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C³APSULe: Cross-FPGA Covert-Channel Attacks through Power Supply Unit Leakage

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Abstract—Field-Programmable Gate Arrays (FPGAs) are versatile, reconfigurable integrated circuits that can be used as hardware accelerators to process highly-sensitive data. Leaking this data and associated cryptographic keys, however, can undermine a system’s security. To prevent potentially unintentional interactions that could break separation of privilege between different data center tenants, FPGAs in cloud environments are currently dedicated on a per-user basis. Nevertheless, while the FPGAs themselves are not shared among different users, other parts of the data center infrastructure are. This paper specifically shows for the first time that powering FPGAs, CPUs, and GPUs through the same power supply unit (PSU) can be exploited in FPGA-to-FPGA, CPU-to-FPGA, and GPU-to-FPGA covert channels between independent boards. These covert channels can operate remotely, without the need for physical access to, or modifications of, the boards. To demonstrate the attacks, this paper uses a novel combination of “sensing” and “stressing” ring oscillators as receivers on the sink FPGA. Further, ring oscillators are used as transmitters on the source FPGA. The transmitting and receiving circuits are used to determine the presence of the leakage on off-the-shelf Xilinx boards containing Artix 7 and Kintex 7 FPGA chips. Experiments are conducted with PSUs by two vendors, as well as CPUs and GPUs of different generations. Moreover, different sizes and types of ring oscillators are also tested. In addition, this work discusses potential countermeasures to mitigate the impact of the cross-board leakage. The results of this paper highlight the dangers of shared power supply units in local and cloud FPGAs, and therefore a fundamental need to re-think FPGA security for shared infrastructures.

Index Terms—Power supply units, voltage regulators, ring oscillators, FPGAs, covert channels, power attacks

I. INTRODUCTION

Field-Programmable Gate Arrays (FPGAs) implement reconfigurable hardware which can speed up application-specific tasks. For example, they are used to process highly-sensitive data in cloud applications including genomic sequencing, cryptography, and financial modeling, among others [2]. Given the increasing proliferation of cloud FPGAs [44], interest in FPGA attacks has also risen. For example, prior work has shown that implementations of cryptographic and other algorithms are susceptible to reverse engineering [39], [46] as well as side-channel attacks [46]. As such research requires physical proximity to the FPGA board and external equipment (e.g., high-end oscilloscopes) to extract keys, it does not reflect modern-day environments, where FPGAs remain inaccessible in data center server racks. When considering cloud FPGAs, however, attacks must instead be performed remotely using only on-chip logic. Recent work has so far shown that remote fault, covert-channel, and side-channel attacks are indeed possible between designs belonging to different users co-located within the same FPGA chip [8]–[11], [24], [26], [28]–[31], [47]. However, as boards are currently allocated on a per-user basis in commercial clouds, this multi-tenant threat model remains theoretical, with little practical impact.

In this paper we instead tackle a more pressing scenario that is applicable to existing cloud FPGA deployments, where boards are co-located within the same server rack. Users renting FPGAs from such FPGA cloud providers assume that their designs are safely isolated from potentially malicious designs by other users running in the same data center. However, as we show in this paper, the assumption of isolation can be broken due to leakage through the shared use of power supply units (PSUs). Specifically, we introduce a new class of remote covert-channel attacks between single-tenant FPGAs on different FPGA boards that are merely powered through the same PSU. Moreover, we show that if this PSU also powers the host computer, the same sink FPGA (receiver) can detect high levels of CPU and GPU activity, creating new CPU-to-FPGA and GPU-to-FPGA channels.

The first crucial observation of our work is that although causing variable power consumption to transmit information is easy, detecting voltage fluctuations without external equipment is non-trivial. However, the reconfigurability of FPGAs provides access to the hardware at a much lower level, and can be used to implement circuits that detect voltage changes which are imperceptible to fixed silicon chips such as CPUs and GPUs. Indeed, cloud providers are aware of the impact of such low-level hardware access, so besides allocating FPGAs on a per-user basis, they also keep several features such as voltage and temperature monitors inaccessible to end-users.

The second key observation is that ring oscillators (ROs) are capable of both causing and sensing voltage fluctuations. This paper therefore introduces a novel way of monitoring changes in voltage caused by the source FPGA, CPU, or GPU. Specifically, both properties of ROs are used in the sink (receiver) FPGA, whereby stressing the voltage regulator of the sink FPGA allows one to detect transmissions by the source (transmitter) FPGA.
Using these insights, we demonstrate the first cross-FPGA covert channel between off-the-shelf, unmodified Xilinx Artix 7 and Kintex 7 boards in either direction of communication. We also characterize the bandwidth-accuracy tradeoffs across different measurement periods and sizes of the covert-channel ROs on the source and sink FPGAs. We further test our covert channel on two PSUs running under normal operating conditions (i.e., without being overloaded), and introduce CPU-to-FPGA and GPU-to-FPGA covert channels by modulating their respective loads. We finally discuss potential countermeasures to mitigate the effects of this leakage.

A. Contributions

Our contributions can be summarized as follows:

1) We identify sharing of PSUs as a new source of vulnerability, even for unprivileged FPGA designs without access to voltage or temperature system monitors.

2) We introduce a novel measurement setup and classification metric that uses ring oscillators (ROs) on the sink FPGA to stress its voltage regulator and therefore reliably detect external voltage fluctuations.

3) We exploit this setup to create the first remote covert-channel attack between FPGAs on distinct physical boards that are dedicated on a per-user basis, reaching accuracies of up to 100%.

4) We evaluate the strength of the information leakage across different architectural choices, and perform a bandwidth-accuracy tradeoff analysis.

5) We introduce the first CPU-to-FPGA and GPU-to-FPGA covert channels using high loads of activity on their respective processors, opening up new avenues for remote FPGA attacks.

6) We propose hardware- and software-level countermeasures to reduce the impact of the leakage.

B. Paper Organization

The rest of the paper is organized as follows. Section II introduces the threat model, while Section III details the experimental setup, including hardware properties, the measurement procedure, and the high-level architectural FPGA design. Section IV then describes the need for our novel classification metric, and explains why it works where the naive approach of looking at absolute ring oscillator counts fails. Section V then evaluates cross-FPGA covert communication over shared PSUs, varying the number of source and sink ring oscillators used, and performing an analysis of bandwidth-accuracy tradeoffs. Section VI then covers CPU-to-FPGA and GPU-to-FPGA information leakage, while Section VII discusses potential defense mechanisms. We place our work in the context of related research in Section VIII, before we conclude in Section IX.

