

Ocular Biometrics in a Mobile Phone

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The Need

-Historically unprecedented global market penetration of mobile devices across all sectors, plus:

-Phones are frequently lost, stolen or hacked

- 1.1B comprised accounts (2004-2013) & 80% of data breaches occur due to exploited or stolen credentials*
- \$221B (\$54B in US) identify related crimes in 2011*
- 96% of all lost phones were accessed by their finder*

-End users are frustrated with passwords, and it's getting worse

- The average person juggles at least 25 passwords*
- 38% of Americans would rather clean a toilet than think of a new password*
- 36% of Americans think world peace would be easier than remembering all their passwords*
- A recent report by Ericsson Labs found that 48% of consumers say they would like to replace passwords with an eye biometric in 2014

-Passwords and pins are easy for thieves and hackers to guess

- 1234, 1111, 0000 account for 25% of all pins in use*
- Stronger passwords don't help much and reduce usability

Biometric Solution

- Biometrics as the most natural and tentatively the best solution to this fundamental question: *Who is Holding the Phone?*
 - *It's something that you ARE, not something the you know or have (surrogate authentication)*
- Available modalities
 - Which ones make more sense?
 - *Hint: users constantly “touch” and “look at” their phones*
 - What are the issues facing mass consumer adoption?
 - User experience (ease, consistency across use case scenarios)
 - Price
 - Privacy, security, revocability
 - ...

Vascular Patterns as a Strong Biometric

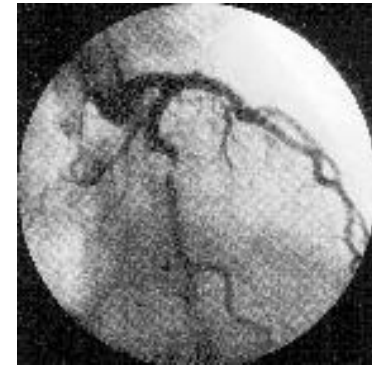
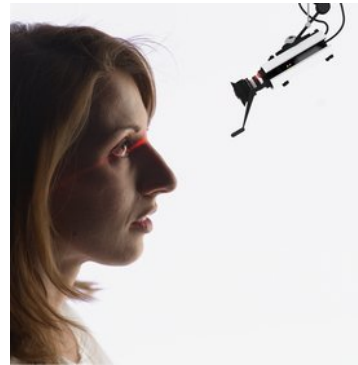
- Complex and unique patterns



Contactless palm vein recognition unit

← Hand and
finger vein

Retinal
vasculature→

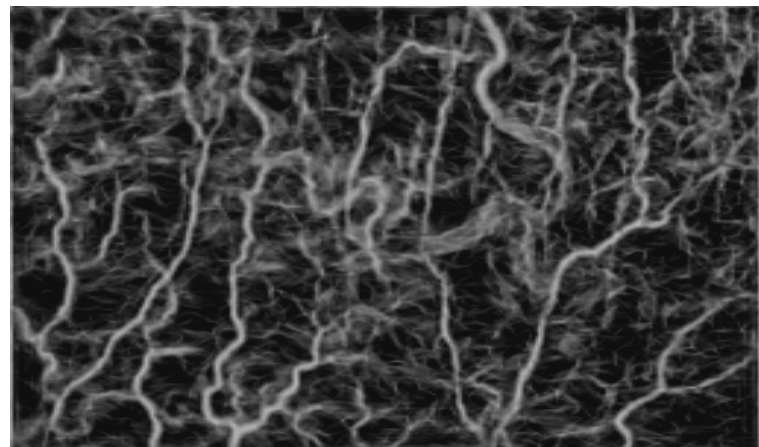


Infrared image



Vein and hand contour

???→

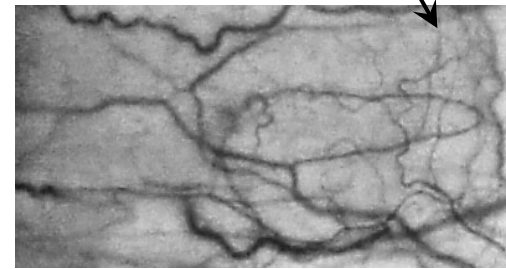
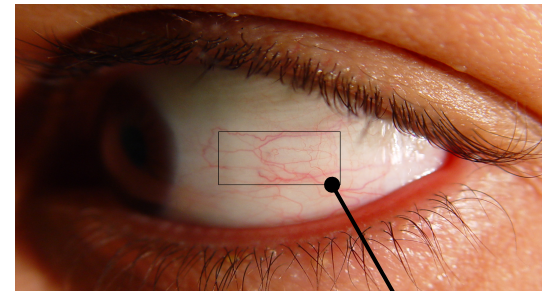
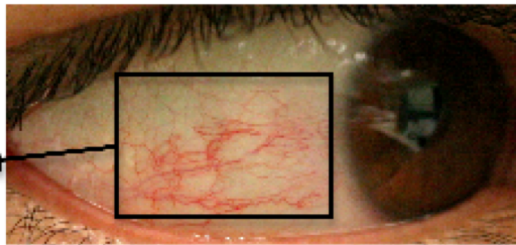
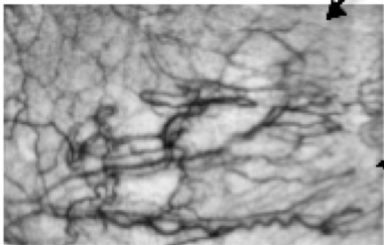
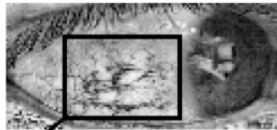
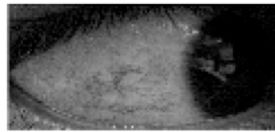
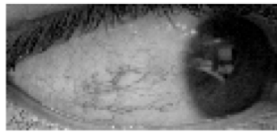
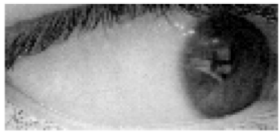


... But hard to get to unless



OR

- Scan the [only] part our body with visible vasculature: whites of the eye
 - Derakhshani R, Ross A, “Conjunctival Scans for Personal Identification,” US Patent number 7,327,860; issued on 2/5/2008.
 - Many thanks to CITEr for providing the seed funding for this research in 2005

