

Identification Technology of Mobile Phone Devices Using RFF

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Abstract: The vulnerability of the device identifiers, such as IP and MAC addresses, IMEI and IMSI codes, etc. creates threat to the information security, integrity and reliability. One of the solutions of this threat is usage of Radio Frequency Fingerprinting (RFF) technology for identifying wireless devices based on their unique radiation “fingerprint” as opposed to their addresses or codes. In this work identification problems of mobile radio stations (from here in – mobile phones) are being analyzed and identification methodology for identifying them based on the mathematical processing of front and rear fronts is proposed. All of this provides new insight in the field of signal detection and identification, thus by using this method only the original data is received. The purpose of this work is identification of a mobile phone, working on the DCS (digital cellular service) frequency, based on the phone’s radiated signal time characteristics.

1 INTRODUCTION

The aim of this work is to explore the identification possibilities of the mobile phone. Currently various identification methods, such as identification of the manufacturer’s model, according to the design of the mobile phone, or identification according to the physical and electrical parameters, or their entirety, are being used. It is known, that each wireless device has its own unique radiation characteristics (Danev and Capkun, 2009), (Hall et al., 2003).

Identification of the wireless devices based on the certain characteristics of the signal (phase, phase and frequency errors, etc.) is proposed by other authors (Hall et al., 2003), (Candore et al., 2009).

Technique to identify wireless device according to its radiation characteristics is known as Radio Frequency Fingerprinting. RFF are energy traces that are left in the radio frequency spectrum. They have certain characteristics that are emitted by every transmitter. RFF allows to separate certain unique characteristics that are radiated by every wireless device even if several devices having the same specifications are produced in the same plant (Danev and Capkun, 2009), (Hall et al., 2003), (Danev et al., 2012), (Danev et al., 2010). The essence of this technique is that wireless devices are identified according to the different radiation parameters such as the phase characteristics of wireless device (Hall et al., 2003), characteristics of various errors (Danev

et al., 2010), radiometric characteristics (Candore et al., 2009). From these characteristics using various mathematical models (such as Bayesian step change detector (Hall et al., 2003) or Fisher linear discriminant analysis (Danev and Capkun, 2009) the certain parameters, that allow determining the unique parameters of the transmitters, are calculated. Identification is done by analyzing initial transient signals.

By using RFF technique, identification system, which can correctly determine the radio transmitter, is formed. This system is an invaluable tool for militaristic and civil purposes, where unauthorized usage of electromagnetic specter is detected. It is very useful to identify and localize the source of the transmitted information. This system provides proof that unsanctioned or illegal radio transmission is being broadcasted (Shaw and Kinsner, 1997).

RFF technique is usually based on edge detection and analysis of their various parameters, because unique characteristics of every wireless device are present within the boundary of these edges.

This technique is easily used to identify transmitters, based on Bluetooth, 802.15.4 standard (Danev and Capkun, 2009), WLAN 802.11 standard (Danev and others, 2012), (Shaw and Kinsner, 1997), GSM (Zanetti and Lenders, 2012) and VHF (Danev and others, 2012) standards.

Analysis of the aforementioned works shows that practical usage of the identification techniques,

provided in these works, is met with certain difficulties. The first problem is that it is quite difficult to automate the identification process; the second problem is the complexity of the mathematical models, used in the identification process (Danev and Capkun, 2009), (Hall et al., 2003). In this work common math equations, which can be easily implemented in the automated identification system, are used. In this work, as well as in many of the previous works, the identification of the cell phone is based on the edge detection. In this work is used a completely new methodology based on discretization of the signals and description on the shape of the edge by using mathematical methods. In this paper we analyze the shape of the signal amplitude in various aspects but do not analyze characteristics of phase or errors.

2 EXPERIMENT AND HARDWARE

This experiment was performed at Kaunas University of Technology, Faculty of Telecommunications, Radio Link laboratory. During the experiment all mobile phones were functioning on DCS frequency (1800MHz). All phones were connected to the mobile network of the Tele2 Ltd. (from herein after Tele2), one of the mobile operators of Lithuania Republic. Algorithm of the experiment is presented in Fig. 1.