II. Threat Model

Prior work on remote FPGA attacks has primarily investigated security in the context of multi-tenant FPGAs. It has shown that when a single FPGA chip is shared among multiple users concurrently, designs are vulnerable to temperature and voltage attacks (Section VIII). Although these attacks highlight potential issues with future architectures, they remain theoretical at the moment, as FPGAs are currently allocated on a per-user basis.

In this paper, we are thus concerned with remote attacks against platforms where the entire logic is allocated to a single user, such as FPGAs on Amazon Web Services, Alibaba Cloud, etc. [44]. Design logic therefore cannot access any voltage or thermal system monitors present on the FPGA fabric, as these inaccessible in a cloud environment.\(^1\) Compared to multi-tenant attacks on FPGA designs which share the same power distribution network, this presents a much stronger threat model, which necessitates that any effect caused by a side- or covert-channel transmitter be measurable across extensive physical separation (as opposed to logic on the same FPGA chip), and with multiple intermediate components (passive capacitors, inductors, voltage regulators, etc.) on the path between the source and sink FPGA boards.

In our work, we specifically investigate remote voltage-based attacks, where a shared PSU provides an indirect (and, as we show, leaky) connection between FPGA boards. We do not consider reverse-engineering attacks on the bitstream itself or the contained logic, but instead focus on how to initiate a communication channel through modulating the load on the PSU itself. We mainly consider FPGA-to-FPGA attacks between otherwise unconnected devices, but also investigate CPU-to-FPGA and GPU-to-FPGA attacks. This is because the same PSU might also power the host computer, and, by extension, its internal components including CPUs and GPUs, as shown in the high-level system model of Figure 1. We make no assumptions regarding how the FPGAs are connected to the computer. In other words, we do not assume that FPGAs are attached to the motherboard over PCIe, to a USB controller over a serial chip, or, in fact, if they are even (logically) connected to the computer at all. Our only assumption is that

\(^1\)In cloud FPGAs, part of the reconfigurable fabric is reserved by a cloud-provided “shell” which hides implementation details, including physical pinouts, identification primitives, and system monitors. User logic is therefore forced to interact with external hardware through the shell’s AXI4 interfaces.
of a shared PSU between the two communicating parties.

Within an FPGA, and in accordance with prior work, (potentially adversarial) users can place and route any designs of their choice, such as different types of ring oscillators. This is allowed by current FPGA cloud deployments, as long as the logic is placed outside of the cloud-provided shell. In this paper, we show that by relying only on on-chip FPGA logic (i.e., ring oscillators), we are able to demonstrate FPGA-to-FPGA, CPU-to-FPGA, and GPU-to-FPGA covert communication, without physical access to the FPGA boards. One of the key contributions of our work is therefore the ability to communicate across unmodified devices, without external equipment or access to internal voltage monitors, which are off-limits to unprivileged FPGA designs.

It should be noted that some cloud providers such as Amazon Web Services (AWS) place restrictions on the types of circuits that can be instantiated on their FPGAs, and prohibit combinatorial loops including ring oscillators. Although in this work we primarily use conventional ring oscillators, Section V-E shows that they can be easily replaced by alternate designs proposed in recent work [8], [10], [11], [34], which bypass cloud countermeasures.

III. EXPERIMENTAL SETUP

In this section, we detail our experimental setup, starting with the ring oscillators employed in the source and sink FPGAs (Section III-A), and delving into the architectural design of the FPGA transmission and reception circuitry (Section III-B). We then describe the hardware properties of the FPGAs used (Section III-C), as well as the computer PSUs, CPUs, and GPUs, which are effectively turned into covert-channel transmitters (Section III-D). We finally discuss the process followed for data collection (Section III-E).

A. Ring Oscillators

Ring Oscillators (ROs) are comprised of an odd number of NOT gates in a ring formation, and therefore form a combinatorial loop, whose value oscillates. The frequency of oscillation changes based on process variations, as well as voltage and temperature conditions [18], making ROs good temperature [35], [37] and voltage [47] monitors. ROs also cause voltage fluctuations, which stress power circuits, and can potentially crash the FPGA or inject faults [14], [24].

In this paper, we use ROs as both transmitters and receivers, and implement them using lookup tables (LUT-RO) with one inverter and three buffer stages as shown in Figure 2. We chose to use this RO design instead of more common ROs with three inverters or one inverter and two buffer stages [9], [10], [35] because preliminary experiments showed that they resulted in more stable measurements. Alternative types of ROs are evaluated in Section V-E.

B. Architectural FPGA Design

We now give a high-level overview of the covert-channel source and sink FPGA designs.

1) Covert-Channel Source: To cause detectable changes on the sink, the source FPGA employs ring oscillators organized as \( T \cdot N_T \) ROs, while the sink (right) uses \( R \cdot N_R \) measurement ROs and \( S \cdot N_S \) stressor ROs. The same power supply unit powers both boards.

2) Covert-Channel Sink: To receive transmissions, we employ \( R \) receivers, placed on separate clock regions of the sink FPGA, and each containing \( N_R \) ROs. We estimate the RO frequency by counting the number of RO signal transitions in a fixed measurement interval of \( 2^t \) clock cycles through counters placed outside of the RO clock regions.

However, this setup is not sufficient to decode covert transmissions, due to inherent noise in the power supply and environmental fluctuations. Instead, it is necessary to introduce additional circuitry on the sink FPGA which stresses the board’s voltage regulator, making maintaining a constant voltage harder. This fact allows us to sense voltage changes induced by the source FPGA, or even by CPU and GPU activity, as presented later in Section VI. Specifically, we include \( S \) stressors, each with \( N_S \) ROs. As with the source transmitters, these \( S \) stressors are placed on separate clock regions, and can also be controlled independently. The block diagram for the sink design is shown in the right part of Figure 3, while Figure 4 shows a concrete instantiation of the sink architecture on the Kintex 7 board. Section IV further demonstrates the need for stressor ROs.

C. FPGA Boards

For our experiments, we use Xilinx Kintex 7 KC705 and Artix 7 AC701 boards. The 28 nm chips these devices contain are similar, but the Kintex 7 is more performant, while the
Artix 7 is optimized for low power [41], [42]. Both FPGAs have a 200 MHz oscillator and operate at a core VCCINT voltage of 1.0 V, but the boards use different regulators to convert the 12 V PSU output to 1.0 V [43], [45].

