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Thoroughly analyzing the use of Ring Oscillators for on-chip Hardware Trojan detection

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Abstract—With the globalization of the IC design flow, structural integrity verification to detect parasitic electrical activities has emerged as an important research domain for testing the genuineness of an Integrated Circuit (IC). Sensors like Ring Oscillators (RO) have been proposed to precisely monitor the internal behaviour of the ICs. In this paper we propose an experimental analysis of the impact of parasitic electrical activities on the frequencies of ROs and on the internal supply voltages measured. Our observations lead us to identify the limits of the usability of ROs for practical and embedded detection of Hardware Trojans.

I. INTRODUCTION

Due to the recent trend of outsourcing Integrated Circuit (IC) manufacturing and design, structural integrity verification of ICs has become a hot topic. From the specification step to that of packaging, and especially during the design step, a circuit can be corrupted by a malicious adversary. This malicious modification of an IC structure is called a Hardware Trojan (HT) insertion and its effects can range from performance degradation (e.g. denial of service) up to more sophisticated effects like memory dumping, etc. [1].

A HT is composed of two parts, the trigger and the payload. The trigger is the mechanism that scans specific signals within the IC until a specific condition is met. When this condition is met, the payload is then executed. The trigger can either be generated externally (an external signal or a special external physical condition) or internally (an internal state of the IC, a special data configuration, etc.). Moreover the trigger can either be combinational (where the sought condition is the result of a logical operation on several signals) or sequential (the triggering of the payload is related to a succession of internal states). The payload is the ‘malicious effect of the HT. The payload is ‘explicit’ when signals or logic blocks are directly added, removed or deactivated. It can also be ‘implicit’ which is the case when the effect cannot be directly observed like, for example, the thinning of particular wires. HT detection is a challenging problem, even after its activation when the payload is implicit; and it is even more difficult if we try to detect the HT before its activation. One reason is that the probability of triggering a HT during functional tests is low and testing is an expensive approach to that end. Inspecting the circuit through reverse engineering is also an expensive process in terms of costs and time which is destructive by nature for the device under inspection. If this method can guarantee with a high level of confidence the absence of HT in the de-processed IC, it doesn’t guarantee at all the absence of HT in all the remaining ICs that have not been (destructively) inspected. Even though latest imaging based methods have proven to offer a really high confidence level [2], the approach is still destructive and does not cover the entire population of ICs.

Several non destructive methods for detecting HTs have been proposed since 2007. The first proposed approaches have suggested to analyze, using statistical techniques, the overall consumption of an IC to detect the impact of a HT. For example, in [3], a detection technique based on the Karhunen-Loève theorem is proposed in order to identify the power consumption of a HT within process variations and noise. However, this paper only reports validations obtained by simulations which unfortunately do not take into account things like measurement noises. In addition, one may wonder if a global monitoring of an IC is an adequate solution. Moreover, the technique may not scale to complex Systems on Chip (SoC).

In order to enhance the detection capabilities of the proposed techniques, in 2008, the authors in [4] analyze locally the propagation delays of logical paths using embedded monitors. However, once again, only simulation results are provided. In order to improve detection capabilities, [5] proposes to integrate a hardware system allowing to monitor important nets of ICs. However, little information is given in this paper about the efficiency of the proposed technique or on the associated silicon overhead. In parallel to this approach, a test solution was also proposed in [6] in order to trigger easily a HT or at least to increase its electrical activity. In 2010, to eliminate process variations from the detection problem, [7] proposes to exploit the strong correlation between the max operating frequency of an IC, \( F_{\text{max}} \), and its dynamic power consumption to detect the presence of an HT. If this approach is sound, it faces the difficult problem of measuring \( F_{\text{max}} \) on a SoC. Finally in 2011, the use of Ring Oscillators (RO) has been proposed to detect HTs. For example, [8] has provided an analysis of RO sensitivity to the presence of HT. However, little information is given in this paper about the efficiency of the proposed technique or on the associated silicon overhead. In parallel to this approach, a test solution was also proposed in [6] in order to trigger easily a HT or at least to increase its electrical activity. In 2010, to eliminate process variations from the detection problem, [7] proposes to exploit the strong correlation between the max operating frequency of an IC, \( F_{\text{max}} \), and its dynamic power consumption to detect the presence of an HT. If this approach is sound, it faces the difficult problem of measuring \( F_{\text{max}} \) on a SoC. Finally in 2011, the use of Ring Oscillators (RO) has been proposed to detect HTs. For example, [8] has provided an analysis of RO sensitivity to the presence of HT but concludes that it seems difficult to detect really small HT. At the same time, [9] suggested the use of an array of RO, used in conjunction with PCA [10] to distinguish infected ICs from genuine ones. This proposal has been experimentally validated on FPGA using a Digital Sampling Oscilloscope (DSO) (thus off-chip) to measure the oscillating frequency of ROs. In 2012, the proposed idea was also evaluated for a true ASIC [11]. However, the results present a lower success rate than those reported in [9]. This is may be caused by the use of an embedded counter to measure the oscillating frequency of RO,
which implies a poor accuracy. That point will be discussed later. This is the reason why we conducted a thorough analysis of the use of RO and counters for on-chip hardware Trojan detection.

