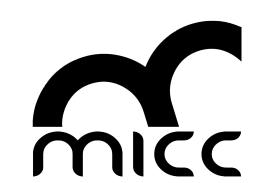


Smartphone Fingerprint by Mixing Device Sensors Features





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Introduction

In the next coming years, many of our basic activities such as reading an e-mail, checking our bank account, buying on-line, etc., will be performed by using a smartphone in a mobile environment. It is quite obvious that the degree of security granted by a classic username-password access is not sufficient and that a stronger level of safeness is required.

We present a possible solution which envisages the use of the user's own smartphone as a mean to grant a safer and easy mobile access. A novel methodology is introduced to obtain a robust smartphone fingerprint by opportunely combining different intrinsic characteristics of each sensor. Modern mobile phones, in fact, have several kinds of sensors such as accelerometer, gyroscope and camera; such sensors can be used to uniquely identify each phone by measuring the specific anomalies left onto the signals they acquire.

Gyroscope $\vec{F}_{Coriolis} = -2m\vec{\Omega} \times \vec{v}$

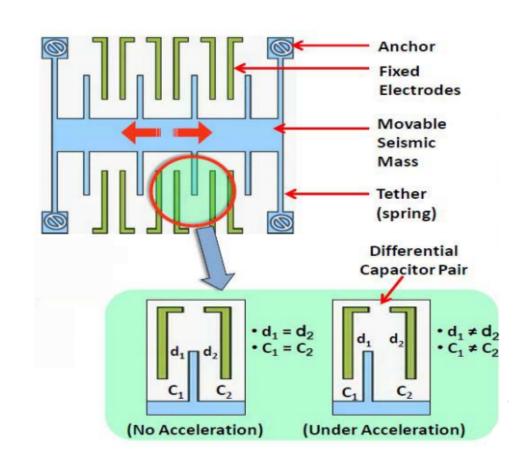
We read the output signals $\omega_x(k)$, $\omega_y(k)$, $\omega_z(k)$.

Features

We extract directly from readings a set of 21×3 features f_g in the temporal and frequency domain using the MIR Toolbox^a.

^ahttps://www.jyu.fi/hum/laitokset/musiikki/en/research/coe/materials/mirtoolbox

Accelerometer



We read signals $a_x(k)$, $a_y(k)$ and $a_z(k)$ and the associated time-stamp $t_a(k)$.

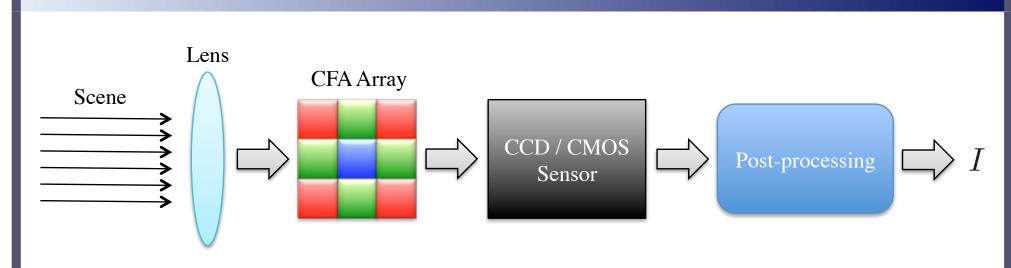
Features

We extract a set of 17×2 features^a f_a in the temporal and frequency domain from

$$T(k) = t_a(k+1) - t_a(k),$$

$$S(k) = \sqrt{a_x^2(k) + a_y^2(k) + a_z^2(k)},$$

Camera



An image can be modelled as

$$I = I^{(0)} + I^{(0)}K + N,$$

where $I^{(0)}$ is a noiseless representation of the scene, N is an additive noise term, and K is the multiplicative PRNU.

