

UNIVERSITY OF POTSDAM

MASTER'S THESIS

**SoPa++: Leveraging explainability from
hybridized RNN, CNN and weighted
finite-state neural architectures**

Author:

Atreya SHANKAR

1st Supervisor:

Dr. Sharid LOÁICIGA
University of Potsdam

2nd Supervisor:

Mathias MÜLLER
University of Zurich

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Chapter 1

Introduction

1.1 Main Section 1

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1.1.1 Subsection 1

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1.2 Main Section 2

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