

#### **GPGPU for side channel attacks**

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#### Plan

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#### Context : Side Channel Analysis



- The required trace number grows faster than the computation power.
- Need for new implementation techniques.

 $\Rightarrow$  An easy solution : OpenMP, API for parallel programming on multicore CPU.

 $\Rightarrow$ A more efficient solution : GPU.



#### GPU

- GPU : Graphic Processing Unit.
- Specialized electronic circuit designed to rapidly manipulate and alter memory in a frame buffer.
- Present in : graphic card, tablet, recent smartphone, etc.

#### GPU API

- NVIDIA : CUDA.
- ATI : ATI Stream.
- Standard : OpenCL.



#### OpenCL platform model



- One host...
- ...connected to computing devices (for example GPU)...
- ...regrouping several computing units.
- For GPU : Single Instruction Multiple Data (SIMD).

#### High computing power but not for free :

- Parallelization constrained by SIMD programming model.
  - $\Rightarrow$  same instruction computed by the computing units



#### **OpenCL** program





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#### Example : Correlation Power Analysis (Brier et al. 2004)

*N* traces C<sub>i</sub> associated to a model H<sub>k</sub>(i) for each key hypothesis k
 The Pearson coefficient is computed for each instant x of the traces.

$$\rho(x)_{k} = \frac{N \sum_{i=1}^{N} C_{i}(x) \cdot H_{k}(i) - \sum_{i=1}^{N} C_{i}(x) \cdot \sum_{i=1}^{N} H_{k}(i)}{\sqrt{N \sum_{i=1}^{N} C_{i}(x)^{2} - \left(\sum_{i=1}^{N} C_{i}(x)\right)^{2}} \cdot \sqrt{N \sum_{i=1}^{N} H_{k}(i)^{2} - \left(\sum_{i=1}^{N} H_{k}(i)\right)^{2}}}$$

Three kernels :

Compute : 
$$\sum_{i=1}^{N} H_k(i)$$
 and  $\sum_{i=1}^{N} H_k(i)^2$ 
Compute :  $\sum_{i=1}^{N} C_i(x)$ ,  $\sum_{i=1}^{N} C_i(x)^2$  and  $\sum_{i=1}^{N} C_i(x)H_k(i)$ 

Compute the Pearson coefficient for each k and each x



Comparisons of different implementations

## Hardware : CPU : XEON X3430

- 4 cores
- clock speed : 2.4GHz
- ► cost : ≈ 200\$
- GPU : ENGTS450
  - 192 CUDA cores
  - clock speed : 0.78GHz
  - ► cost : ≈ 200\$

#### Results : order of magnitude

Algorithm	Sequential	OpenMP	OpenCL
CPA	Т	T/3	<i>T /</i> 12
$\frac{\text{Variance}}{\overline{C(x)^2} - \overline{C(x)}^2}$	T'	T'/3	<i>T'/</i> 12



#### Example : Rank Correlation (Batina et al. 2008)

- Same computation than the CPA but the values of  $C_i(x)$  and  $H_k(i)$  are replaced by their rank.
- Many sets of size N have to be sorted.
- SIMD not designed for conditional function and sorting.
  - $\Rightarrow$  "only" 5 times faster than the sequential version.





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- Solution : Sorts by the CPU and the correlation computation by the GPU.
  - $\Rightarrow$  8 times faster than the sequential version.





#### Conclusion

- GPU usefull for SCA.
- Many algorithms from the Side Channel Analysis and from the Signal Processing.
   But not all of them.

 $\Rightarrow$  It is important to respect SIMD concepts and correctly determine the parallelism in the algorithm.

 $\Rightarrow$  Need for a prior analysis.



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### Thank you!

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