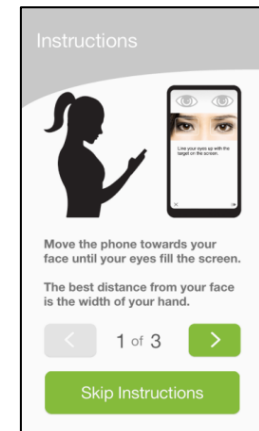
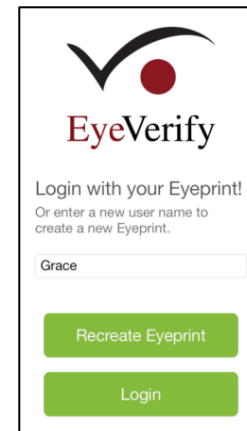
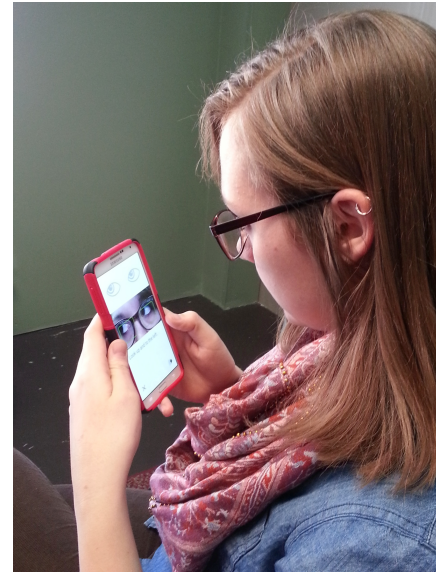


From dSLRs to Cell Phones

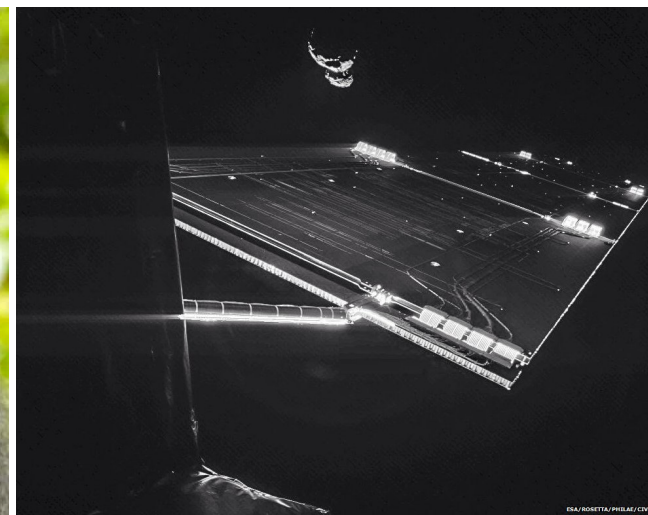
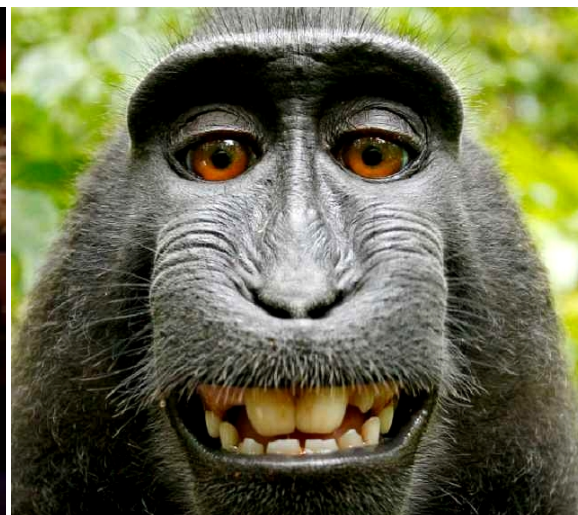
2005:



Now:



Selfies are in...



Rosetta takes 'selfie' ahead of landing site selection (BBC News)