This algorithm consists of three main parts.

1. Edge Detection. In the first part detection of rising and falling edges is performed. This is necessary to perform further calculations for determining the edge curvature.

In this part the rising is detected. Firstly the spectrum of the edge is calculated and 1ms length part of the spectrum is taken. This part of the spectrum is further discretized each 1 μ s. For each discrete point the first derivative according to the last discrete value is calculated. Rising edge boundaries are detected according to the change of this derivative.

After that boundaries of the falling edge are detected according to the change of this derivative. When change becomes significant enough, it is considered, that the falling edge has begun. End of the falling edge is considered as a point, when change of derivative becomes low.

2. Calculation of Parameters. Calculation of parameters is performed in the second part.

Calculation of the following rising edge parameters is performed: the first and second derivatives, curvature, slope coefficient. These parameters are the central part of the work, since they mathematically determine the shape of the edge. If edge is a line, edge parameters of every cell phone would be similar, but in reality they are different.

3. Identification of the Device. In the third part a matrix for identification of the device is created.

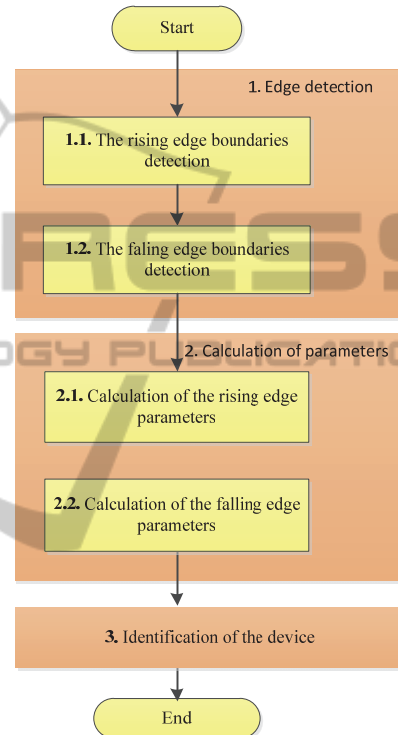


Figure 1: Algorithm of the experiment.

The following cell phones were tested: Nokia X3, Nokia C3, Nokia 3600, Nokia 7260, two Nokia 6230i (further referred to as A and B), Samsung E390.

During the work, spectrum analyzer Rohde&Schwarz FSH8 was used. Its operating frequency band is 9 kHz – 10 GHz. This device ensures good sensitivity without additional amplifier (up to – 141 dBm). Antenna used in this work was omnidirectional and calibrated for 18700 MHz frequency band.

During experiments in the laboratory, WLAN networks were detected, but they did not interfere with the experiments because their frequencies were different (WLAN operates in 2400 MHz band and experiments were performed in 1800 MHz band).

Experiment was performed in strictly controlled environment and on identical system settings. During all experiments the same antenna, connector cables and spectrum analyzer were used. To enhance the transmitted signal and to avoid using additional amplifying equipment we chose to use a small distance between cell phone and spectrum analyzer is 0.5 m. For each phone 80 measurements were performed. Correlation coefficient of the measurements was >95%, and error less than 8%.

Transmitted signal was initiated by call from the mobile phone, thus monitored transmitting frequency band was 1758-1782 MHz. Transmitted signal frequency characteristics were collected from 3hr long calls by making a call each 2 minutes. Results were collected on different days. No significant differences were detected. Thus it can be said that further experiments for certain band should be easily performed. Stable conditions for all the cell phones, which were used in the experiment, must be assured. For the remaining part of the experiment a certain frequency, in which all the following steps of the experiment will be performed, is chosen. A 1765 MHz frequency was chosen, because signal amplitude (strength) is strongest on this frequency.

3 RESULTS

As shown by the experiments, rising and falling edges of all cell phones (power vs time), even of the same brand (Nokia 6230i A and B), were different (Fig. 2, 3).

