.paravision.ai/whitepaper-enterprise-grade-1n-face-recognition/>
<https://neurotechnology.com/awards-frvt-1-n.html>
<https://nvlpubs.nist.gov/nistpubs/ir/2019/nist.ir.8280.pdf>

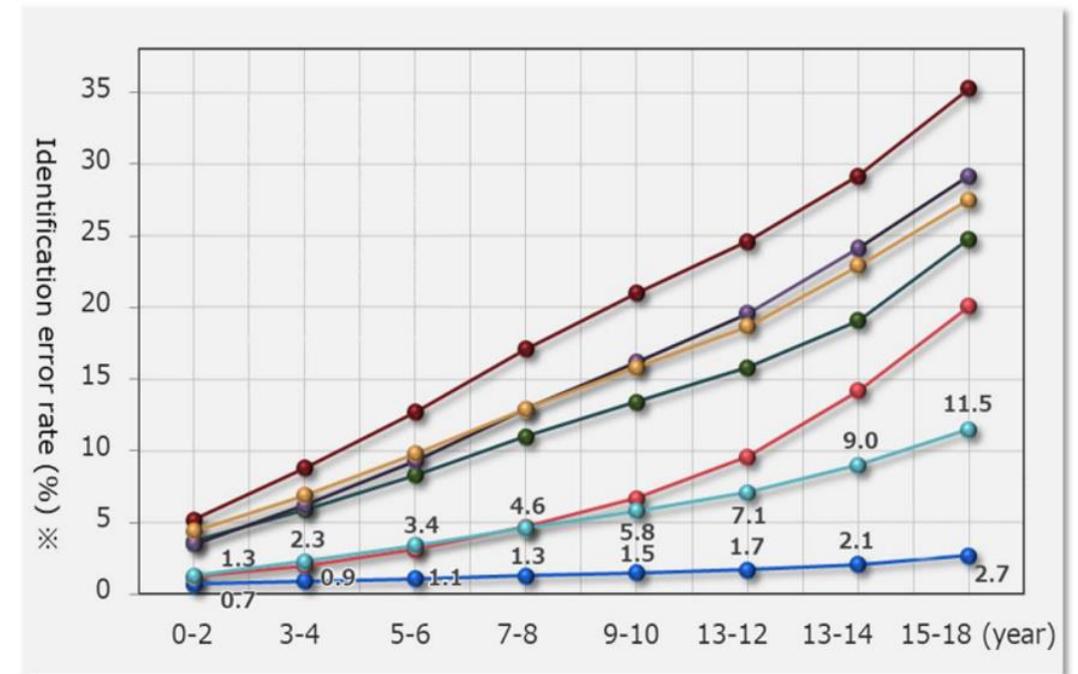
Demographic Bias & Age Impact

Demographic Bias

Race /Ethnicity	Sample Images	Verification Accuracy (%)
East Asian		93.72
Black		94.67
South Asian		93.98
Caucasian		96.18