Challenge 1

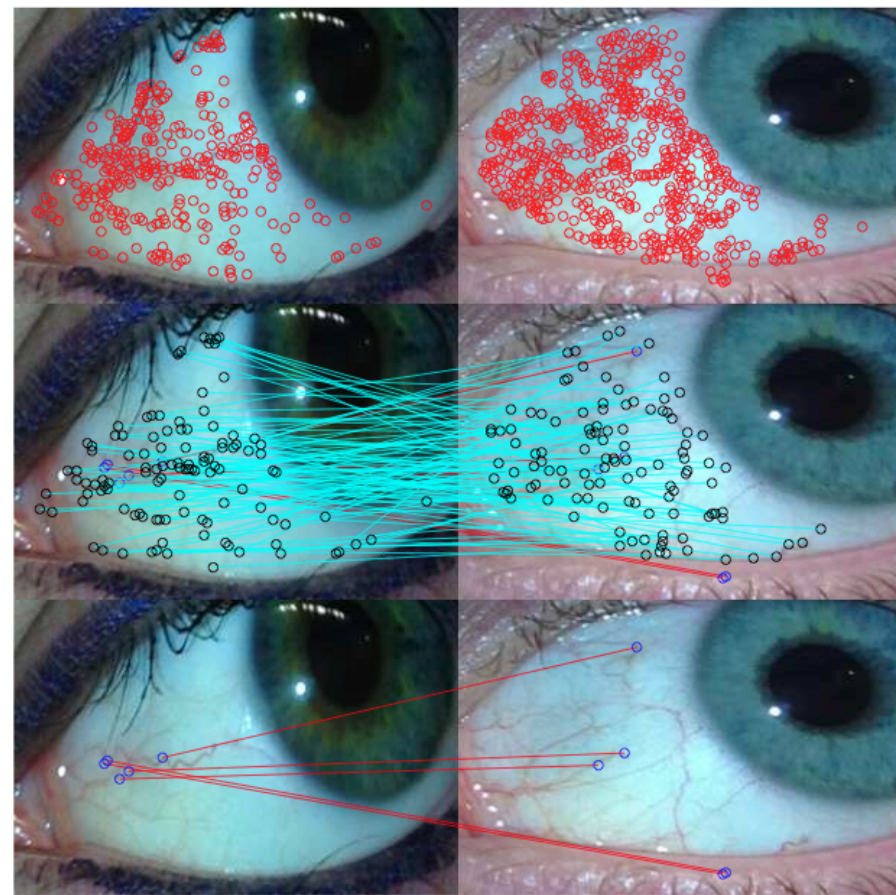
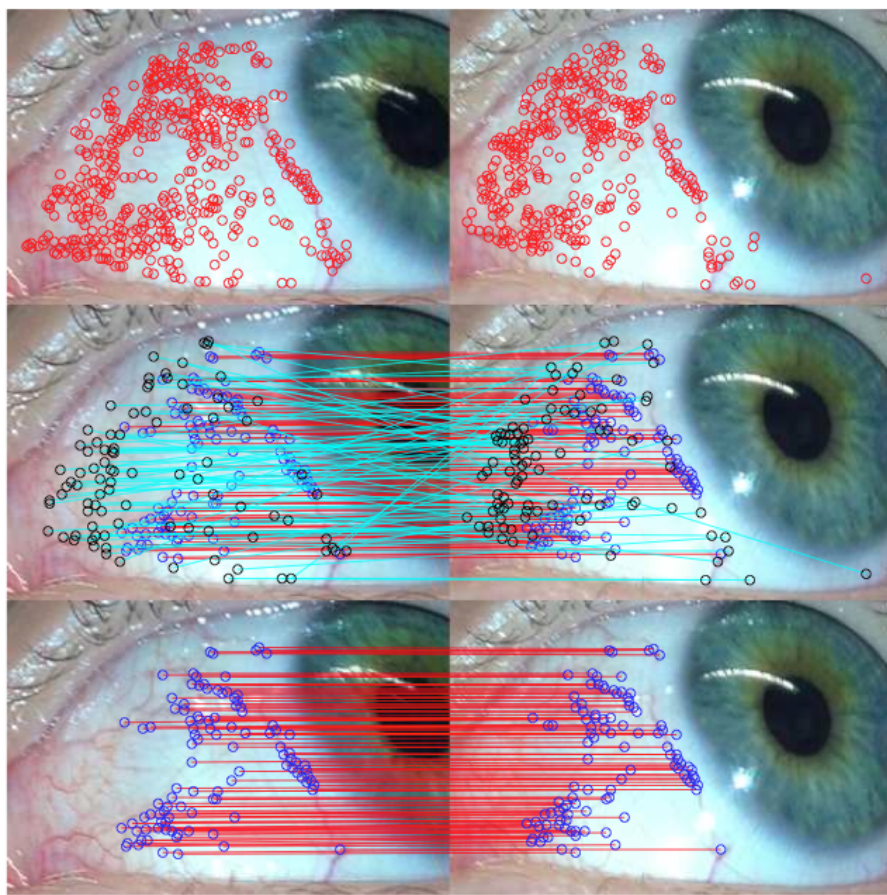
- *Front facing cameras* have lower quality sensors and optics
- Mitigation: computational photography
 - Multi-capture SNR boosting (exposure stepping and averaging)
 - PSF recovery/blind deconvolution
 - Multi-capture feature aggregation
 - Vesselness/eyevein-tuned filter banks



Challenge 2

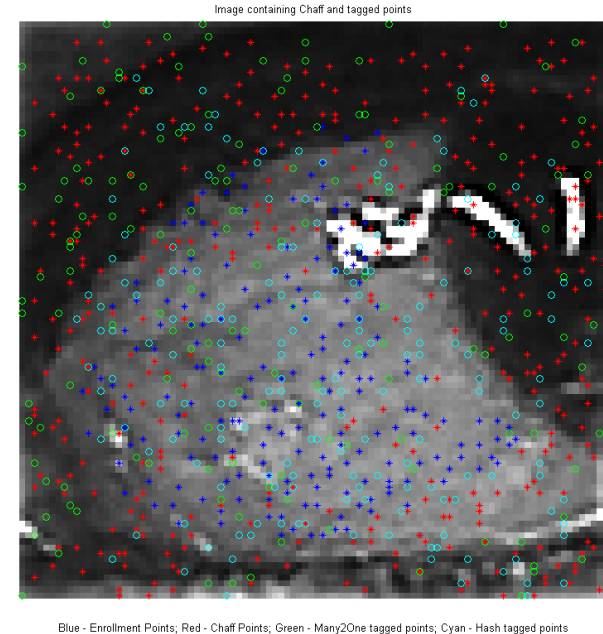
- *Mobile use case* imaging artifacts
 - Lighting, motion blur, glare and specularities, eyelid/eyelash/glass occlusions
- Mitigation:
 - Lucky imaging/no reference quality metrics
 - Robust features and matching/fusion routines amenable to partial data
 - Multi-sample and context aware multi-template match score fusion

Genuine and Impostor Comparisons



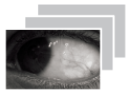
Challenge 3

- Template security and revocability
 - Steganography
 - Device-bound isometric transforms
 - Revocable, private, and high entropy private key generation
 - Stable despite biometric intra-class variations
 - Always resolves to a private key, but valid sequence only achieved during successful genuine comparison



To Recap:

Enrollment



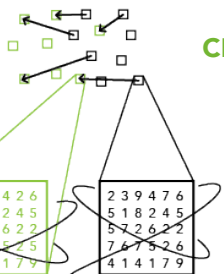
Step 1:
IMAGE CAPTURE



Step 2:
SEGMENTATION & ENHANCEMENT

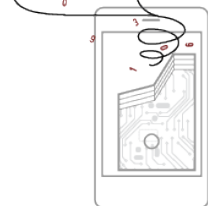


Step 3:
INTEREST POINT DETECTION & FEATURE EXTRACTION

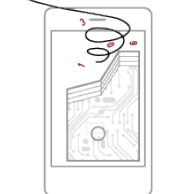
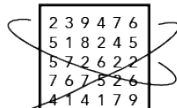
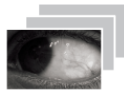