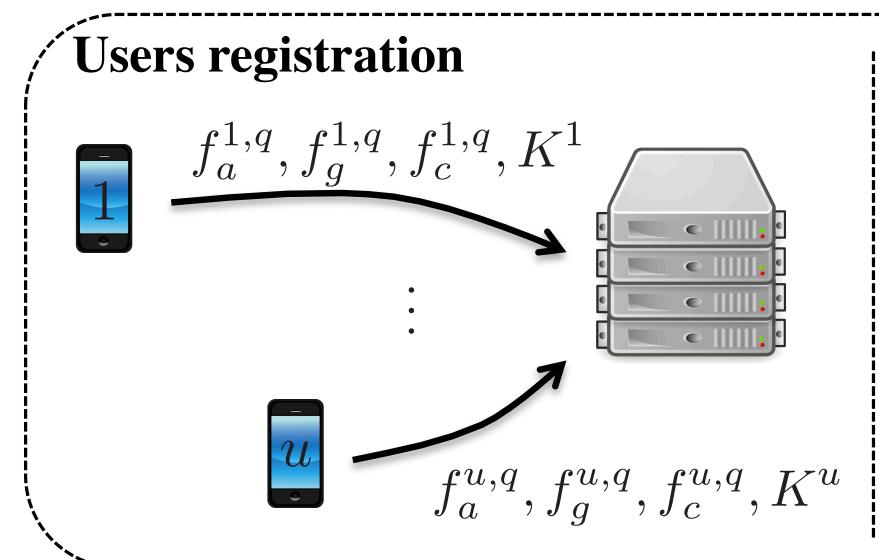
Features

Each camera is described by the vector

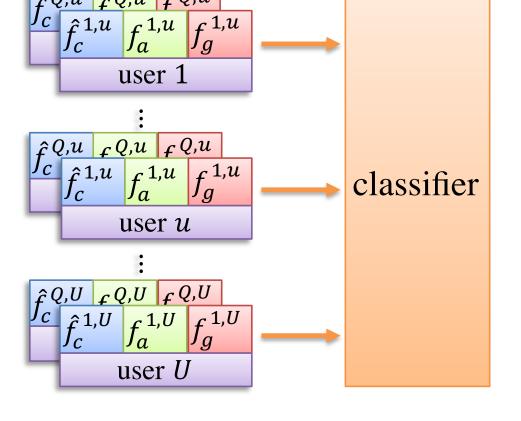
$$f_c = \operatorname{sign}(W_{512 \times 512}),$$

where $W_{512\times512}$ is the 512×512 central portion of the noise term W extracted from I.

Proposed system



Training $\hat{f}_{c}^{Q,u} f_{c}^{Q,u} f_{c}^{Q,u}$

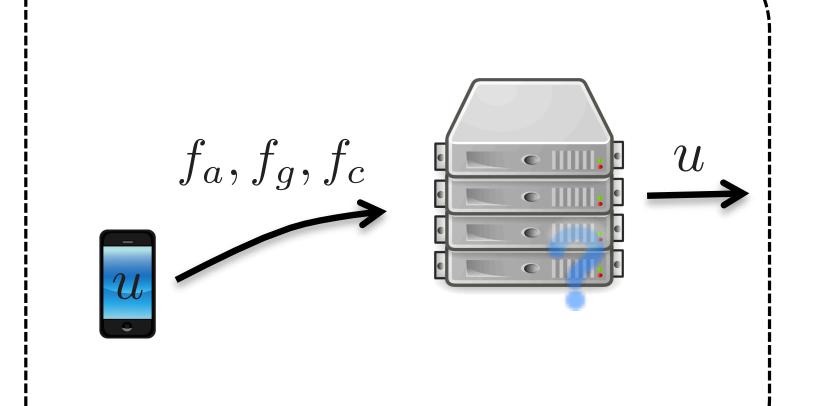


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10

CONF1

. User identification



PRNU estimation

$$K = \frac{\sum_{p} W_{p} I_{p}}{\sum_{p} I_{p}^{2}}$$

PRNU correlations

$$\hat{f}_c^{u,q} = \begin{bmatrix} \rho(f_c^{u,q}, K^1) \\ \rho(f_c^{u,q}, K^2) \\ \vdots \\ \rho(f_c^{u,q}, K^U) \end{bmatrix}$$

Results

Devices

Device	Amount
LG Nexus 5	5
Motorola Moto G 2015	1
Samsung Galaxy S3	2
Samsung Galaxy S4	1
Samsung Galaxy S2plus	1

Configurations

	Training set		Test set	
	Position	Vibration	Position	Vibration
CONF1	Table	ON	Table	ON
CONF2	Table	OFF	Table	OFF
CONF3	Hand-held	ON	Hand-held	ON
CONF4	Table	ON	Hand-held	ON
CONF5	Hand-held	ON	Table	ON

100 90 80 70 60 40 30

CONF3

CONF4

CONF5

CONF2