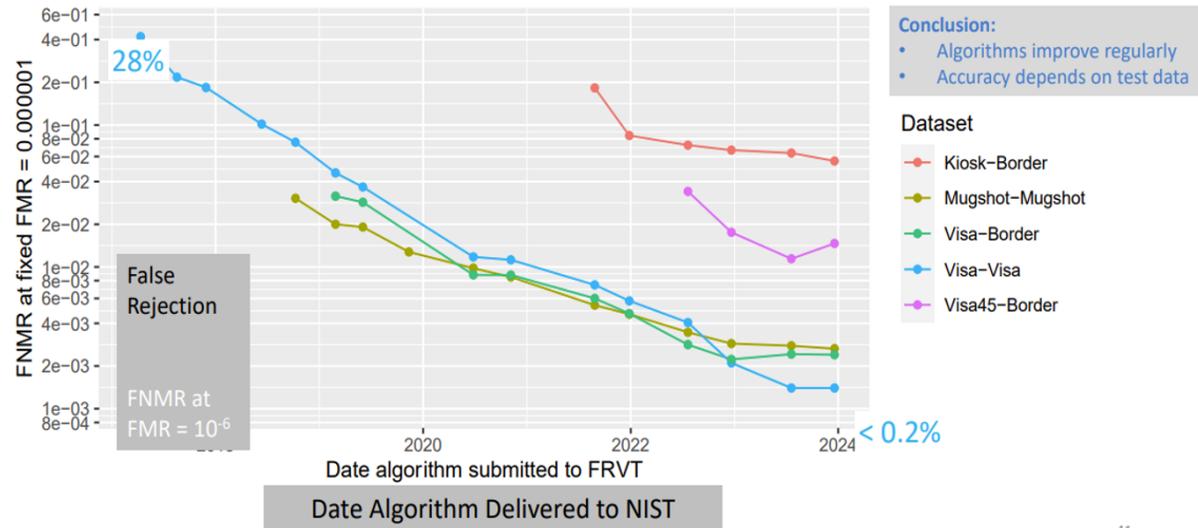
Face Verification performance by ArcFace [1] on each race/ethnicity cohort in RFW dataset [2]

Algorithm Accuracies over Aging Photos

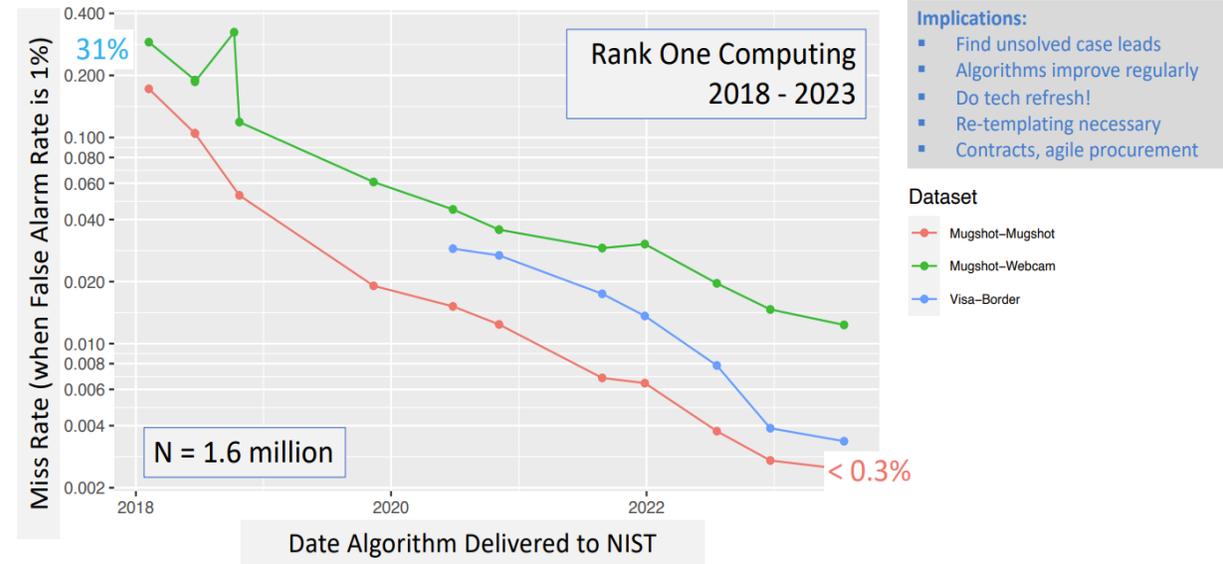


※ False-rejection discrimination rate at a false acceptance discrimination rate of 0.1% at the time of registration of 31000 people