Theoretically, duration of the rising and falling edges are not accurately determined (but maximum duration of the edge can be 28 μ s), thus it can be different (Molisch, 2011). Result analysis clearly shows that longest duration of the edge change was 17 μ s (Fig. 2). Thus it was decided that rising edge will be sampled for a period of 17 μ s with a sampling time of 1 μ s.

By sampling signal in the time axis, we obtain specific signal amplitudes for each sampling step. In this case we chose to discretize signal each 1 μ s.

Falling edge of the signal is shown in Fig. 3. As in the case of the rising edge, falling edges of all 7 cell phones are shown next to each other. Maximum duration of the edge change was 16 μ s, but, to meet the identical conditions, discretization period was set to 17 μ s.

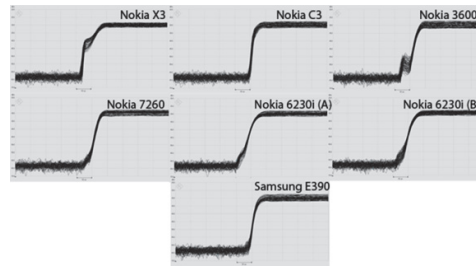


Figure 2: Rising edge.

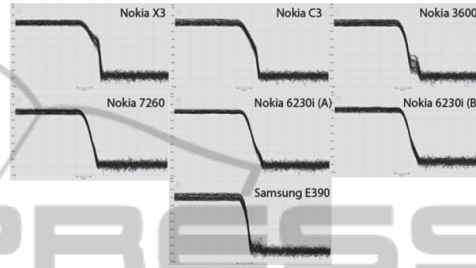


Figure 3: Falling edge.

Average amplitudes of all rising edge signals are shown in Fig. 4. Also noise level is addressed here. It is obvious, that edge shapes of individual cell phones are different. We can clearly see differences between different mobile phones: their uniformity, curvature, etc. In example, cell phone “Nokia X3” has extra clear and wide signal edge curve. During the experiment this was one of the most visually apparent differences between rising fronts, created by sending signals from this call phone. On the other hand rising edge of the “Nokia C3” cell phone is below the noise level until approximately 7 μ s, thus we can conclude that the rising edge of this cell phone is shortest as well as steepest. Also rising edge on a “Nokia 3600” cell phone was particularly interesting: it has clear directivity and breaking points.

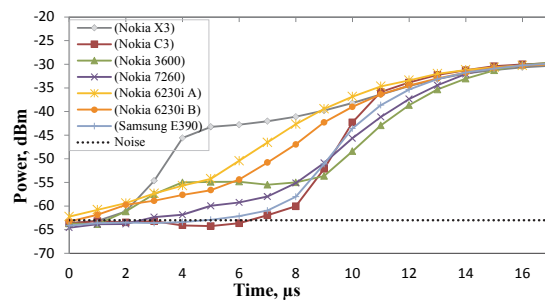


Figure 4: Power of the rising edge of the cell phone signal as function of time (averages).

Falling edges of 7 cell phones are shown in Fig. 5. As we can see from the charts, as in the case

of rising edge, “Nokia X3” has a clear breaking point. Signal curve of cell phone “Nokia S3” is very similar to the curve of “Nokia X3”.

Experiment results of two identical cell phones “Nokia 6230i” should be specifically mentioned. The first call phone was marked with letter “A”, second – with letter “B”.

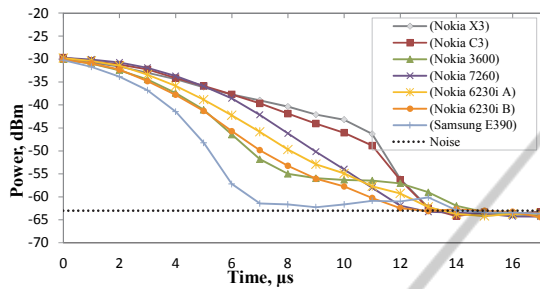


Figure 5: Power of the falling edge of the cell phone signal as function of time (averages).