Step 4:
CHAFF POINTS ADDED & TAGGED

Step 5:
SCRAMBLE DESCRIPTORS

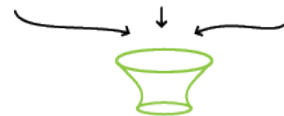


Verification



Enrollment

Eyeprint Key = 1113?2wqps.!2394^%2HrW481K&hHr5t0+1ds,7944vGaq%>5\$



Step 6:
EYEPRINT KEY ENCODING



Step 7:
MATCHING



Step 8:
EYEPRINT KEY DECODING



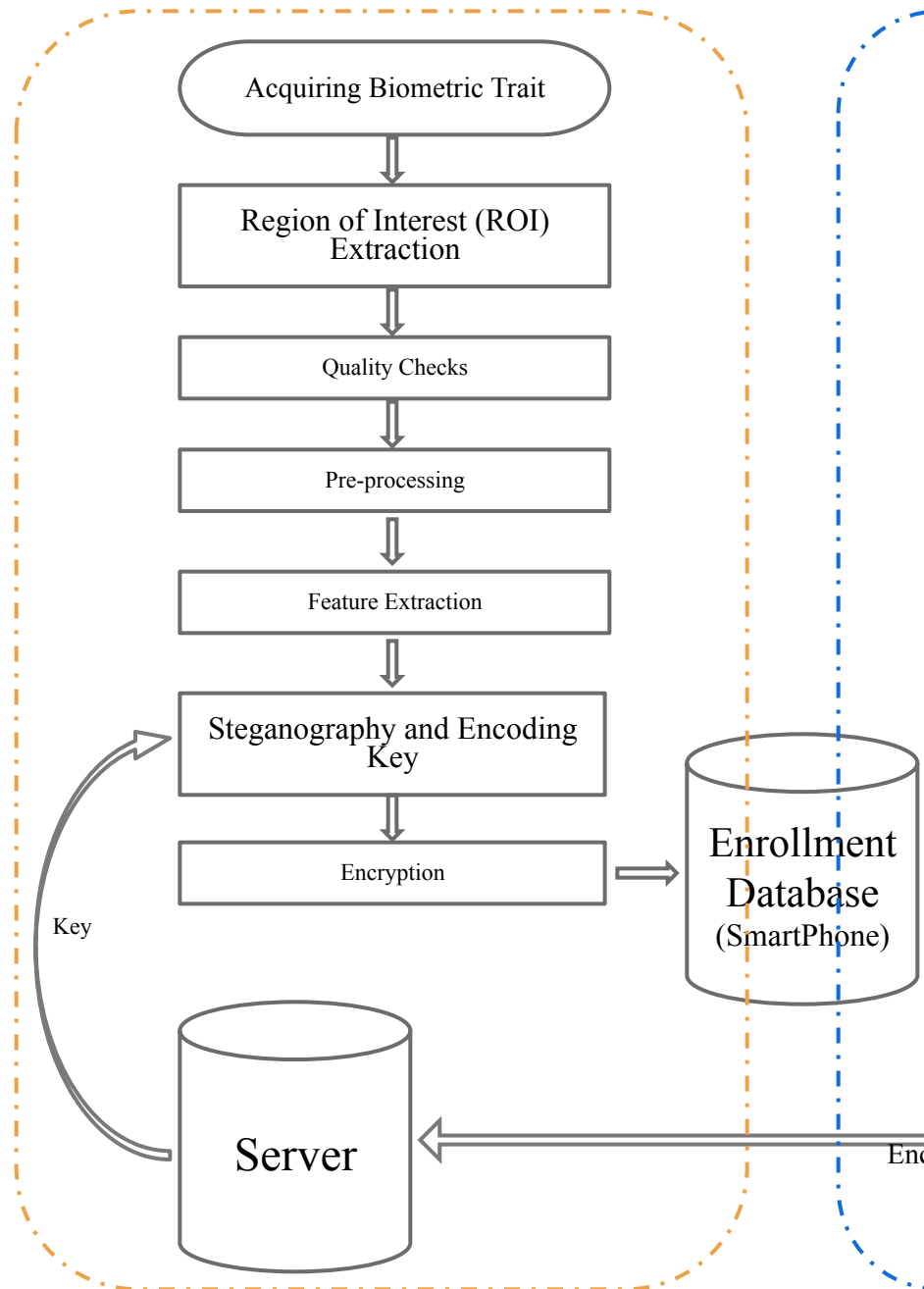
Eyeprint Key = 1113?2wqps.!2394^%2HrW481K&hHr5t0+1ds,7944vGaq%>5\$

Thanks

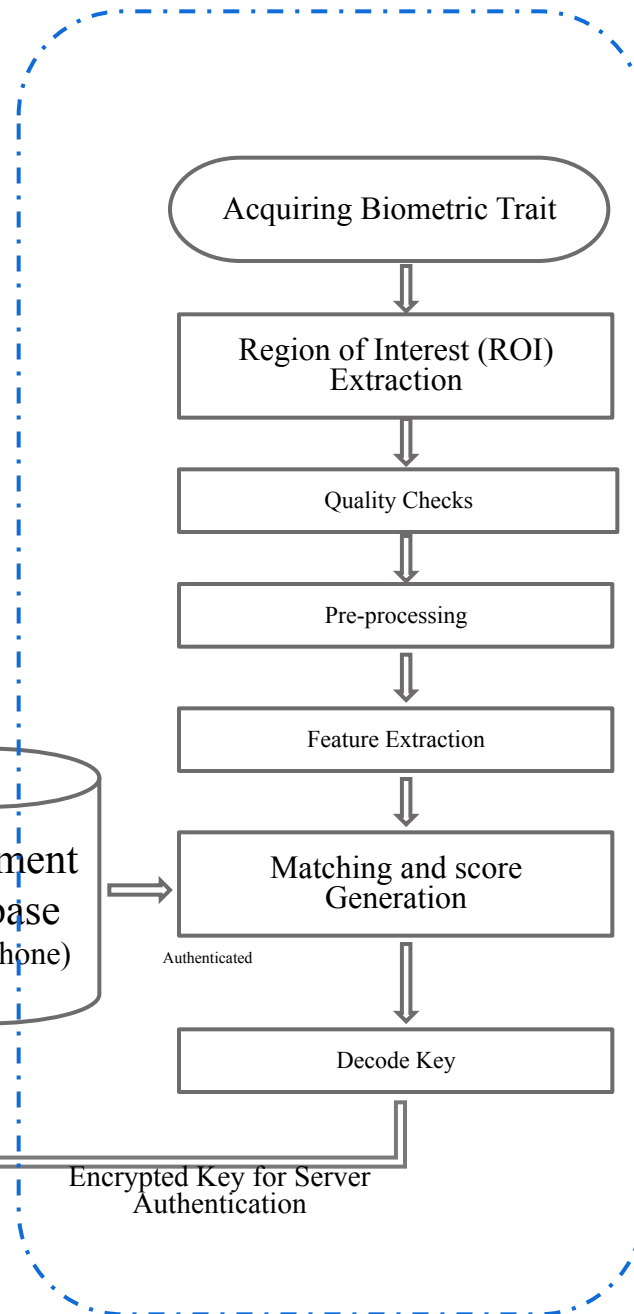
- To my R&D team and colleagues at EyeVerify, UMKC, MSU, and CITeR for their contributions and support
- Questions?