Gain Continues for Face



1:1



1:N

Other Areas

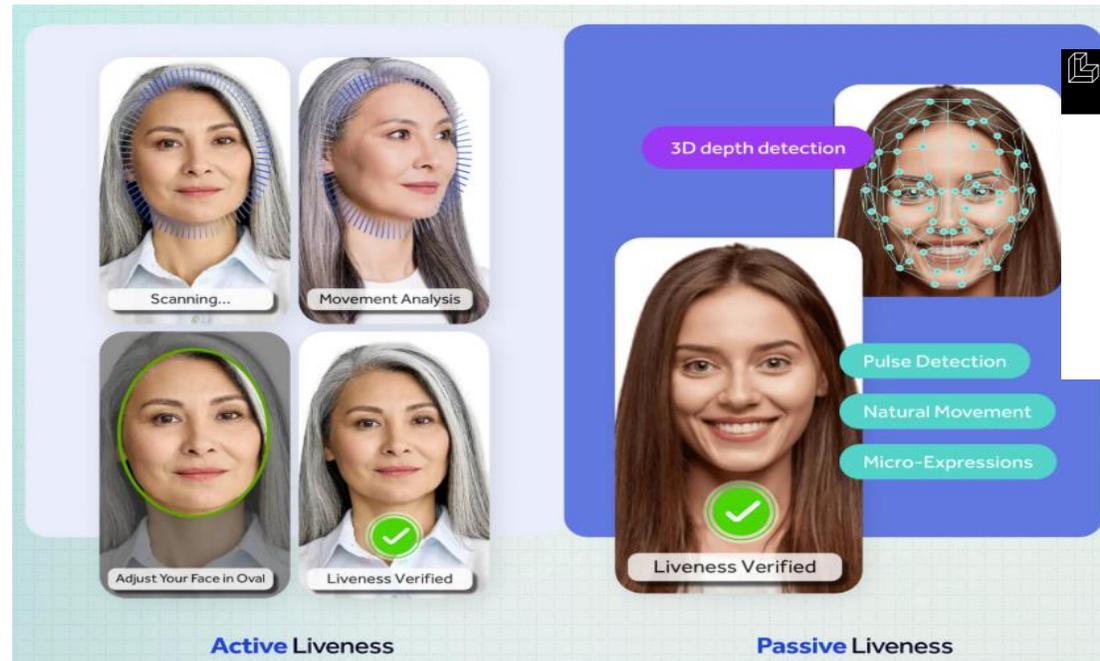
Liveness

Privacy

Consent



ISO/IEC 30107-1:2023
Information technology — Biometric presentation attack detection



Active Liveness

- Scanning...
- Movement Analysis
- Adjust Your Face in Oval
- Liveness Verified

Passive Liveness

- 3D depth detection
- Pulse Detection
- Natural Movement
- Micro-Expressions
- Liveness Verified

TECH & RIGHTS

7 Biggest Privacy Concerns Around Facial Recognition Technology

<https://facia.ai/blog/active-liveness-vs-passive-liveness-key-differences-and-how-they-work/>

<https://www.liberties.eu/en/stories/facial-recognition-privacy-concerns/44518>

Contactless Fingerprinting

Advantages	Challenges
<ul style="list-style-type: none"> • Hygienic (no contact) • Portable & potentially low-cost • Software-driven (mobile capture possible) <ul style="list-style-type: none"> • Smudge-free, less hardware • Larger capture area & multi finger possible 	<ul style="list-style-type: none"> • No global standards, vendor lock-in risks <ul style="list-style-type: none"> • Lower accuracy due to 3D-to-2D conversion artifacts • Poor interoperability with legacy ABIS • Capture issues: motion blur, lighting, background noise • Device variability (camera optics, OS) <ul style="list-style-type: none"> • Longer capture time, higher failure rates • Weak liveness detection vs. traditional sensors