In references (Danev et al., 2010), (Shaw and Kinsner, 1997) it is noted, that signal characteristics of the cell phones of identical manufacturer and brand should be quite similar or have slight differences.

As we can see from Fig. 4 and 5, the difference between such cell phones is not too big, but it is quite significant. It will be later shown that this difference, after mathematical processing, will become much more apparent. Falling edge of the cell phone “Nokia 3600”, as its rising edge, has an apparent directional variation and several breaking points during a 17 μs period. On the other hand falling edge of the “Samsung E90” rapidly falls down close to noise level but, as shown in the diagram, later rises slightly after a rapid fall. Duration of the falling edges is relatively slightly shorter than the duration of the rising edges because falling edges of most mobile phones drop down to noise level after 14 -15 μs.

From Fig. 4 and 5 we can form an opinion that GSM packets of a specific tested cell phone have visually similar rising and falling edges. In example, looking at “Nokia 3600” transmitted signal edges it is visually apparent that shape of the falling edge is similar to the shape of the rising edge – breaking points and directional variations are clearly visible in both rising and falling edges. The same applies to both “Nokia 6230i” cell phones marked by orange and yellow colors. It has to be said that in the cases of other cell phones, i.e. “Samsung E930”, rising edge is different from the falling edge. Despite that later mathematical processing showed, that these

fronts are not their own “mirror images” according to their shape.

As we can see from Fig. 2-5, rising and falling edges of each phone are curves and they can be described by the 1st and 2nd order derivatives, edge curvature and edge slope coefficient.

The First Derivative. It is known that the first derivative shows the speed of the functions quantity change. Signal edges, obtained during this experiment can be described by derivative, showing the speed of the amplitude change. The higher this number, the greater the change in comparison with the previous value. As the derivative approaches zero, the edge of the signal becomes straighter (the form change stops). By using the results of the first derivative we can detect the borders of the signal edge, used to assure identical conditions to all tested cell phones.

A combined chart of all 7 tested cell phones first derivatives (in relative units) of the rising edge is shown in Fig. 6.

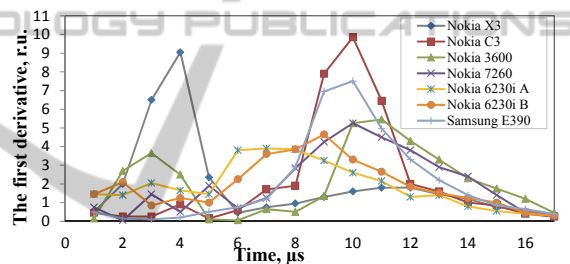


Figure 6: The first derivative of the rising edge.

As we can see from the combined chart, the first derivatives of each cell phone are different. It is most visible with the cell phones “Nokia X3” and “Nokia C3”. The first derivatives of other cell phones are slightly lower, which means that the changes of certain intervals are lower. A combined chart of all 7 tested cell phones first derivatives of the falling edge is shown in Fig. 7.

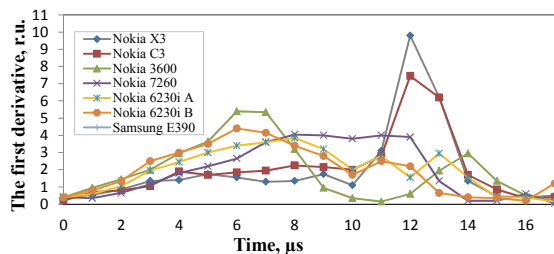


Figure 7: The first derivative of the falling edge.

Thus the charts of the first derivatives shows the intensity of the function change over time. The

higher the intensity of this change results in higher change of the amplitude and more apparent edge curvature. In Fig. 6 and 7 we can see that derivatives of the cell phones “Nokia X3” and “Nokia C3” are the highest and in Fig. 4 and 5 we can see that the edges of these phones are distinguished by their individuality.

The Second Derivative. The second derivative shows if a function has a breaking point (a point where it’s direction changes) in the function change interval. Existence of the breaking point is considered the point where function of the second derivative crosses zero axis. Results of the second derivative of the rising edge are shown in Fig. 8 and of the falling edge in Fig. 9.