Extras Slides

ENROLLMENT



VERIFICATION



Segmentation

Back Facing Captures (8-Megapixel)

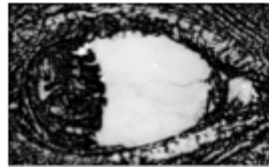
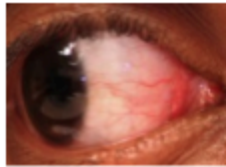
- Done using morphological operations and by measuring intensity variations in an image.

* Led by Pavan Tankasala

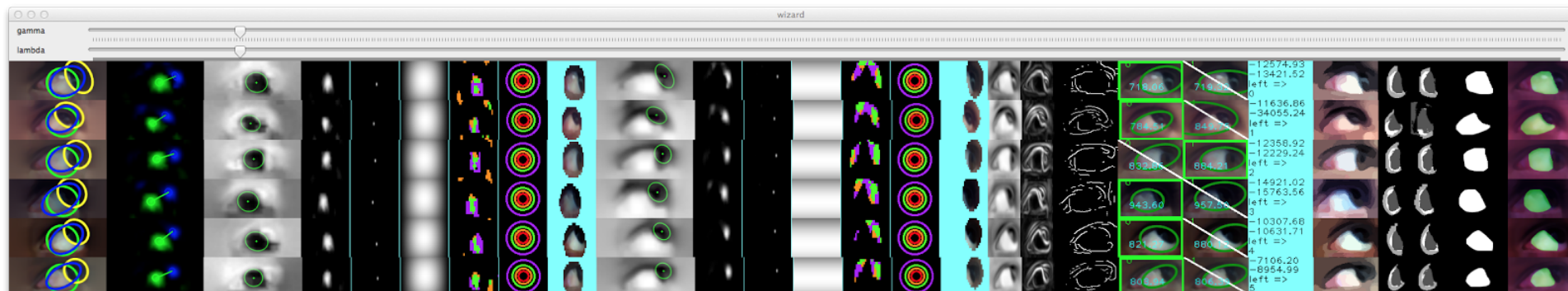
Front Facing Captures (1.3-Megapixel)

- Using the estimated eye shapes, upper eyelid, lower eyelid and limbus boundary across various color spaces, the definitive foreground and background are used to derive the final segmentation using GrabCut.

* Led by David Hirvonen



Above - Back facing captures. Below - Front facing captures



Accuracy

3rd Party Testing by Purdue University

[Biometric Standards, Performance and Assurance Lab](#)

- 250 subjects, broad cross section of races and ethnicities
- Independent data capture and analysis by Purdue
- Eyeprint FAR False Accept Rate = 2/100,000 at 0.88% FAR

[Detailed White
Paper available
here](#)

Stable Biometric Feature Set

All vein patterns including eye veins are absolutely unique to each human

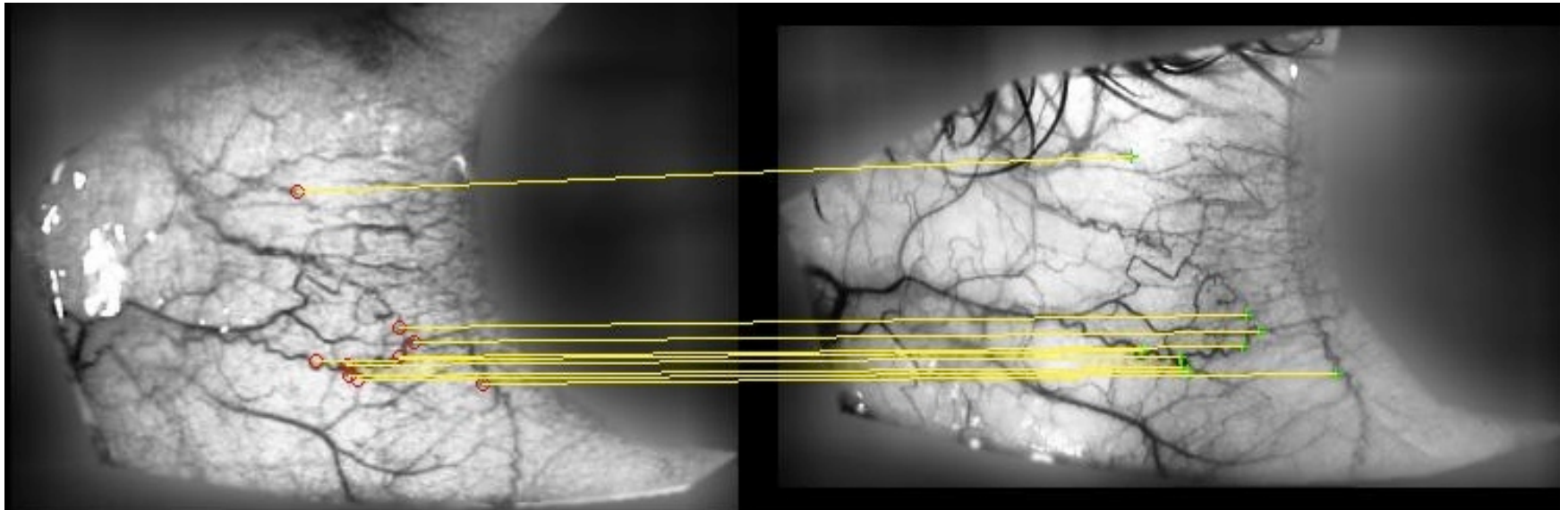
- Even identical twins and clones have unique vein patterns

Eyeprints are stable

- Research shows stability for a minimum of 2 years
- Anecdotal evidence indicates seven year plus stability
- Templates can be updated over time to ensure aging of user and templates remain consistent