For the source FPGA designs, we place a transmitter on each clock region of the FPGA. As the Artix 7 board has 10 clock regions, while the Kintex 7 has 14, the number of transmitters on these devices is \( T = 10 \) and \( T = 14 \) respectively. The sink FPGAs contain \( R = 4 \) receivers in the corners of each chip, each with \( N_R = 5 \) ROs. Sink FPGAs also contain \( S = 5 \) stressors, one of which is placed in the center of the device, while the remaining four next to the receiver clock regions (Figure 4 shows an example with \( N_S = 500 \)). Although not shown to be significant in our experiments, these early architectural choices were made to ensure that the power draw was approximately equally spread across the power distribution network of the FPGA fabric.

These decisions and other FPGA properties are summarized in Table I. More compile- and run-time parameters, such as the measurement period and the number of source transmitters ROs \( N_T \) and sink stressor ROs \( N_S \), are varied in Section V.

### D. Power Supply Units & Computer Transmitters

To verify that the covert channel is not due to faulty design in a line of specific power supply units, we test communication on two PSUs made by different manufacturers (Corsair and Dell), rated for different loads (850 W and 1,300 W respectively), and both with a Gold 80 Plus Certification (which guarantees 90% efficiency at 50% load). These PSUs are integrated in two computers, the first of which contains two Xeon E5645 CPUs for a total of 24 threads, while the second a single Xeon E5-2609 with 4 threads. They also contain Nvidia GeForce GPUs, with 96 and 640 CUDA cores respectively. The CPU and GPU cores are used as the covert-channel sources in Section VI for CPU-to-FPGA and GPU-to-FPGA communication over the shared power supply. The properties of the computers used are summarized in Table II.

#### Table I: Properties of the FPGA boards used, along with fixed compile-time choices for the source and sink circuit configurations.

<table>
<thead>
<tr>
<th>Property</th>
<th>Artix 7</th>
<th>Kintex 7</th>
</tr>
</thead>
<tbody>
<tr>
<td>Board</td>
<td>AC701</td>
<td>KC705</td>
</tr>
<tr>
<td>Boards Tested</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>Part Number</td>
<td>XC7A200T</td>
<td>XC7K256T</td>
</tr>
<tr>
<td>Slices</td>
<td>33,650</td>
<td>50,950</td>
</tr>
<tr>
<td>Clock Regions</td>
<td>( 2 \times 5 )</td>
<td>( 2 \times 7 )</td>
</tr>
<tr>
<td>Core Voltage, VCCINT</td>
<td>1.0 V</td>
<td>1.0 V</td>
</tr>
<tr>
<td>Voltage Regulator</td>
<td>LMZ31710</td>
<td>PTD08A020W</td>
</tr>
<tr>
<td>Clock Frequency</td>
<td>200 MHz</td>
<td>200 MHz</td>
</tr>
<tr>
<td>Transmitters, ( T )</td>
<td>10</td>
<td>14</td>
</tr>
<tr>
<td>Stressors, ( S )</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>Receivers, ( R )</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>Receiver ROs, ( N_R )</td>
<td>5</td>
<td>5</td>
</tr>
</tbody>
</table>

#### Table II: Hardware properties of the two computers used, with their corresponding PSUs, CPUs, and GPUs.

<table>
<thead>
<tr>
<th>Property</th>
<th>PC-A</th>
<th>PC-B</th>
</tr>
</thead>
<tbody>
<tr>
<td>PSU Brand</td>
<td>Corsair</td>
<td>Dell</td>
</tr>
<tr>
<td>Power Rating</td>
<td>850 W</td>
<td>1,300 W</td>
</tr>
<tr>
<td>80 Plus Certification</td>
<td>Gold</td>
<td>Gold</td>
</tr>
<tr>
<td>Motherboard</td>
<td>SuperMicro</td>
<td>Dell Precision</td>
</tr>
<tr>
<td>X8DAL-i</td>
<td>E5645</td>
<td>E5-2609</td>
</tr>
<tr>
<td>Xeon CPU Model</td>
<td>E5645</td>
<td>E5-2609</td>
</tr>
<tr>
<td>CPU Cores</td>
<td>6 @ 2.4 GHz</td>
<td>4 @ 2.4 GHz</td>
</tr>
<tr>
<td>Number of Threads</td>
<td>12</td>
<td>4</td>
</tr>
<tr>
<td>Number of CPUs</td>
<td>2</td>
<td>1</td>
</tr>
<tr>
<td>GeForce GPU</td>
<td>ZOTAC GT 430</td>
<td>EVGA GTX 750 Ti</td>
</tr>
<tr>
<td>GPU Memory</td>
<td>1 GB GDDR3</td>
<td>2 GB GDDR5</td>
</tr>
<tr>
<td>CUDA Cores</td>
<td>96 @ 0.7 GHz</td>
<td>640 @ 1.0 GHz</td>
</tr>
</tbody>
</table>

#### E. Data Collection and Encoding

For our data collection process, we made several choices to make the communication scenario realistic. For instance, the computers attached to the PSUs were used normally during experimentation, including running and installing other software. Moreover, to ensure leakage is not due to temperature, the FPGAs were placed outside the computer case, and away from computer fans, which may affect measurements by turning on or off based on the computer temperature. We similarly placed the FPGAs next to each other horizontally (as opposed to stacking them vertically), further minimizing cross-FPGA temperature effects. In addition, to control for other voltage effects, the FPGAs were not connected to the computer over PCIe, which would likely increase the potential for leakage. However, as we show in Section V-E, our covert channel operates with similar accuracy, even when the FPGAs are connected to the computer over PCIe, and are enclosed in it without accounting for temperature variations. Finally, to verify that the leakage is not caused through the UART interface, we often used one computer to take the measurements, and the other to power the source and sink boards through its PSU.

As there is inherent noise in the measurements, (a) the absolute RO frequency is not well-suited for comparison, and (b) the RO counts need to be averaged over repeated measurements to produce meaningful results. To address both concerns, we use Manchester encoding, where to send a 1, the
source transmitters are enabled for one measurement period and disabled for the next (a 0 is similarly encoded by first disabling transmitters during the first measurement period and enabling them in the second period). These measurement periods are \( M \cdot 2^i \) clock cycles long, where we average \( M \) RO counts collected by ROs enabled for \( 2^i \) clock cycles (see Section IV). The bandwidth can thus be calculated as:

\[
b = \frac{f_c}{2 \cdot 2^i \cdot M}
\]

where \( f_c = 200 \text{ MHz} \) is the FPGA clock frequency.

In most experiments, we transmit the 20-bit number \( 0\text{xf3ed1} \), which is Manchester-encoded in 40 bits to be sent across the covert channel. Additional patterns are evaluated in Section V-E. To ensure that perfect synchronization is not needed between the source and the sink, for each of the 40 periods, we take four sets of \( M \) measurements, where \( M \) is in the order of a few hundred counts (see Table III and Section V-D). The four sets of repetitions create \( 4^2 = 16 \) Manchester encoded pairs per bit to be transferred, for a total of \( 16 \times 20 = 320 \) pairs to estimate the covert-channel accuracy.