Thus in these figures we can see a clear differences between the second derivatives of the edges of each phone. Changes of the second derivative are especially clear in the case of a “Nokia X3” with highest change being at 5 μs. Also the value of the second derivative is big for the cell phone “Nokia C3”. It is also apparent that the second derivatives of various phones cross the y = 0 axis different number of times: 3 times for “Nokia X3”, 4 for “Nokia C3” (Fig. 8). The time of these crossings is different in each case as well.

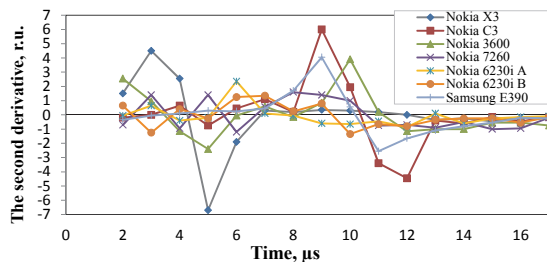


Figure 8: The second derivative of the rising edge.

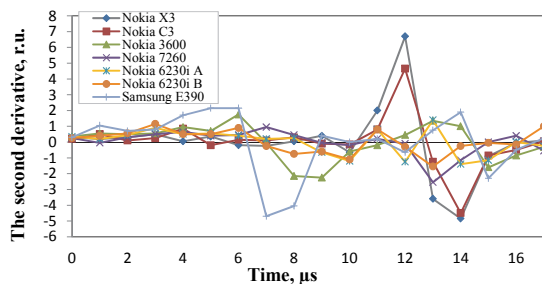


Figure 9: The second derivative of the falling edge.

The second derivatives of all phones first points are not large, since as it is shown in Fig. 6, the shapes of falling edges for all phones are bending down slowly and do not change the direction of approximately up to 10 s time interval. In this case,

mobile phone “Samsung E390” stands out, as the second derivative here has a breaking point and changes sign already at 7 μs. The highest points of the second derivative are also at 12 μs of falling edges curves for “Nokia X3” and “Nokia C3” mobile phones. Curvature of these phones has been observed at the first derivative as well.

Calculated the second derivatives of two mobile phone signals for the same model “Nokia 6230i” visually looks similar. However, under the analysis it can be seen a sufficient difference. This difference apparently could be influenced by a wider shape (form) of mobile phone falling edge (Fig. 5).

Curvature and slope coefficient. The curvature K of the curve can be calculated using the equation (1) (Curvature definition, 2010):

$$K = \frac{|y''|}{[1 + (y')^2]^{3/2}} \quad (1)$$

where y' and y'' is first-and second-order function's (in our case it is power) derivatives respectively.

Slope coefficient b is an expression showing the dependence of average of the random variable on the other variable (several variables). Slope coefficient is calculated using equation (2). It shows how the signal edge steeply arises (or descends) (SLOPE function definition):

$$b = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sum (x - \bar{x})^2} \quad (2)$$

where x (in our case it is time) and y (in our case it is power) are argument and function respectively.

Curvature is calculated using the first and second derivatives. The curvature is calculated at each point of the sampling and shows sharpness of the curve is bending. In this case, the greater the curvature at that point is the stronger deformation is observed. All that is well seen by looking at the waveform. Meanwhile, the curvature is less, the shape of signal edge is straighter and smoother. Curvature is equal to 0 if the waveform is straight. The curvature of rising and falling edges are shown in Fig. 10. As it can be seen from this figure, curvatures of different edges are significantly different. “Nokia 3600” and “Nokia X3” have large curvatures for the rising edge, while phone model „Samsung E390“has the largest curvature of falling edge. Curvatures of the two mobile phones model of “Nokia 6230i” are also clearly different.

As mentioned above, the slope coefficient will show the signal edge slope. This means that the slope coefficient is higher, the mobile phone edge is steeper. Slope coefficient is calculated from all sampling points, i.e. using 18 points. In Fig. 11 we can see that the lowest slope coefficient is in the case of mobile phone “Nokia X3” raising edge. Meanwhile, the highest slope coefficient is in the cases of “Nokia C3” and “Samsung E390” signal rising edges.