2006

2013

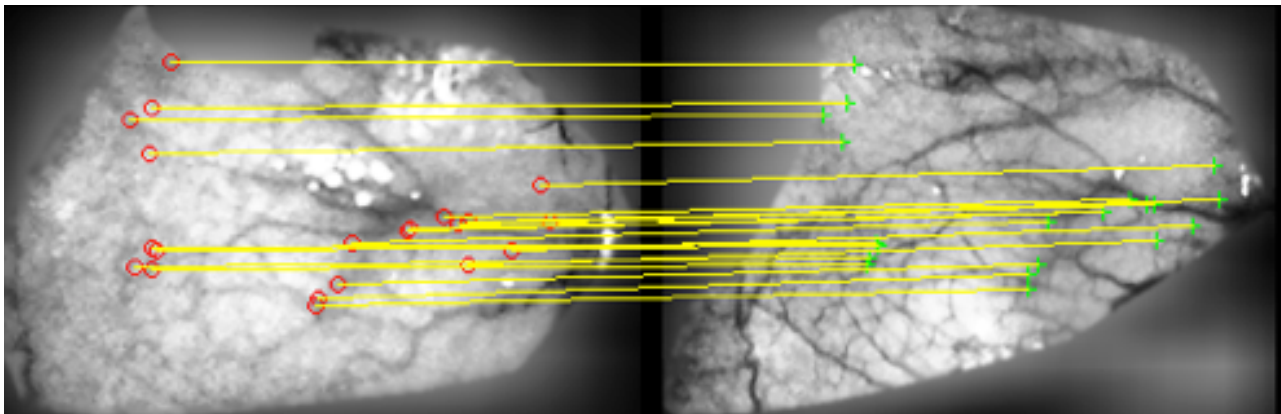


Images of the same person – seven years later - the vein patterns match perfectly

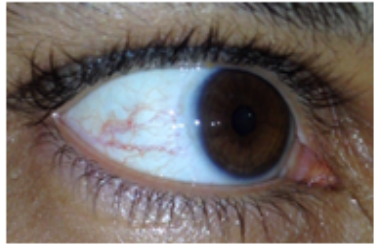
Stable Biometric Feature Set

Eyeprint users with contact lenses and eye glasses, even red and bloodshot eyes are no problem (hangover, dust, pink eyes)

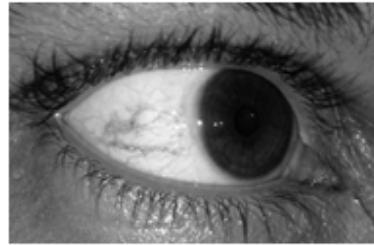
- Eyeprint algorithms look for strong vein patterns and descriptors
- Small capillaries or more pronounced veins have very little impact



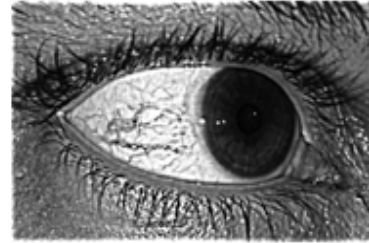
Preprocessing and Interest Point Finders



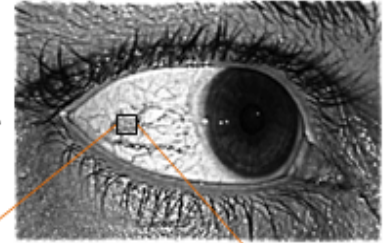
A



B



C



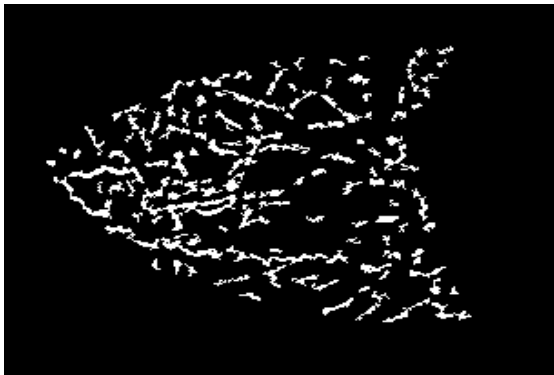
D

P3'			P2'			P1'
	P3		P2			P1
P4'	P4		P0		P8	P8'
	P5		P6		P7	
P5'			P6'			P7'

166	182	189	141	166	177	193
150	167	165	150	160	178	175
173	168	138	114	130	167	173
180	205	175	87	118	148	167
176	184	206	163	99	129	179
177	182	202	227	169	142	179
166	209	200	217	198	156	158

