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Linux capable RISC-V CPU for IOb-SoC

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Declaration

I declare that this document is an original work of my own authorship and that it fulfills all the requirements of the Code of Conduct and Good Practices of the Universidade de Lisboa.

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Resumo

Com o avanço das tecnologias desenvolvidas em código aberto torna-se necessário estudar tanto o novo hardware desenvolvido como o software que tira partido deste novo hardware. Nesta dissertação de mestrado pretende-se conseguir correr um sistema operativo baseado em Linux numa variante do *IOb-SoC*. Ao longo deste trabalho à de-se realizar a implementação de um processador *RISC-V* de 32-*bits* capaz de correr o Linux no *IOb-SoC*. No final desta tese pretende-se: em primeiro lugar, ser capaz de correr uma simulação do sistema criado, que mostre o seu correto funcionamento; e em segundo lugar, implementar a variante do *IOb-SoC* desenvolvida numa FPGA e a partir desta FPGA correr o Linux.

Palavras-chave: RISC-V, Linux, Sistema num Chip (SoC), Verilog

Abstract

With the advances in new open-source technologies, it's imperial that the new hardware solutions and software implementation on the new hardware is studied. The aim of this thesis is to successfully run a Linux based OS (Operative System) on an IOB-SoC variant. During this work, the implementation of a 32-bit RISC-V CPU capable of running Linux on the Iob-SoC is going to be developed. At the end of this thesis, it's expected to: firstly, be able to run a simulation of the SoC (System on Chip) used to run the Linux kernel and verify its correct functionality; and secondly, implement the IOB-SoC variant developed in an FPGA and successfully boot Linux.

Keywords: RISC-V, Linux, Systems on Chip (SoC), Verilog

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Listings

List of Acronyms

AGU	Address Generator Unit
ALU	Arithmetic and Logic Unit
AP	Average Precision
API	Application Programming Interface
AXI	Advanced eXtensible Interface
BRAM	Block RAM
CISC	Complex Instruction Set Computer
CM	Configuration Module
CPU	Central Processing Unit
DMA	Direct Memory Access
DSP	Digital Signal Processing
FF	Flip-Flop
FM	Feature Map
FP	Fixed-Point
FPGA	Field Programmable Gate Array
FPS	Frames Per Second
FPU	Floating-point unit
FU	Functional Unit
GPP	General-Purpose Processor
GPU	Graphical Processing Unit
IOb-SoC	IObundle SoC
IoU	Intersection over Union
IPv4	Internet Protocol
LRU	Least Recently Used
LUT	Look-Up Table
MMU	Memory Management Unit
PE	Processing Element
RAM	Random Memory Access
ReLU	Rectified Linear Unit
RISC	Reduced Instruction Set Computer
ROM	Read Only Memory
SFP	Static Fixed Point
SIMD	Single Instruction Multiple Data
SoC	System on a Chip
SRAM	Static Random Access Memory
UART	Universal Asynchronous Receiver-Transmitter

Chapter 1

Introduction

Introdução

1.1 Motivation

Motivação

1.2 Objectives

Objectivo

1.3 Thesis Outline

Estutura do documento

Chapter 2

Linux on RISC-V Background

Things I need to talk about:

- To do

Chapter 3

Conclusions

Concluded

3.1 Achievements

Talk about what was achieved

3.2 Future Work

Talk about what can be improved in future works

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