The rest of this paper is structured as follows. First in Section II, the conducted experiments and the measurement setups used are described. The accuracy of the oscillation frequency of RO is also quantified in this section as well as the impact of intra-die and inter-die variations based on measurements done on several boards. In Section III, the influence of a parasitic electrical activity on the oscillation frequency of ROs is quantified. To that end, an LFSR, whose length could be varied between 4 and 64 bits, is used to mimic the effect of a HT with an electrical activity equivalent corresponding to that of 2 to 32 D Flip-Flops (DFF). Then the influence of two main parameters, the electrical activity amplitude and the physical distance between this activity and the ROs under measurement, is analyzed experimentally. From these experiments the value of the internal voltage drop induced by a parasitic activity is finally inferred. Because all results of section III were obtained without any normal electrical activity in the IC, an un-realistic situation with respect to the operation of an infected IC, all experiments were repeated with an AES performing a ciphering during the measurements. The corresponding results are reported in Section V, before drawing a global conclusion in Section VI.

II. EXPERIMENTAL SETUP

In order to analyze precisely the impact of a parasitic electrical activity on RO’s frequency, a target was designed and implemented on a Xilinx Spartan-3E 1600E FPGA [12] using Xilinx tools. It includes a Finite State Machine (FSM), a serial communication block (RS232) that handles the communication between the computer and the chip and 14 RO that can be enabled and measured separately. The FSM and the RS232 are placed far enough from the ROs so as not to influence them.

The 14 RO are placed as shown Fig. 1 showing the floorplan of the implemented circuit. This placement was adopted to allow a spatial analysis of the impact of an electrical activity on a RO, and more precisely on the supply voltage distribution in the chip. The 14 ROs have the same design composed of 4 inverters and 1 NAND2 gate in order to enable or disable each RO separately. The oscillating frequency \( F_{RO} \) of each RO is around 150 MHz depending on the local quality of the process. To improve the quality of the signal through the FPGA output pin, each RO is connected to a clock divider that allows a clean measurement of \( F_{RO}/2 \) and hence \( F_{RO} \) (corresponding to a period \( T_{RO} \)). The measurements are done using a DSO from Lecroy with a sample rate of 40 GS/s (25 ps between 2 samples). In order to get an accuracy of \( \pm 0.25 \) ps on the \( F_{RO}/2 \) measurements, we measured 100 oscillations of \( F_{RO}/2 \) and we repeated the process 100 times in order to calculate an average value, \( < F_{RO} > \), and a standard deviation \( \sigma_{FRO} \). Each measurement lasts at most 7 s, in order to keep the temperature constant. To ensure a stabilized voltage, a DC power supply with a 0.05% accuracy was used to directly power the FPGA core after the removal of the voltage regulator of the board. To generate a parasitic electrical activity, a Linear Feedback Shift Register (LFSR) was placed all around the RO3 as shown Fig. 1. This RO was therefore considered to be at the coordinates (0,0) during all experiments. This LFSR was designed in such a way that we could change the word size between 4 and 64 bits. This was done to mimic a parasitic electrical activity equivalent to the flipping of 2, 8, 16 and 32 bits in order to analyze the influence of the amplitude of a parasitic electrical activity during the study.

Prior to any analysis, we first quantified the accuracy (including all measurement noise sources: circuit, laboratory, power supply, etc.) of our measurements and the impact of intra-die and inter-die variations. To that end we measured the mean value \( < \Delta T_{ROI} > \) and the standard deviation \( \sigma_{ROI} \) of the different ROs and this was repeated for several boards. Fig. 2 gives the RO periods for two different boards only to ease the reading. It shows that there are significant inter-die and intra-die variations which are responsible of changes in the RO periods of more than 100 ps. However, the most important point drawn from Fig. 2 is that the standard deviation of our measurements (independently of the considered board) varies from one ring oscillator to another on a range between 2.5 and 3.7 ps.