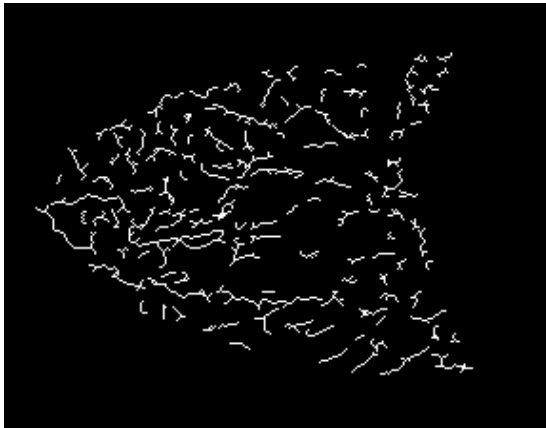
E

Finding Points of Interest

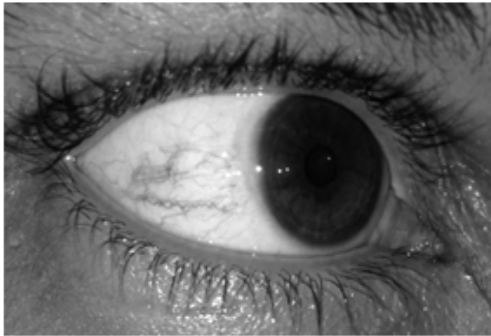


-1	0	1
-2	0	2
-1	0	1

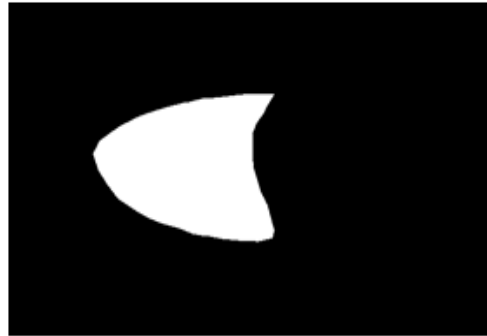
1	2	1
0	0	0
-1	-2	-1



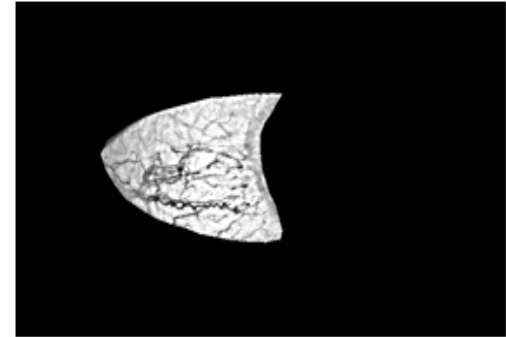
Result For Stage 1 Feature Extraction



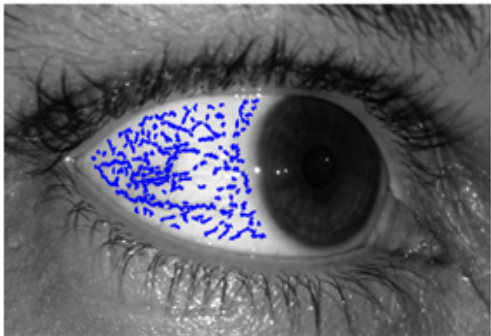
A



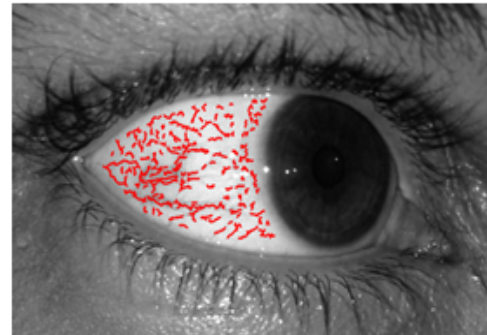
B



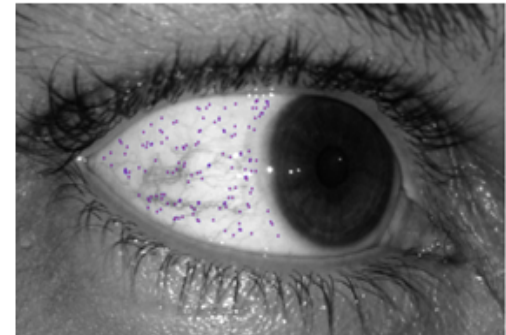
C



D



E



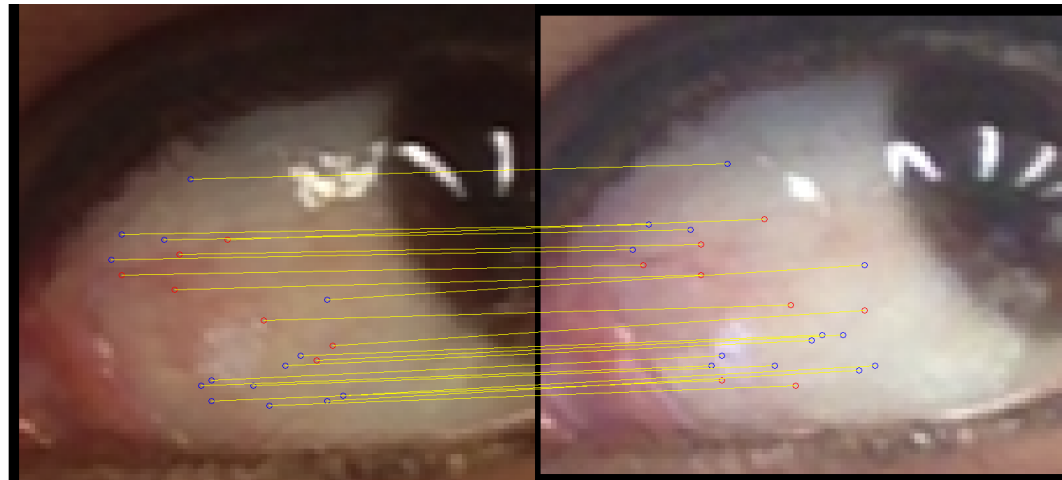
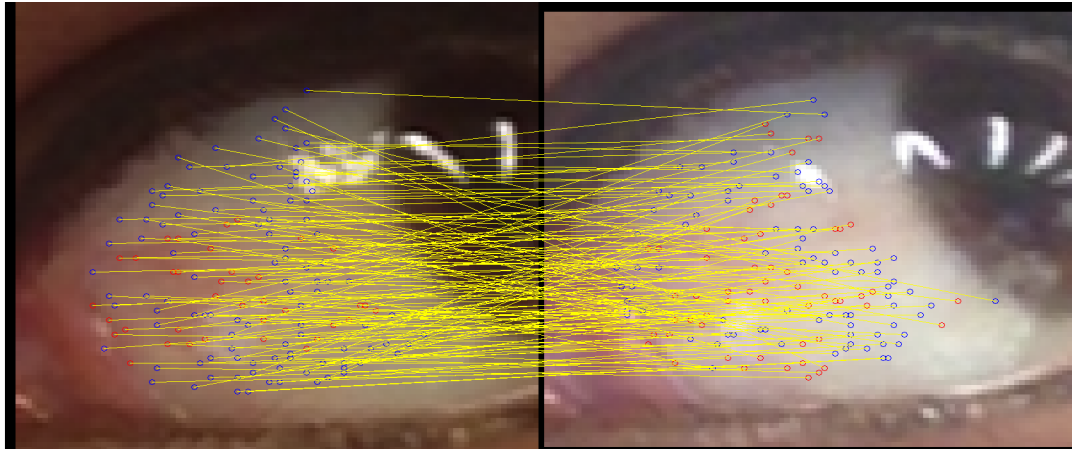
F

- Next: multi descriptor fused match score followed by hypothesis based registration and outlier detection for robust matching