IV. Classification Metric

This section introduces a novel methodology to detect changes in the power supply voltage through the sink’s “stressor” ROs. Section IV-A first motivates why the naive approach of using the absolute ring oscillator counts is insufficient for classification of transmissions in this scenario. Section IV-B then introduces the metric using stressors, while Section IV-C finally explains why our technique works.

A. Motivation

Broadly speaking, when the transmitters are activated on the source FPGA, there is a voltage drop that is visible not just at the board regulator, but also at the 12 V rail PSU input to the FPGA. Indeed, Figure 5 demonstrates this for a Kintex 7 source without a sink FPGA present across multiple input voltages and different numbers of enabled transmitters \( T \). Specifically, we power the board using a Keithley 2231A power supply, and measure the voltage at the power rail of the board using a Tektronix MDO3104 Mixed Domain Oscilloscope with TPP1000 1 GHz passive probes, taking 10,000 data points. Figure 5 indicates that at any voltage level provided by the power supply (11.5-12.5 V), as the number of enabled source transmitters \( T \) increases, the voltage measured by the oscilloscope decreases. For example, at 12.5 V, the oscilloscope measures 12.539 V when no transmitters are enabled, but only 12.521 V when 14 transmitters are enabled, for a voltage drop of approximately 18 mV. At 11.5 V, the measured voltage similarly drops from 11.525 V to 11.507 V.

Although one would expect RO frequency to increase with higher voltages [18], this is not the case. For a ring oscillator \( i \), let its average count be \( C_i^v \). When the voltage provided by the power supply is \( 11.5 \text{ V} \leq V \leq 12.5 \text{ V} \), we would expect that \( C_{i_1}^v > C_{i_2}^v \) whenever \( V_1 > V_2 \), but Figure 6 suggests that the RO counts remain approximately the same for all eight ring oscillators and voltages \( V \) tested on an Artix 7 sink. As a result, the absolute RO frequency cannot be used to decode cross-FPGA covert-channel transmissions.

B. Description

To solve the issues identified above, we introduce ROs to “stress” the voltage regulator and make external changes in the power supply voltage measurable. For any bit transmission (say the \( i \)-th one), we take \( M \) measurements as follows:

1. For the first measurement period, we disable all stressor ROs, and let the receiver ROs run for \( 2^i \) clock cycles, producing counts \( C'_0 = (C'_0, \ldots, C'_R \cdot N_R^{-1}) \).
2. In the second measurement period, we enable all (or some, see Sections IV-C and V-D) stressor ROs, and estimate the RO frequencies through their counts, \( C^r_i \).
3) In the third measurement period, we disable all stressor ROs, re-enable them in the fourth period, and so forth.

This procedure produces \(M/2\) measurements \(C_0, C_1, \ldots\) corresponding to disabled stressors, and \(M/2\) measurements \(C'_1, C'_2, \ldots\) corresponding to enabled stressors, as also shown in the timing diagram of Figure 7. Figure 7 represents Manchester-encoded transmissions of the 3 bits \(100\), averaging over \(M = 4\) measurements and only repeating transmissions once (actual measurements have \(M = 500\), with 4 repetitions). We take the average of each set per RO, thereby calculating the disabled-stressor average \(D^i = 2/M \cdot \sum_{k=0}^{M/2-1} C_{2k}\) and the enabled-stressor average \(E^i = 2/M \cdot \sum_{k=0}^{M/2-1} C'_{2k+1}\). We then use \(\Delta^i = D^i - E^i\) to recover the transmitted bit.

Specifically, assume that we wish to recover the \(n\)-th bit, corresponding to transmissions \(2n\) and \(2n+1\), as each bit \(b\) is Manchester-encoded as the pair \((b, 1-b)\). In each transmission pair, there is always a 1-bit and a 0-bit, so we can compare the \(R \cdot N_{RO}\) counts of \(\Delta^{2n}\) and \(\Delta^{2n+1}\). If the majority of the RO differences in the first set of measurements is bigger than the corresponding differences in the second set of measurements (i.e., \(\Delta^{2n} > \Delta^{2n+1}\) for most ROs), we classify the \(n\)-th bit as a 1, while if the majority is smaller, we classify it as a 0.

Figure 8 demonstrates the need for this more complicated procedure in practice for a transmission of a Manchester-encoded 1-bit. Specifically, it compares our new metric with stressor ROs, \(\Delta^{2n} - \Delta^{2n+1}\), against the naive bit-recovery metric \(D^{2n} - D^{2n+1}\) for all 20 receiver ROs. As Figure 8 (blue circles) shows, \(\Delta^{2n} - \Delta^{2n+1} > 0\) for all 20 receiver ROs \(R_0, R_1, \ldots\), so our metric correctly recovers this bit transmission. However, the \(D^{2n} - D^{2n+1}\) values with stressors disabled (orange diamonds) behave randomly, and indeed, in the experiment in which these measurements originated, our metric successfully recovered over 98% of transmissions, compared to 53% using the naive method without the stressors. Section IV-C further expands on why the new technique makes for a good approach in detecting transmissions.

C. Explanation

In this section we test the receiving circuit (sink FPGA) on its own to characterize its behavior. We first plot in Figure 9 the average metric \(\Delta^i\) for the eight ring oscillators of Figure 6 across the same power supply voltages \(11.5\ \text{V} \leq V \leq 12.5\ \text{V}\).
Fig. 10: Difference between the average $\Delta$ metric as measured at 11.5 V and 12.5 V for different measurement times and numbers of stressors enabled on the AC701-1 sink.

As expected, for all ROs, $\Delta_{12.5}^i < \Delta_{11.5}^i$ whenever $V_i > V_2$: when there is an external voltage drop (e.g., when the source FPGA enables the transmitter ROs), the $\Delta$ metric increases compared to when there are no external transmissions.

We additionally test the behavior of the receiver FPGA across different measurement times of $2^n$ clock cycles and numbers of enabled stressors $S$. Specifically, we conduct measurements on an Artix 7 sink, and calculate the average value of our $\Delta$ metric over all 20 receiver ROs at two voltage levels: 11.5 V and 12.5 V. Figure 10 plots our results, which lead to several observations.