Contactless Fingerprinting

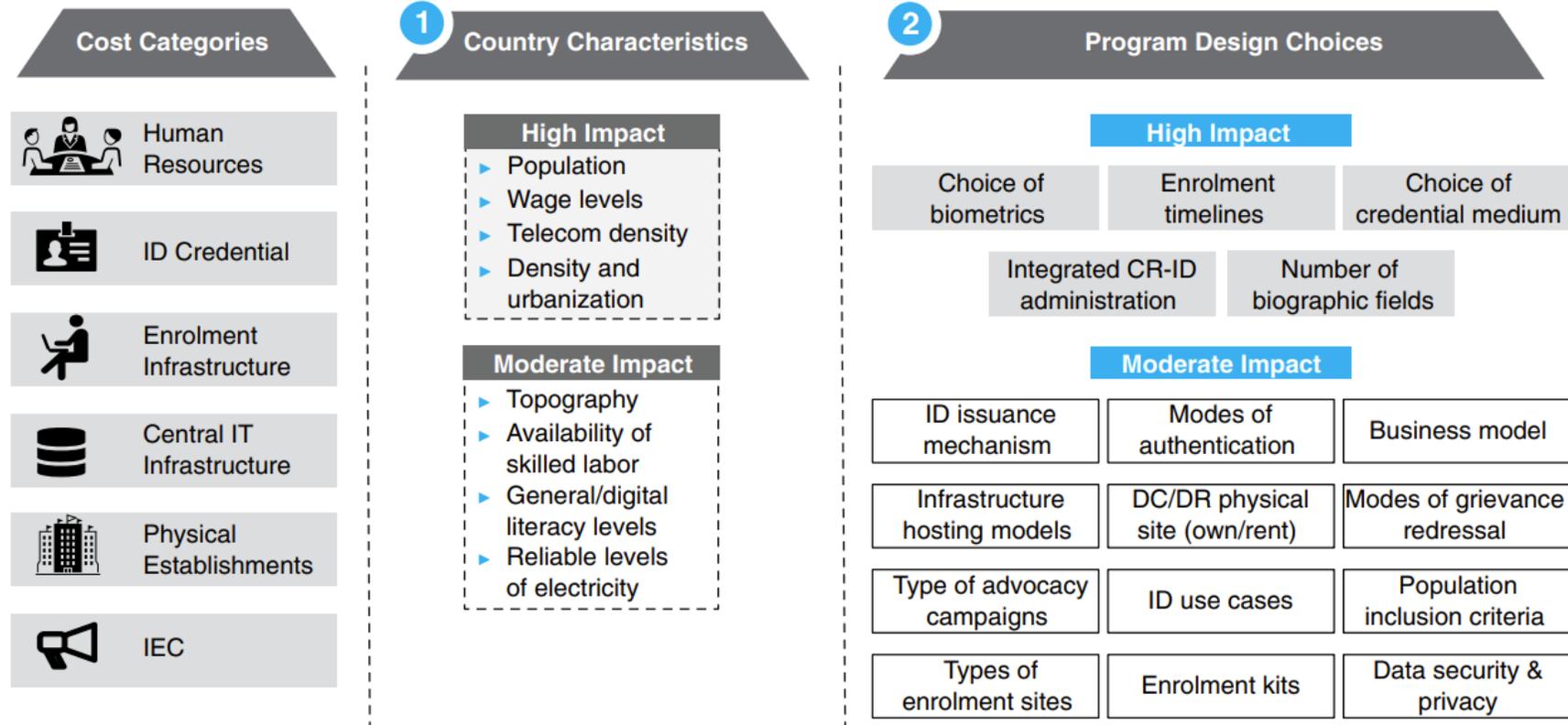
Key Evaluations:

- NIST (2020): Wide performance range (20–80%); only 1 hardware based solution >90% (multi-finger)
- NIST (2022–24): Warns against reusing contact-based quality algorithms; no cert like FBI Appendix F yet; legacy impact must be evaluated