III. IMPACT OF A PARASITIC ELECTRICAL ACTIVITY ON \( < T_{ROI} > \)

The impact of a parasitic electrical activity on the period of a RO is analyzed using the experimental set-up described previously. This analysis is done according to two main parameters. The first one is the amplitude of this electrical activity. The second one is the distance between the RO and the location of the parasitic electrical activity.
A. Amplitude of the parasitic electrical activity

Fig. 3 shows the evolution of the mean, \( \langle T_{RO9} \rangle \), of the period measured from RO9 with respect to the amplitude of the parasitic electrical activity expressed in terms of switching bits (and more precisely switching DFF). Fig. 4 shows the difference, \( \Delta \langle T_{RO9} \rangle \), between the period measured with and without parasitic electrical activity for RO9. As shown, \( \langle T_{RO9} \rangle \) increases linearly from 13.892 ns up to 13.897 ns with the amplitude of the parasitic electrical activity. This linear increase is due to the parasitic electrical activity as highlighted in Fig. 3 which shows that \( \Delta \langle T_{RO9} \rangle \) also increases linearly from 0 to 5 ps, i.e. only 0.043% of the \( \langle T_{RO9} \rangle \). This value is very small compared to the effect of process variations, even intra-die ones, whose effect can reach 500 ps on this board. However, the RO9 is relatively far from the parasitic source. Let us analyze what happens on RO closer from the parasitic electrical activity.

B. Distance between RO and the parasitic electrical activity

To study the behaviour of the other ROs, we analyzed the impact of the distance between the parasitic electrical activity and the ROs on the latter’s frequencies. Fig. 6 shows two cartographies of \( \Delta \langle T_{ROi} \rangle \) obtained with two different Spartan3E-1600 devices for a parasitic equivalent activity equivalent to the switching of 32 DFF (LFSR configured to work on 64 bits). The distance unit of these cartographies is the slice which was estimated to have a height equal to \( \sim 120 \mu m \) according to the number of slices (along X and Y) embedded in these devices and to their dimensions (measured using a microscope after chemical removal of the package).

As shown in Fig. 6, for RO3 placed right in the middle of the parasitic activity, the increase of its period is only of 8 ps. For the farthest ROs (RO1, RO6 and RO14) the influence is of about 5 ps. This value seems to be the minimum effect of the parasitic electrical effect; effect which is distributed over the whole device. This trend is confirmed by Fig. 5 that gives a projection of the two cartographies along Y axis (see Fig. 1 for the orientation). These results suggest that a parasitic electrical activity has a global effect (5 ps) on the power network of the device but also has a slightly more pronounced local effect (8 ps).

Nevertheless, these results also suggest that the effect of the switching of 32 DFF on the RO’s oscillation frequency is extremely small for the ROs (5 ps) furthest from the parasitic electrical activity as for the closest ones (8 ps). This is reassuring since these results reflect the fact that the power/ground networks are designed to be as little resistive as possible in
order to avoid the occurrence of important voltage drops that can compromise the timing constraints. Hence this shows that the induced voltage perturbation is global and with very low amplitudes.

Another interesting point highlighted by these experimental results is that the influence of the parasitic electrical activity on the ROs period ($\Delta < T_{ROI} >$) is relatively constant at a given distance from the activity. This means that its effect seems to be relatively independent of the intra-die variations, and this is confirmed on the Fig. 6. This observation might be the starting point for new Trojan detection methods that would be naturally immune to process variations.

C. Parasitic electrical activity and internal supply voltage

In order to evaluate the effective voltage near each RO, we characterized the sensitivity of a RO to its supply voltage. This was done with a DC power supply with an accuracy of 0.05% and for a voltage range of $[1.19 \text{ V}, 1.2 \text{ V}]$, considering that the circuit’s nominal voltage supply is 1.2 V. Fig. 7 shows the evolution of the mean oscillation period $< T_{ROI} >$ against Vdd, the supply voltage.

As expected, the period decreases linearly ($-14 \text{ps/mV}$) with Vdd on this small voltage range. From there, we made a cartography of the voltage drop induced by a parasitic activity equivalent to the switching of 32 DFF. This cartography is represented in Fig. 8. It shows the maximum voltage drop is reached at the parasitic activity’s source location with a value of 0.5 mV. This voltage drop decreases as we move away from the parasitic activity with the distance with a minimum value of 0.25 mV. These values are far below the values considered, at the design stage, when applying corner analyses during static timing analyses ($\pm 10\%$ of Vdd).