First of all, the average difference $\Delta = \Delta_{11.5} - \Delta_{12.5}$ is close to zero for time periods up to 41 $\mu$s, indicating that prolonged measurement times are necessary to distinguish between transmissions of zero and one, which in practice result in much smaller voltage drops of $\approx 20$ mV. Moreover, until 2.6 ms, $\Delta > 0$ for all choices of how many stressors $S$ to enable simultaneously, with fewer stressors resulting in a larger effect. However, for even larger time periods, $\Delta < 0$, with more stressors resulting in a bigger effect in magnitude. Consequently, the choice of number of stressors and measurement time is intricately linked with the accuracy of the covert channel, and, in fact, helps explain why in some experimental setups (e.g., the KC705-1 receiver on PSU-B of Table III), the recovered pattern is flipped, i.e., a 0-bit is identified as a 1-bit and vice-versa.

V. CROSS-FPGA COMMUNICATION

In this section, we explore FPGA-to-FPGA covert communication, presenting a summary of our results with the default experimental parameters in Section V-A. We then vary the number of source transmitter ROs in Section V-B and the number of sink stressor ROs in Section V-C. We further evaluate bandwidth-accuracy tradeoffs in Section V-D, and the performance of the covert channel across different patterns, types of ROs, and measurement setups in Section V-E.

TABLE III: Default values for accuracy- and bandwidth-related parameters, and the sections of the paper in which they are varied. Bandwidth is calculated using Equation (1).

<table>
<thead>
<tr>
<th>Property</th>
<th>Artix 7</th>
<th>Kintex 7</th>
<th>Section</th>
</tr>
</thead>
<tbody>
<tr>
<td>Transmitter ROs, $N_T$</td>
<td>1,000</td>
<td>1,000</td>
<td>V-B</td>
</tr>
<tr>
<td>Enabled Transmitters</td>
<td>10</td>
<td>14</td>
<td>V-B</td>
</tr>
<tr>
<td>Transmitted Pattern</td>
<td>0xf3ed1</td>
<td>0xf3ed1</td>
<td>V-E</td>
</tr>
<tr>
<td>Transmitter Types</td>
<td>LUT-RO</td>
<td>LUT-RO</td>
<td>V-E</td>
</tr>
<tr>
<td>Stressor ROs, $N_S$</td>
<td>500</td>
<td>500</td>
<td>V-C</td>
</tr>
<tr>
<td>Enabled Stressors</td>
<td>5</td>
<td>5</td>
<td>V-D</td>
</tr>
<tr>
<td>Stressor &amp; Receiver Types</td>
<td>LUT-RO</td>
<td>LUT-RO</td>
<td>V-E</td>
</tr>
<tr>
<td>Measurement Cycles, $2^t$</td>
<td>$2^{15}$</td>
<td>$2^{21}$</td>
<td>V-D</td>
</tr>
<tr>
<td>Repetitions per Bit, $M$</td>
<td>500</td>
<td>500</td>
<td>V-D</td>
</tr>
<tr>
<td>Channel Bandwidth $b$ (bps)</td>
<td>6.1</td>
<td>0.1</td>
<td>V-D</td>
</tr>
</tbody>
</table>

TABLE IV: Accuracy for cross-FPGA covert channels on PSUs A and B, using the default experimental parameters. † signifies that the recovered bit-pattern is flipped.

A. Overview of Results

In this section, we give an overview of our cross-FPGA results. The values for the default experimental parameters used in these experiments and the corresponding covert-channel bandwidths are summarized in Table III. These values were chosen based on exploratory testing, as they represent a good tradeoff between accuracy and bandwidth. However, in some cases, better accuracy can be achieved at the cost of bandwidth, or the same accuracy can be maintained despite increasing the bandwidth (see Section V-D).

The results of our measurements across all 12 combinations of source and sink FGPAis on both PSUs are summarized in Table IV. As the table shows, covert communication is possible with high accuracy between any two boards, in either direction, and on both PSUs. The table also allows us to draw various conclusions. First of all, the behavior is not the same for identical boards. This is likely due to both process variations internal to the FPGA chip (which affect RO measurements), and because of different component tolerances. As an example, the AC701-2 board is a worse sink than the AC701-1 board, while the KC705-1 board is a worse source than the KC705-2 board.

Moreover, the Kintex 7 boards are generally better sources than the Artix 7 boards, due to the higher count of transmitters they contain ($T = 14$ as opposed to $T = 10$). As we show in Section V-B, more transmitters tend to improve the quality of the covert channel. Finally, we notice that although the information leakage remains strong in both PSUs, the accuracy
Fig. 11: Increasing the number of simultaneously-enabled transmitters on the KC705-1 source board generally increases the accuracy of the cross-FPGA covert channel.

Fig. 12: Increasing the number of transmitter ROs $N_T$ on the KC705-1 source board also generally increases accuracy.

Fig. 13: Increasing the number of stressor ROs $N_S$ on the AC701-2 sink board can decrease accuracy, as the additional activity can hide external transmissions under the noise floor.

$T \cdot N_T$ increases the accuracy of the covert channel. This is because the ensuing voltage drops are more pronounced, and can thus be more easily detected by the receiving boards. However, for the KC705-2 sink board, too much activity on the transmitter can decrease the accuracy of the channel. This is because although the magnitude of the voltage drop increases in isolation (Figure 5), the stressor ROs are also causing a voltage drop that can overshadow that of the source FPGA.

C. Stressor ROs

Although we keep the number of stressors fixed at $S = 5$, in this section we evaluate the effect of changing the number of stressor ROs $N_S$ on the sink AC701-2 board, and plot the accuracy of the covert channel in Figure 13. Consistently with Figure 10, although stressor ROs are necessary to detect covert transmissions, further increasing $N_S$ can have the opposite effect: the voltage drop caused by the stressors overpowers any effect caused by the source transmissions, and starts pushing the average difference from positive to negative.

D. Bandwidth-Accuracy Tradeoffs

In this section we investigate accuracy-bandwidth tradeoffs by varying both the measurement period of $2^t$ clock cycles and the number of measurements $M$ over which the RO counts are averaged. We first experiment with both the AC701-1 and the AC701-2 boards as sinks, and plot the results from all other possible FPGA sources in Figure 14. In general, increasing the number of measurements increases the accuracy of the covert channel, but at a cost of lower bandwidth. $M = 500$ represents a good tradeoff between accuracy and bandwidth (over 90% accuracy at 6.1 bps for the Artix 7 boards), but $M \geq 1000$ results in higher accuracy at half the bandwidth.

The second aspect we investigate is varying the number of clock cycles $2^t$ for which each RO is counting. At the same time, we also change the number of enabled stressors on the sink FPGA, and test the accuracy of the covert channel with
Fig. 14: Increasing the number of measurements $M$ improves accuracy to any AC701 sink $R$, from any FPGA source $T$.

Fig. 15: Accuracy for different measurement times and number of enabled stressors on the KC705-1 sink.