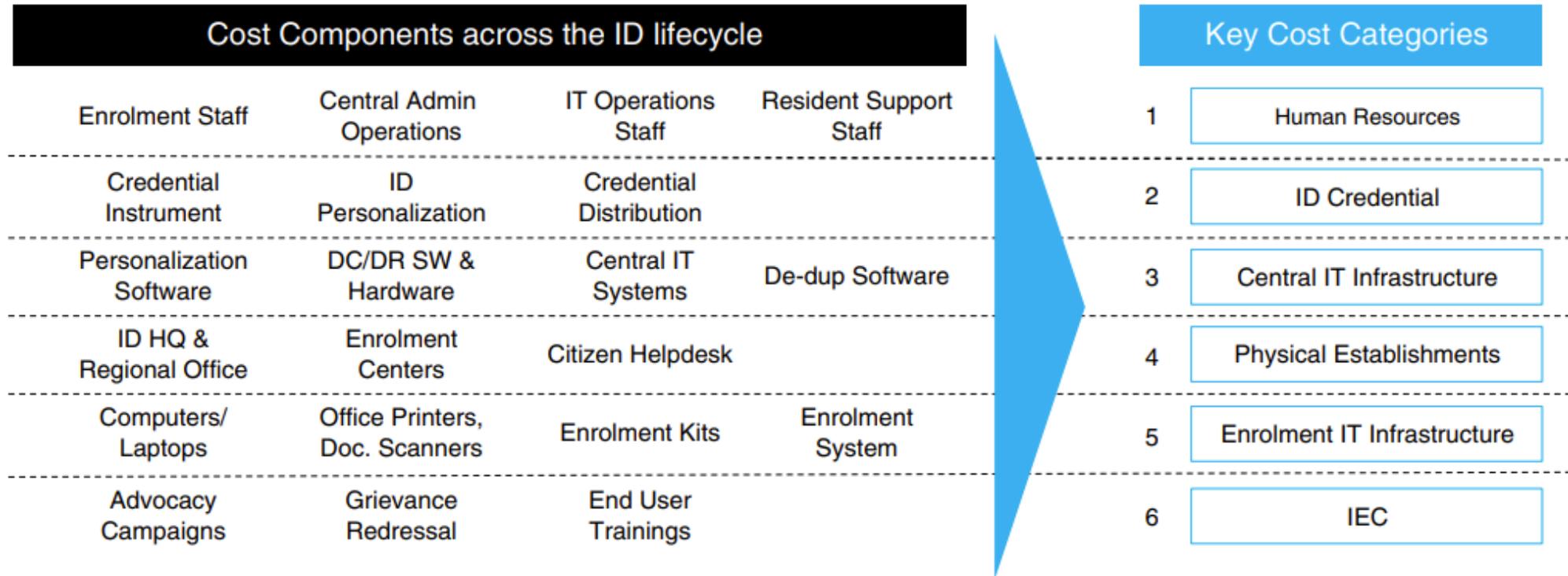
Conclusion:

While promising for self-service or low-assurance use cases, contactless fingerprinting is not yet reliable enough for deduplication or secure foundational ID enrollment without further standardization and validation.

