
tf_onnx Documentation

Release

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This is an introduction tutorial to TF_ONNX. TF_ONNX is a conversion module to let a protobuffer defined on a protocol buffer another protobuffer on ONNX.

This supports not only just another straightforward conversion, but enables you to customize a given graph structure in a concise but very flexible manner to let the conversion job very tidy.

For instance, it will support following features in a least confusing manner.

- operation conversion (controlling number of arguments)
- sub-branch pruning (removing identity node on a tensorflow protobuf)
- sub-branch evaluation (useful for quantization function)
- visualization (dot language provided by graphviz library)

Note: This is still not a general converter from tensorflow pb to onnx pb as the combination of current onnx supported operation cannot cover the operator defined under the Nodedef object on tf-pb.

That being said, numbers of kinds of “op” on Nodedef() would be much less than seeming number of operators written in a tf-tutorial which is nearly 1000 as a function defined as python would be decomposed as a set of basic operators written in c++ on tf.

1.1 Installation

I am very afraid to let this library be dependedent on other big libraries such as Tensorflow or ONNX because it may give it some side effects to conversion process.

Both protocol buffer is therefore extracted from a snapshot of both.

only requirement so far is "numpy"

1.1.1 Install with docker

some command1

1.1.2 Install with virtualenv

some command2

1.2 An Explanation of relavant Protocol buffers

This page is for quick understanding of protocol buffer format.

[Tensorflow Protocol Buffer](#)

[ONNX Protocol Buffer](#)

1.3 Structure of Conversion

A conversion will be done by a tree structure and a set of higher order function on it.

Tree Structure

Higher order function on a tree

1.4 Examples

This page will introduce some basic examples for conversion and a few tools to make your life easier.

As this explanation will trace example codes which are put on a `./examples/` * refer them with this.

`./example/ex1.py`

So far, if somebody needs more explanations more than what was written on a each script,

I will add more on it.

1.5 tf_onnx

1.5.1 Submodules

`tf_onnx.core_`

Submodules

`tf_onnx.core_.node`

Classes

- Node: Base node object.

`tf_onnx.core_.node_`

Submodules

`tf_onnx.core_.node_.inner_`

Submodules

`tf_onnx.core_.node_.inner_.arg1_`

Classes

- Arithmetic: Arithmetic class has axes and keepdims as an attribute.
- Cast: Cast arg1 does not have any attributes to be added on ONNX
- Compare: individual axis are provided in each specific class
- F_shape: individual axis are provided in each specific class
- IO: IO has no role in this graph.

Classes

- Arg1: this class contains just a child.
- Arg2: this class has two children.
- ArgMany: this class will be set for an operator which has more than 3 arguments.

`tf_onnx.core_.node_.leaf_`

Submodules

`tf_onnx.core_.node_.leaf_.const`

Classes

- Const: Const node is one of leafs which contains values on a tensor.

`tf_onnx.core_.node_.leaf_.placeholder`

Classes

- Placeholder: Placeholder is a leaf but not initialized like data on bss.

Classes

- Leaf: leaf node has no children on a graph.
- Inner: there is still no method on this abstraction level but class itself

`tf_onnx.type_`

Submodules

`tf_onnx.type_.cast_`

Functions

- `pb_to_json(): TYPE :: pb object -> json pbjec`
- `json_to_pb(): TYPE :: json object on python -> pbobject on python -> pb_objec`
- `protobuf_to_dict(): Undocumented.`

Variables

- `TYPE_CALLABLE_MAP`
- `REVERSE_TYPE_CALLABLE_MAP`

`tf_onnx.type_.dict2pb`

Functions

- `protobuf_to_dict()`: Undocumented.
- `dict_to_protobuf()`: Populates a protobuf model from a dictionary.

Variables

- `TYPE_CALLABLE_MAP`
- `REVERSE_TYPE_CALLABLE_MAP`

CHAPTER 2

Indices and tables

- `genindex`
- `modindex`
- `search`