Fig. 16: Accuracy for different measurement times and number of enabled stressors on the AC701-1 sink.

The AC701-2 FPGA source. The results for the KC705-1 and AC701-1 sinks are shown in Figures 15 and 16 respectively. These results indicate that the parameters for the receivers need to be carefully tuned for different types of boards. For example, the Artix 7 board necessitates that fewer stressors be driven, which is consistent with the results of Sections IV-C and V-C. On the other hand, the KC705-1 sink remains accurate across a wider range of enabled stressors, but requires longer measurement periods for acceptable accuracies.

E. Other Parameters

We finally test the accuracy of the covert channel when varying the patterns transmitted, changing the cable layout, and using different types of ROs in more realistic conditions.

1) Transmitted Patterns: We test the transmission of longer patterns by communicating five 32-bit patterns (64 encoded bits). The patterns were chosen to have different Hamming Weights and runs of zeros and ones to show that the channel does not fundamentally depend on the values transmitted. The results, plotted in Figure 17 for the AC701-2 source, indicate that the covert channel remains similarly accurate for all three sink boards and five transmitted patterns.

2) Measurement Layouts: In the majority of the previous experiments, the source and sink FPGA boards were connected to the same PSU output through a Corsair peripheral cable with four Molex connectors. This cable was attached to one of the “bottom” 6-pin outputs of the PSU. However, to verify that the information leakage persists across different cable setups, we also use a 12-pin output of the PSU splitting into two 6-pin PCIe cables, denoted by “left” and “right”. We then test communication from the KC705-1 board to the KC705-2 board across different cable setups, using the default measurement time of $2^{21}$ clock cycles, enabling all 5 stressors, but also increasing the number of measurements to $M = 1,000$. The results of our experiments are summarized in Figure 18, which demonstrates that a covert channel is
3) Ring Oscillator Types \& Realistic Conditions: We finally test communication using alternative types of ROs on the Kintex 7 boards, which we measure in a more realistic setup. Specifically, both boards are connected to PC-A over PCIe and are enclosed in the computer tower to avoid isolating thermal effects. The ROs used were proposed by Giechaskiel et al. [10], [11] to bypass currently-deployed cloud countermeasures that prohibit combinatorial loops such as the LUT-RO used so far. One of them replaces a buffer gate with a latch (LD-RO), while the other one with an inverter and a flip-flop (FF-RO). The setup otherwise uses the default experimental parameters of Table III. Figure 19 first shows that for all three types of transmitter ROs, the accuracy of the cross-KC705 channel remains above 95%, despite potential noise introduced by thermal conditions and the shared PCIe buses. Similarly, Figure 20 shows that accuracy remains above 95% when using these alternative ROs for stressors and receivers on a KC705 sink. Although in many cases, bits are again flipped, blocking combinatorial loops and introducing environmental noise cannot prevent our channel from operating.

Possible in all setups tested. This is perhaps to be expected, since the PSU uses a “dedicated single +12 V rail” [5], but the results further indicate that there are differences among the ports tested. Specifically, the covert channel is most accurate between FPGA boards on the same cable (as they are at exactly the same electric potential difference) and least accurate between the single location on the bottom of the PSU and either of the dual outputs. Finally, it should be noted that the recovered pattern is flipped in all setups, except when sharing the cable on the bottom output.

6) Ring Oscillator Types \& Realistic Conditions: We finally test communication using alternative types of ROs on the Kintex 7 boards, which we measure in a more realistic setup. Specifically, both boards are connected to PC-A over PCIe and are enclosed in the computer tower to avoid isolating thermal effects. The ROs used were proposed by Giechaskiel et al. [10], [11] to bypass currently-deployed cloud countermeasures that prohibit combinatorial loops such as the LUT-RO used so far. One of them replaces a buffer gate with a latch (LD-RO), while the other one with an inverter and a flip-flop (FF-RO). The setup otherwise uses the default experimental parameters of Table III. Figure 19 first shows that for all three types of transmitter ROs, the accuracy of the cross-KC705 channel remains above 95%, despite potential noise introduced by thermal conditions and the shared PCIe buses. Similarly, Figure 20 shows that accuracy remains above 95% when using these alternative ROs for stressors and receivers on a KC705 sink. Although in many cases, bits are again flipped, blocking combinatorial loops and introducing environmental noise cannot prevent our channel from operating.

### Table V: CPU-to-FPGA covert-channel results on the two PSUs, along with the number of threads for which maximum accuracy is achieved, and the corresponding bandwidth. † signifies that the recovered bit-pattern is flipped.

<table>
<thead>
<tr>
<th>PSU Property</th>
<th>AC701-1</th>
<th>AC701-2</th>
<th>KC705-1</th>
<th>KC705-2</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>Accuracy</td>
<td>95%</td>
<td>97%</td>
<td>95%</td>
</tr>
<tr>
<td>B</td>
<td># Threads</td>
<td>10</td>
<td>14</td>
<td>11</td>
</tr>
<tr>
<td>C</td>
<td># Stressors</td>
<td>1</td>
<td>1</td>
<td>4</td>
</tr>
<tr>
<td>D</td>
<td>Meas. Cycles</td>
<td>$2^{15}$</td>
<td>$2^{15}$</td>
<td>$2^{18}$</td>
</tr>
<tr>
<td>E</td>
<td>Measurements</td>
<td>500</td>
<td>500</td>
<td>500</td>
</tr>
<tr>
<td>F</td>
<td>Bandwidth</td>
<td>6.1 bps</td>
<td>6.1 bps</td>
<td>0.8 bps</td>
</tr>
<tr>
<td>G</td>
<td>Accuracy</td>
<td>81%</td>
<td>70%</td>
<td>$\dagger$84%</td>
</tr>
<tr>
<td>H</td>
<td># Threads</td>
<td>4</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>I</td>
<td># Stressors</td>
<td>1</td>
<td>1</td>
<td>5</td>
</tr>
<tr>
<td>J</td>
<td>Meas. Cycles</td>
<td>$2^{15}$</td>
<td>$2^{15}$</td>
<td>$2^{21}$</td>
</tr>
<tr>
<td>K</td>
<td>Measurements</td>
<td>1,200</td>
<td>1,200</td>
<td>500</td>
</tr>
<tr>
<td>L</td>
<td>Bandwidth</td>
<td>2.5 bps</td>
<td>2.5 bps</td>
<td>0.1 bps</td>
</tr>
</tbody>
</table>

### VI. Additional Covert Channels

In this section, we explore CPU-to-FPGA (Section VI-A) and GPU-to-FPGA covert channels (Section VI-B).