IV. IMPACT OF A PARASITIC ELECTRICAL ACTIVITY ON $T_{ROI}$ IN AN AES

Until now, we experimentally studied the impact of a parasitic electrical activity on the RO’s oscillation period and on the internal supply voltage. This impact can be observed using measurements made with a high sampling rate Digital Storage Oscilloscope (DSO). However, this impact remains low relative to the effects of the intra-die and inter-die variations. This study was conducted on a circuit without normal computing activity and hence does not accurately reflect the reality and in particular the effect of a hardware trojan whose trigger activity is synchronized by a part of the circuit. The previous measurements were repeated with an AES (placed in the neighbourhood of the structure shown Fig. 1) running in parallel. The Advanced Encryption Standard (AES) used is a hardware implementation of the NIST encryption standard specified in [13]. Moreover, the LFSR (as a HT) is clocked with a clock net of an AES register.

Fig. 9 shows the cartographies of $\Delta < T_{ROI} >$, obtained from two different Spartan3E-1600 devices with the AES
running in parallel. The first observation is that the figures with AES (Fig. 9) and those without AES (Fig. 6) are quite similar. As it can be seen, \( \Delta <T_{ROI} > \) varies between \( \sim 5 \)ps and \( \sim 8 \)ps with respect to the distance from the parasitic electrical activity. The average impact of the HT (of the parasitic electrical activity) on the RO is therefore not altered at all by the activity of the AES. This is compliant with the superposition principle in linear systems. However, we noticed that the AES activity increases the dispersion of the \( <T_{ROI} > \) measurements resulting in an increase of the standard deviation \( \sigma_{ROI} \) from \( \sim 3 \)ps to \( \sim 9 \)ps. Consequently, the AES computing activity imposes to do more precise estimations (measurements) of the \( <T_{ROI} > \) values. In practice this can be achieved with more measurements (for example 200 oscillations instead of 100). These considerations prompt to examine the issues of integrated measurements, measurements that are done with a reduced sampling rate 2 GS/s for high speed circuits) compared to those measurements done with modern Digital Storage Oscilloscopes (40 GS/s up to 120 GS/s).

V. DETECTING TROJANS WITH RO AND COUNTERS

The results presented above show that an electrical activity equivalent to an 8-bit (resp. 32-bit) switching activity increases the oscillation period by \( \sim 3 \)ps (resp. \( \sim 8 \)ps) through the induction of a 0.5 mV voltage drop. Because these values are really small, one can question the relevance of the choice of the couple (RO,counter) as an efficient sensor for integrated HT detection. Indeed, regardless of the oscillation period of the RO, the duration of the measurements (of the counting) that allows capturing an oscillation period drift of 5ps would be too high to be realistic. Hence one needs to ensure that measurement conditions are stable and that the measurement duration is compatible with the computing times of the functional blocks embedded in ICs. Also, the measurement time can be high compared to the active period of the HT’s trigger. In order to clarify this point, we estimated the required number of clock cycles (with a period of \( T_{CK} \)) needed to measure a
drift of 3ps. We used the following formula:

\[ n = \frac{p \cdot T_{CK}}{\Delta T_{RO}} \]  \hspace{1cm} (1)

with \( p \) being the difference between the number of rising edges counted in the presence of the HT and that counted without the HT, this value must be high enough regarding the fluctuations induced by the measurement noise. In our case for \( T_{CK} = 20\,\text{ns} \) (the FPGA design runs at 50 MHz) and \( \Delta T_{RO} = 3\,\text{ps} \) we obtain: \( n = 6667 \). The counter must be enabled during 6667 clock cycles (resp. 666700) to measure a counter difference of 1 (resp. 100). This represents a measurement time of 133\,\mu s (resp. 13.3 ms) for one RO. During this period the experimental conditions have to remain stable. Based on such values, it seems hard to use the couple (RO,counter) to make an integrated method for HT detection. It appears that we need reliable sensors with a much more important sensitivity to local variations of the supply voltage for efficient and embedded HT detection techniques.

VI. CONCLUSION

This paper describes experiments conducted in order to answer two questions. The first is relative to the effect a HT on the ICs internal voltage and delays. The second is relative to the use of the couple (RO,counter), as a tool for embedded HT detection. With regard to the first question the experimental results show that the effect of Hardware Trojans on the supply voltage and delays are significantly below the effect of process variations. As a result, concerning the second question, too-high on-chip sampling rate values, or too long measurement times, are therefore required in order to capture the timing degradation induced by HT using RO. The couple (RO,counter) is not an adequate choice for embedded Hardware Trojan detection. ROs can be used if their frequencies are measured using off-chip DSOs. For completely on-chip solutions, sensors with higher sensitivity to supply voltage variations must hence be sought.

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