Enrollment Image



Verification Image



Our New vs. Existing Descriptor Matches

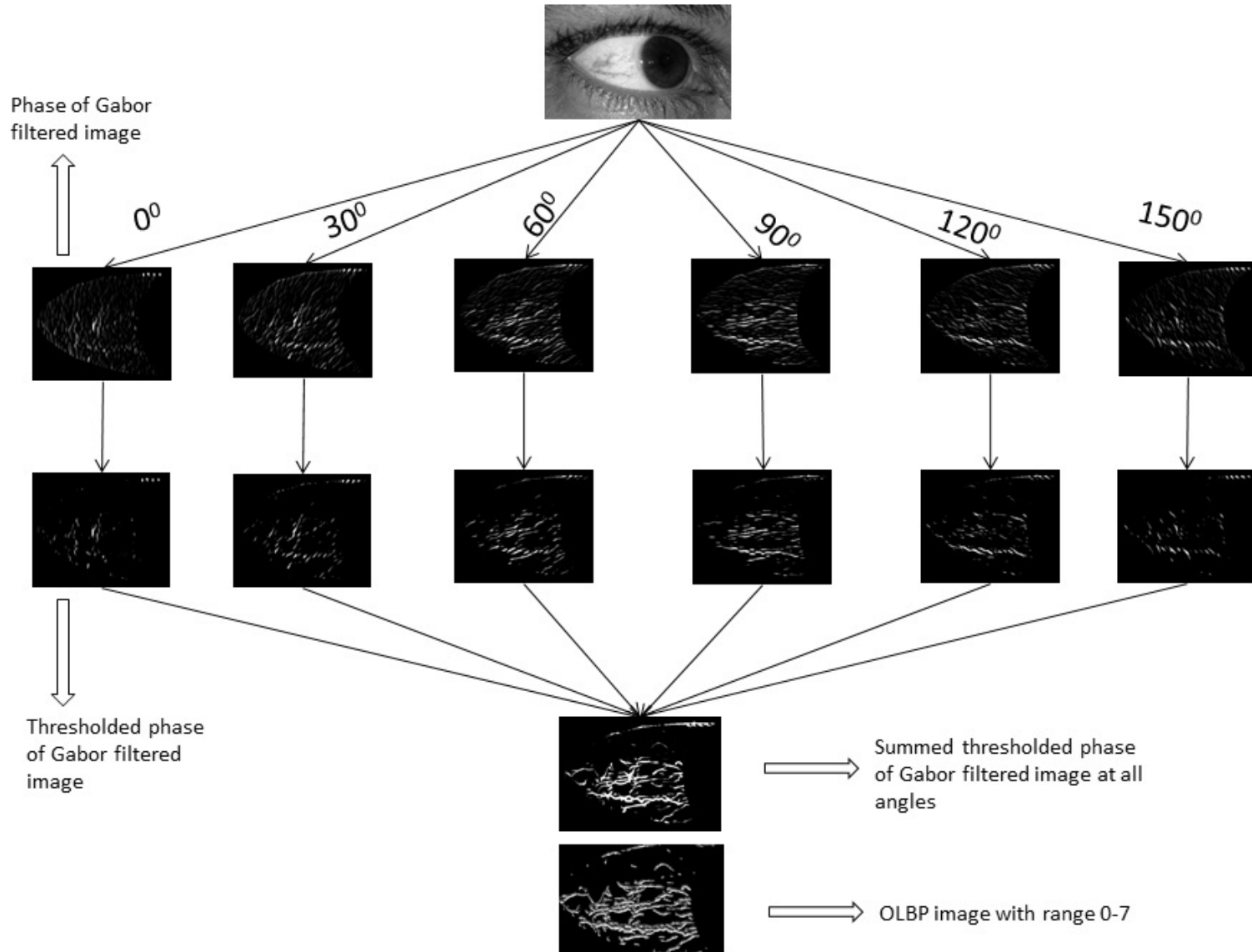
SURF Matches

LBP Matches

HLBP Matches

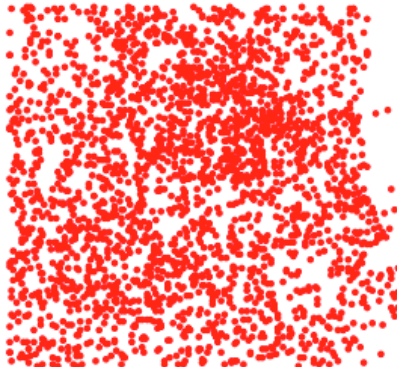
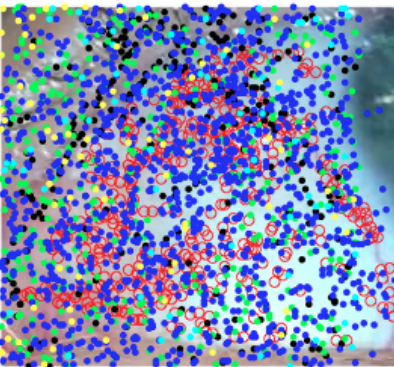
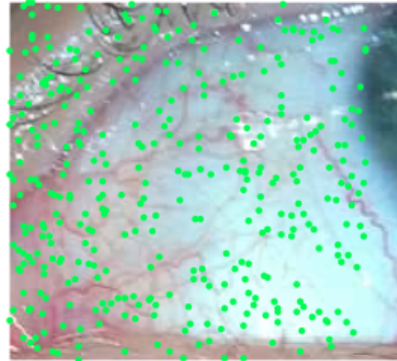
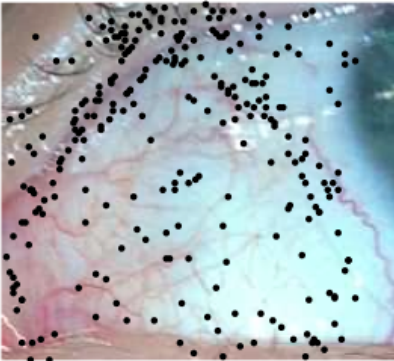
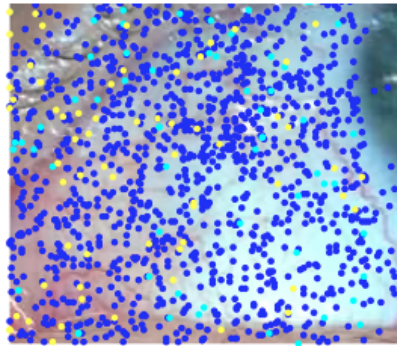
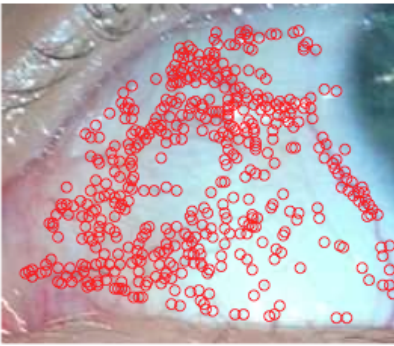
HDLBP Matches

Gabor Preprocessing



Template Security

- Templates are steganographed by insertion of chaff points and fake but statistically indistinguishable descriptors.
- The template is device-bound by additive distance preserving transforms, lossy compression loadings, and whitening.
- Valid released chaff handshake (using OTP hashes, different from key material) upon genuine verification with a third party for private server side surrogate verification and revocability, and possible attack vector detection.
- System of equations based key encoding and retrieval from properly tagged chaff (only upon successful authentication)



Obfuscation of original points (Left to right and top to bottom:

- OP
- {CP, TCP, SCP}
- TP
- SP
- {OP, TP, SP, CP, TCP, SCP}
- {OP, TP, SP, CP, TCP, SCP}