#### A. CPU Transmissions

In order to test the CPU-to-FPGA communication channel, we replace the power draw of the FPGA source with heavy CPU loads. To that end, we use the open-source stress program, which is available on Debian-based Linux distribution package managers [40]. We vary the number of threads that stress uses from 0 (i.e., no transmissions, corresponding to random measurements), up to the number of threads available on each computer, i.e., 24 on the CPU attached to PSU-A, and 4 on the CPU attached to PSU-B.

The measurement process and classification metric remain the same as for the cross-FPGA channels, but we introduce an additional delay of three seconds after the stress program has started to ensure full utilization of the cores, and an additional three seconds after killing the process, to ensure that
time for the KC705 receivers to PSU-A, and to increase accuracy, we reduce the measurement period. Moreover, when testing with GPUs with the open-source [38] program, which uses Nvidia’s CUDA architecture to fully utilize the GPU cores. As the two GPUs use different architectures, we compile and run the gpu_burn program against different Nvidia drivers and CUDA versions. These differences are summarized in Table VI. Moreover, we return to the default measurement period of $2^1 = 2^{21}$ cycles for the Kintex 7 boards, and increase the number of measurements for all boards to 1,500, reducing bandwidth by a factor of $3 \times$. These parameters and the corresponding results are summarized in Table VII. As in the CPU case, three seconds of delay are added after before and after the program, to allow usage to return to normal.

We plot the results for the two PSUs in Figure 21, which allows us to draw three main conclusions. First of all, there is a critical CPU activity threshold which is necessary to make the covert channel possible. On PSU-A, this requires about 4 threads for the AC701 boards, and 7 threads for the KC705 boards. Moreover, increasing the number of threads does not always make the covert channel more accurate. For example, increasing the number of CPU threads from 0 to 10 increases accuracy, but the accuracy generally plateaus between 10 and 17 CPU threads, and then decreases, perhaps due to hyper-threading. Finally, we notice that for a similar number of threads used, the accuracy on PSU-B is often higher compared to that for PSU-A. This parallels our cross-FPGA results of Section V, and indicates that PSU-B is generally more prone to covert communication. The maximum accuracy achieved, the number of CPU threads used, and other experimental parameters are summarized in Table V.

### B. GPU Transmissions

The process for testing GPU-to-FPGA transmissions is similar to that of CPU-to-FPGA transmissions. We stress the GPUs with the open-source `gpu_burn` [38] program, which uses Nvidia’s CUDA platform to fully utilize the GPU cores. As the two GPUs use different architectures, we compile and run the `gpu_burn` program against different Nvidia drivers and CUDA versions. These differences are summarized in Table VI. Moreover, we return to the default measurement period of $2^1 = 2^{21}$ cycles for the Kintex 7 boards, and increase the number of measurements for all boards to 1,500, reducing bandwidth by a factor of $3 \times$. These parameters and the corresponding results are summarized in Table VII. As in the CPU case, three seconds of delay are added after before and after the program, to allow usage to return to normal.

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TABLE VIII: Time to leak cryptographic keys of different sizes to two types of FPGA boards, without error correction.

<table>
<thead>
<tr>
<th>Algorithm</th>
<th>Key Size</th>
<th>AC701</th>
<th>KC705</th>
</tr>
</thead>
<tbody>
<tr>
<td>AES</td>
<td>256</td>
<td>0.7 min</td>
<td>44.7 min</td>
</tr>
<tr>
<td>ECDSA</td>
<td>521</td>
<td>1.4 min</td>
<td>91.1 min</td>
</tr>
<tr>
<td>RSA</td>
<td>1,024</td>
<td>2.8 min</td>
<td>179.0 min</td>
</tr>
</tbody>
</table>

To increase accuracy, one can either tweak the parameters of the source and sink FPGA designs (including the number of measurements $M$ over which RO counts are averaged), or instead change the communication scheme itself. For example, a 3-repetition code decreases bandwidth by a factor of 3, but also lowers the error rate to $3e^{-2} - 2e^3$; a 10% error rate is reduced to under 3%. The channel capacity is $1 - H(e) = 1 + e \log_2 e + (1-e) \log_2(1-e)$, and for smaller bitflip probabilities, other error correcting codes such as Hamming and Golay codes can be used to improve accuracy.

B. Defense Mechanisms

In this section, we discuss potential software and hardware defense countermeasures against voltage-based covert- and side-channel attacks. To start with, some countermeasures might revolve around preventing intentional transmissions from the covert-channel source. However, doing so would be particularly hard without huge sacrifices in terms of power and performance. Although we used ring oscillators to cause fluctuations in the voltage of FPGAs sharing the same PSU, other switching activity can also result in voltage over- and under-shoots. For example, prior work has shown that switching large sets of programmable interconnect points [48] or flip-flops [12], [13] can cause voltage fluctuations outside of the allowed operating voltage range for an FPGA device. Moreover, we demonstrated CPU-to-FPGA and GPU-to-FPGA channels, which show that the problem is not FPGA-specific, but can be found in other types of activities which result in large power draws. Consequently, unless power is equalized among all possible algorithm implementations, some leakage which can differentiate between levels of activity will persist.

To prevent side-channel attacks from being possible, designers may remove the power-draw dependence on the data being processed and increase the noise level. Although several masking and hiding techniques have been proposed, leakage on FPGAs persists due to variations in placement and routing [6]. Consequently, a better approach would be to prevent the leakage from being measurable on the FPGA sinks.

Current FPGA cloud providers prevent voltage and temperature monitors from being accessible by user logic and prohibit traditional LUT-ROs from being instantiated on their infrastructure [3]. However, alternative ring oscillator designs can bypass cloud restrictions [8], [10], [11], [34], and can also replace LUT-ROs (Section V-E). Moreover, time-to-digital converters (TDCs) can also be used instead of ring oscillators to monitor voltage fluctuations and conduct side-channel attacks [31], [32]. Although a compiler tool that checks for combinatorial loops and latches [15] would prevent some of the above monitoring logic, it would not catch FF-ROs.

Given that designing effective countermeasures against side- and covert-channel receivers is an arms race, defense-in-depth would dictate run-time solutions in addition to any preventive approach. One feature of the covert channel is the high switching activity on the receiver. Built-in voltage monitors could thus be used by cloud providers to detect abnormal fluctuations—with the caveat that legitimate circuits may also cause similar patterns, and that, at least on the AC701 boards, the number of enabled stressor ROs was small ($N_S = 500$). In fact, proposals to “detect the insertion of power measurement circuits onto a device’s power rail” [25] are similar, though the challenge is to reduce false positives.