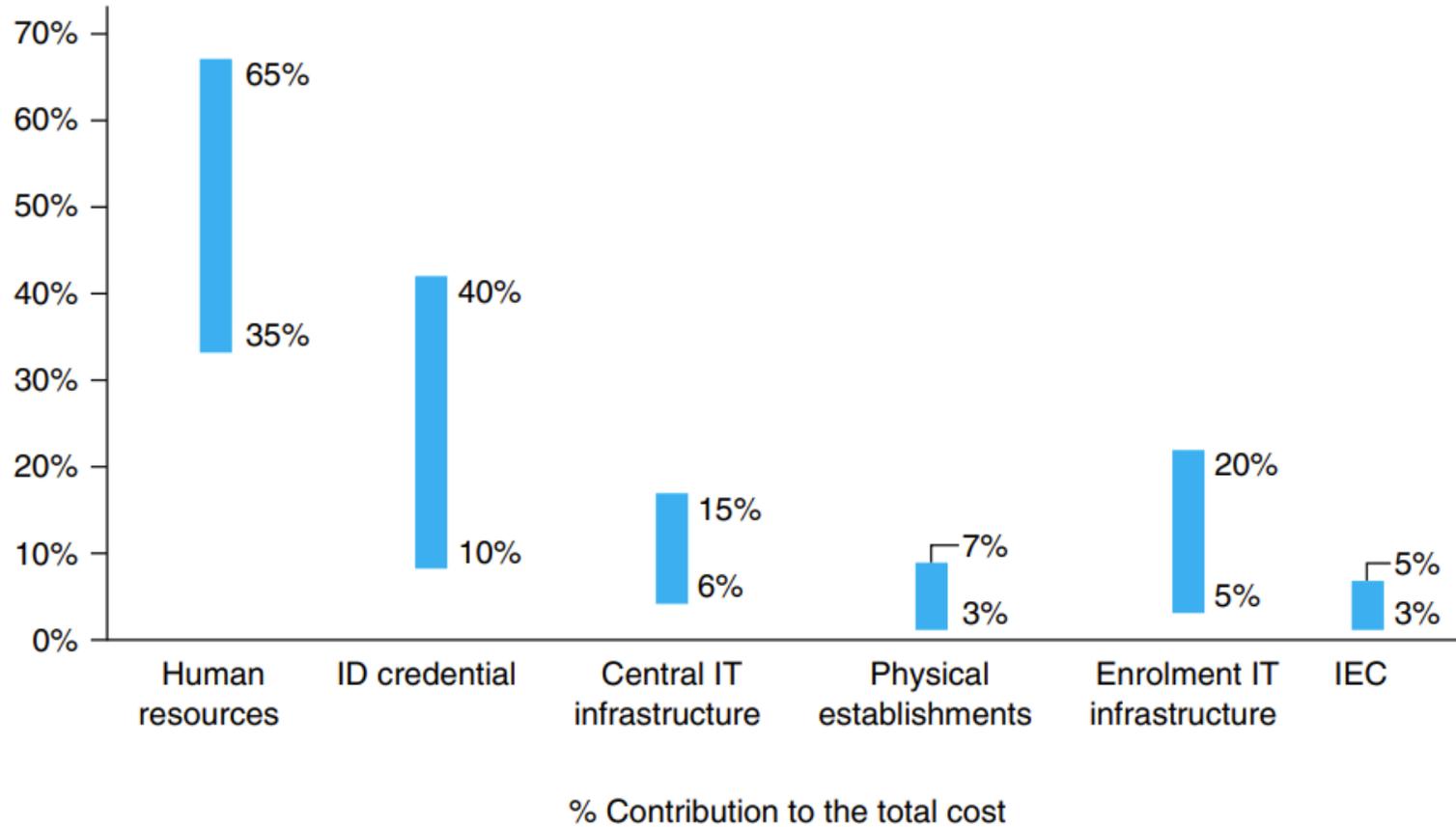
Cost Categories – ID System



Key Cost Categories



Cost Contribution



Cost – Additional Biometric Modality

“Each additional biometric modality increases accuracy and inclusion, and is estimated to increase enrolment costs by only about 5–10%.”