Thus, the presented results have shown that it is possible to describe the rising and falling edges of mobile phone transmitting signals using simple mathematical categories. Following the creation of corresponding matrix of such data, identification of the mobile phones should be possible. The greater number of parameters in a matrix, the higher probability of the mobile phone identification. However, the assessment of a larger number of parameters and the formation of the matrix will be done in subsequent works.

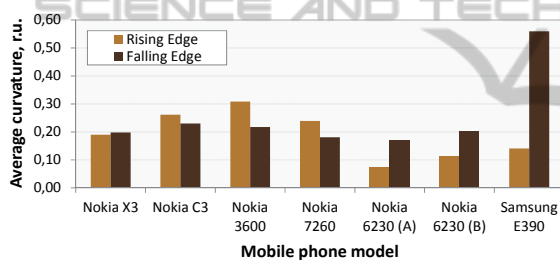


Figure 10: Edges average curvature dependence on the mobile phone model.

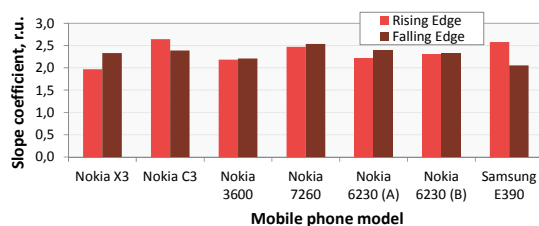


Figure 11: Edges slope coefficient dependence on the mobile phone model.

Some issues are planned for further works: to describe accurately identification algorithm for mobile station, perform measurements with the mobile station in further distance from the spectrum analyzer, perform measurements with other wireless devices.

4 CONCLUSIONS

Measurements have shown that rising and falling edges of transmitted signal for different mobile phones vary even when the phone models are the same.

The experimental results of measurement are estimated to not exceed 8% of the error; the correlation coefficient of measurement results for each model is > 0.95 .

New digital wireless device identification method based on assessment of signals rising and falling edges form has been proposed. This method is based on the calculation of rising and falling edges parameters using mathematical methods – the first and second derivatives, curvature and slope coefficient.

Number of identification parameters will be increased and the algorithm for the identification of devices will be proposed in further works.

REFERENCES

- Danev B., Capkun S., 2009. *Transient-based Identification of Wireless Sensor Nodes*. IPSN'09, San Francisco, California, USA.
- Hall J., Barbeau M., Kranakis E., 2003. *Proceedings of LASTED International Conference on Wireless and Optical Communications (WOC), Banff, Alberta*.
- Candore A., Kocabas O., Koushanfar F., 2009. *Robust Stable Radiometric Fingerprinting for Wireless Devices*. IEEE International Workshop on Hardware-Oriented Security and Trust, San Francisco, CA, USA.
- Danev B., Zanetti D., Capkun S., 2012. *On Physical-layer Identification of Wireless Devices*. ACM Computing Surveys (CSUR).
- Danev B., Luecken H., Capkun S., 2010. *Attacks on Physical-layer Identification*. WiSec '10 Proceedings of the third ACM conference on Wireless network security.
- Shaw D., Kinsner W., 1997. *Multifractal Modelling of Radio Transmitter Transients for Classification*. WESCANEX 97: Communications, Power and Computing. Conference Proceedings., IEEE
- Zanetti D., Lenders V., 2012. *Exploring the Physical-layer Identification of GSM Devices*.
- Molisch A. F., 2011. *Wireless Communications*, John Wiley and Sons, 2nd Edition.
- Curvature. Definitions and Examples. College of Science and Mathematics 2010. URL: <http://science.kennesaw.edu/~plaval/math2203/curvature.pdf>
- SLOPE function definition on Microsoft Office support resources. URL: <http://office.microsoft.com/en-au/excel-help/slope-HP005209264.aspx>