Finally, better hardware (at a higher cost) can also help hide the useful signal under the noise floor. For example, independent, fully separate power supplies for different boards would require that the leakage be detectable even over the AC power line, and through two different AC-to-DC rectifiers. Moreover, better isolation of power circuits within the same PSU, as well as voltage regulators with better transient responses on both the source and the sink FPGAs, or differently-designed powering circuits with more filters and smoothing capacitors can also reduce the signal available to an attacker.

Overall, with prior work showing that FPGA are vulnerable to multi-tenants attacks [8], [9], [11], [28]–[31], [47] and with this article demonstrating cross-FPGA attacks, better architectural designs appear to be necessary both for FPGA chips and for the boards on which they are deployed.

VIII. RELATED WORK

This section summarizes prior work in remote FPGA attacks without physical access (Section VIII-A), as well as voltage- and temperature-based covert channels (Section VIII-B).

A. Remote FPGA Attacks

Although attacks on FPGA systems have traditionally required physical access to the FPGA board [46], a recent class of remote attacks has emerged. These attacks have used ROs as covert- and side-channel receivers, and also as covert-channel transmitters and fault attack inductors.

In the former category, Giechaskiel et al. showed that ROs can detect the value of nearby long wires in Xilinx FPGAs, and used this effect to create covert- and side-channel attacks on shared FPGAs [8]–[10]. It was later shown that the same phenomenon is present on Intel devices [29], [30], and that it can be used to extract keys from AES cores using on-chip Differential Power Analysis (DPA). These attacks have a weak threat model, as the source and sink long wires are adjacent.
A stronger attacker was performed by Zhao and Suh [47], who used ring oscillators as voltage monitors. Zhao and Suh conducted an intra-FPGA side-channel attack on RSA and AES cores, even with physical isolation of the attacker and victim circuits (on the same FPGA chip). Provelengios et al. then characterized voltage drops internal to an FPGA due to power waster RO circuits [28], while Giechaskiel et al. showed that covert channels are possible between separate dies on 2.5D-integrated FPGA chips (“Super Logic Regions”) [11]. Schellenberg et al. similarly recovered keys from an AES core in a multi-tenant setting, but used Time-to-Digital Converters (TDCs) to do so instead of ROs [31].

To the best of our knowledge, the only work which has considered cross-chip attacks is also by Schellenberg et al., where TDCs were used to conduct a cross-FPGA RSA key recovery [32]. However, the chips were located on the same FPGA board, and hence shared the same voltage regulator, making them easier to influence directly, due to the lack of additional intermediate components between their power distribution networks. By contrast, we introduced a novel way of using ring oscillators to detect FPGA-to-FPGA, CPU-to-FPGA, and GPU-to-FPGA covert communications between devices which are merely powered by the same power supply unit, but do not share voltage regulators. Moreover, unlike the work by Schellenberg et al., which used an FPGA board that is explicitly “designed for external side-channel analysis research” [32], we used off-the-shelf Xilinx-designed boards.

Ring oscillators have also been shown to be effective as temperature monitors. This was first demonstrated for thermal covert channels with physical access [35], but was later also shown for intra-FPGA covert channels, where ROs were used both as source heaters and sink monitors [19]. More recently, it was also shown that temporal thermal channels using ring oscillators are possible in cloud environments, where different users get time-shared access to the same FPGA fabric [37]. Ring oscillators can also be used to remotely cause fault attacks on FPGAs. These attacks can merely crash the FPGA [14], or cause timing violations, which can then be used to extract cryptographic keys [24]. Moreover, they can also be used to bias True Random Number Generators (TRNGs) [26].

B. Power and Temperature Covert Channels

It is well-known that data-dependent power consumption can be used to recover cryptographic keys through differential power analysis and other techniques by acquiring and analyzing power traces [23]. The same principles can be applied to create covert communication, for example from a malware app on a phone to a malicious USB charger [33], or from a program that modulates CPU utilization to an attacker measuring the current consumption of the computer [17]. Similarly, measuring voltage ripple on the power lines can be used to track the power usage pattern of other data center tenants [20]. Although these works exploit the same source of information leakage, they require external equipment to detect these data-dependent power variations, and are thus not applicable to cloud environments in practice. However, as we showed in this paper, it is possible to use the reconfigurable fabric of FPGAs as a covert-channel sink, allowing for accurate transmission of data remotely, without physical access.

Another category of power attacks which has recently been discovered is related to Dynamic Voltage and Frequency Scaling (DVFS) on modern processors, which regulates the voltage and frequency of CPUs in accordance with usage demands. Malicious software can exploit DVFS to cause faults in computations [36], or create covert channels between CPU cores, where the source core modulates frequency, and the sink core measures a reduction in its own performance [22].

Thermal attacks can also be used to create covert channels between CPU cores [27], but they require access to CPU thermal sensors, and are slower than their power counterparts, having a capacity of up to 300 bps [4]. Temperature-based covert channels need not be limited to communication within a single computer. Assuming computers are sufficiently close, a covert channel between nearby yet air-gapped devices is also possible with access to temperature sensors on the sink computer [16]. Finally, thermal information can also be used as a proxy estimate for power consumption in data centers. This information can alert potential adversaries to opportune moments to attack the availability of servers, either by exceeding the power capacity [21], or by more generally degrading performance [7]. Although these attacks require privileged thermal sensors, FPGA ROs could also be used for similar purposes, complementing our work in this paper.

IX. Conclusion

In this paper, we introduced the first FPGA-to-FPGA, CPU-to-FPGA, and GPU-to-FPGA voltage-based covert channels, achieving transmission accuracies of up to 100%. Unlike prior work, which unrealistically assumes that different users share the same FPGA fabric, our work considered a stronger threat model, where the FPGA chip and board are allocated on a per-user basis. Our covert channel exploited properties of the response of power supply units (PSUs) and voltage regulators to changes in their load. To detect these changes, we introduced a novel architectural design and classification metric which depends on stressor ring oscillators on the covert-channel sink FPGA. We showed that ring oscillators also performed well in the source FPGA, and further showed that heavy CPU and GPU activity could also be used as an effective transmitter. We demonstrated our covert channel on four Artix 7 and Kintex 7 boards, creating a channel of communication between any two of them in either direction, with high accuracy. We also performed an analysis of bandwidth-accuracy tradeoffs, and further explored the accuracy of the covert channel across different sizes and types of the sink and source FPGA circuits, different measurement patterns and setup layouts, and PSUs with different power ratings from two manufacturers. We finally also proposed potential countermeasures to prevent the information leakage we discovered from being exploitable. Overall, our remote covert-channel attacks highlight the dangers of shared power supply units, and therefore a need to rethink FPGA security, even for single-user monolithic designs.