“Although multimodal biometrics may represent an added cost (compared with single mode biometrics, e.g., fingerprints), their use can—depending on population size and other characteristics—reduce overall costs because it will reduce the rate of manual adjudications during deduplication, as well as improve the accuracy and flexibility of authentication”

(study covers a group of 15 countries across Europe, South America, Africa, and Asia)

<https://documents1.worldbank.org/curated/en/702641544730830097/pdf/Understanding-Cost-Drivers-of-Identification-Systems.pdf> (2018)

Cost – Biometric Devices

Name of the item / Service	Unit cost (US \$)
Computer / Laptop	\$ 1,200.00
Mobile / tablet	\$ 200.00
Camera / webcam	\$ 100.00
Multifunction Printer	\$ 200.00
Power back up	\$ 100.00
Finger print scanner - One finger	\$ 200.00
Finger print scanner - Slab scanner	\$ 1,000.00
Iris scanner - one	\$ 300.00
Iris scanner - two	\$ 1,000.00
Signature pad	\$ 150.00
GPS dongle	\$ 100.00
Additional Screen	\$ 150.00
Voice recording device	\$ 150.00
Case for kit	\$ 200.00

Tri Modal Enrolment
Kit ~ 3K -4K USD

Depends on
Volume, Geography,
Businesss Risk etc.

Cost – Biometric Deduplication

Cost per deduplication (For first biometric)	\$	0.25
Add on dedup cost for additional biometric	\$	0.10



ID4D

Vendor 1

Records	Cost	Cost/ID
10K	20K	2
50K	35K	0.7
500K	100K	0.2
1M	170K	0.17
10M	600K	0.06
100M	3.4M	0.034

Vendor 3

Records	Cost/ID	
1M	1	With Hardware
5M	0.9	
10M	0.8	
1M	0.5	Only software
5M	0.45	
10M	0.4	

Matcher SDK could be 10 % of ABIS costing
10-12% AMC
40% for DR
30-40% AMC

Vendor 2

Records	Cost	Cost/ID
1 M	500K	0.5
5M	1.4 M	0.28
10M	1.6M	0.16

Vendor 4

Records	Cost	Cost/ID
1M	200K	0.2
5M	800K	0.16
10M	1.5M	0.15

Average

1 M = 0.35

5 M = 0.29

10 M = 0.19

<https://documents1.worldbank.org/curated/en/702641544730830097/pdf/Understanding-Cost-Drivers-of-Identification-Systems.pdf> (2018)

TriModal Data from Vendor Interviews; software cost by default

Identity friction – say no...or?

- UX – I am not a number
- versus
- Re-Decentralised Viz Estonia...



Recall John Perry Barlow

- “Governments of the Industrial World, you weary giants of flesh and steel, I come from Cyberspace, the new home of Mind. On behalf of the future, I ask you of the past to leave us alone. You are not welcome among us. You have no sovereignty where we gather.” ... + ...
- “We will create a civilization of the Mind in Cyberspace. May it be more humane and fair than the world your governments have made before.”

And Dave Clark

- “We reject: kings, presidents, and voting. We believe in: rough consensus and running code.”
- And Larry Lessig: “Code is Law” (in Code and Other Laws of Cyberspace)
- See also (1997)
<https://web.archive.org/web/20020420162518/http://www.telegraph.co.uk/connected/main.jhtml?xml=/connected/1997/04/22/ecdip22.xml>

What went wrong?

- Centralisation (rich get richer) +
 - Moore's Law + Metcalfe's "Law"
 - i.e. growth 100%/18months, and super-linear value proposition
 - Hyperscale (scale out) is a natural consequence of these net fx
- Imagine if Musk didn't just own Starlink (and supply via SpaceX)
 - But also, say, GPS. Or a nation's power grid?
- The forces of centralization are remarkably strong

Apple (hyperscalars) market cap >> UK GDP

Lots of other examples

AI startup valuations >> GDP of small nations

Yet energy consumption to have MVP also >> small nations

Few people (even US e.g. FBI v. Apple) can co-erce compliance

Impact on Nations

Back to Kings
(or at least
robber barons)

Maybe teach
diplomats to
code?

Or re-
decentralise?

Hybrid cyber-
physical
decentralised:

Estonia
decentralised
digital state, +

encrypted backup
to their embassies
in several other
countries

Now for some SF

What if....we could mod our biometrics?

- Link between bio- and digital no longer immutable (or unique)

