TopogrElow

TensorFlow API r1.4

tf.contrib.distributions.bijectors.SoftmaxCentered

Contents
Class SoftmaxCentered
Properties
dtype
event_ndims

Class SoftmaxCentered

Inherits From: **Bijector**

Defined in tensorflow/contrib/distributions/python/ops/bijectors/softmax_centered_impl.py.

See the guide: Random variable transformations (contrib) > Bijectors

Bijector which computes $Y = g(X) = exp([X \ 0]) / sum(exp([X \ 0]))$.

To implement softmax as a bijection, the forward transformation appends a value to the input and the inverse removes this coordinate. The appended coordinate represents a pivot, e.g., softmax(x) = exp(x-c) / sum(exp(x-c)) where c is the implicit last coordinate.

Because we append a coordinate, this bijector only supports event_ndim in [0, 1], i.e., scalars and vectors.

Example Use:

```
bijector.SoftmaxCentered(event_ndims=1).forward(tf.log([2, 3, 4]))
# Result: [0.2, 0.3, 0.4, 0.1]
# Extra result: 0.1

bijector.SoftmaxCentered(event_ndims=1).inverse([0.2, 0.3, 0.4, 0.1])
# Result: tf.log([2, 3, 4])
# Extra coordinate removed.
```

At first blush it may seem like the Invariance of domain theorem implies this implementation is not a bijection. However, the appended dimension makes the (forward) image non-open and the theorem does not directly apply.

Properties

dtype

dtype of **Tensor** s transformable by this distribution.

event_ndims

Returns then number of event dimensions this bijector operates on.

graph_parents

Returns this **Bijector** 's graph_parents as a Python list.

is_constant_jacobian

Returns true iff the Jacobian is not a function of x.



Note: Jacobian is either constant for both forward and inverse or neither.

Returns:

• is_constant_jacobian: Python bool.

name

Returns the string name of this **Bijector**.

validate_args

Returns True if Tensor arguments will be validated.

Methods

__init__

```
__init__(
   event_ndims=0,
   validate_args=False,
   name='softmax_centered'
```

forward

```
forward(
    Х,
    name='forward'
```

Returns the forward **Bijector** evaluation, i.e., X = g(Y).

Args:

- x: **Tensor** . The input to the "forward" evaluation.
- name: The name to give this op.

Returns:

Tensor.

Raises:

- TypeError: if self.dtype is specified and x.dtype is not self.dtype.
- NotImplementedError: if _forward is not implemented.

forward_event_shape

```
forward_event_shape(input_shape)
```

Shape of a single sample from a single batch as a TensorShape.

Same meaning as forward_event_shape_tensor. May be only partially defined.

Args:

• input_shape: TensorShape indicating event-portion shape passed into forward function.

Returns:

• forward_event_shape_tensor: **TensorShape** indicating event-portion shape after applying **forward**. Possibly unknown.

forward_event_shape_tensor

```
forward_event_shape_tensor(
   input_shape,
   name='forward_event_shape_tensor'
)
```

Shape of a single sample from a single batch as an int32 1D Tensor.

Args:

- input_shape: Tensor, int32 vector indicating event-portion shape passed into forward function.
- name: name to give to the op

Returns:

forward_event_shape_tensor: Tensor, int32 vector indicating event-portion shape after applying forward.

forward_log_det_jacobian

```
forward_log_det_jacobian(
    x,
    name='forward_log_det_jacobian'
)
```

Returns both the forward_log_det_jacobian.

Args:

• x: Tensor. The input to the "forward" Jacobian evaluation.

• name: The name to give this op.

Returns:

Tensor, if this bijector is injective. If not injective this is not implemented.

Raises:

- TypeError: if self.dtype is specified and y.dtype is not self.dtype.
- NotImplementedError: if neither _forward_log_det_jacobian nor{_inverse, _inverse_log_det_jacobian} are implemented, or this is a non-injective bijector.

inverse

```
inverse(
    y,
    name='inverse'
)
```

Returns the inverse **Bijector** evaluation, i.e., $X = g^{-1}(Y)$.

Args:

- y: **Tensor**. The input to the "inverse" evaluation.
- name: The name to give this op.

Returns:

Tensor, if this bijector is injective. If not injective, returns the k-tuple containing the unique k points $(x1, \ldots, xk)$ such that g(xi) = y.

Raises:

- TypeError: if self.dtype is specified and y.dtype is not self.dtype.
- NotImplementedError: if _inverse is not implemented.

inverse_event_shape

```
inverse_event_shape(output_shape)
```

Shape of a single sample from a single batch as a TensorShape.

Same meaning as inverse_event_shape_tensor . May be only partially defined.

Args:

output_shape: TensorShape indicating event-portion shape passed into inverse function.

Returns:

• inverse_event_shape_tensor: TensorShape indicating event-portion shape after applying inverse. Possibly

unknown.

inverse_event_shape_tensor

```
inverse_event_shape_tensor(
    output_shape,
    name='inverse_event_shape_tensor'
)
```

Shape of a single sample from a single batch as an int32 1D Tensor.

Args:

- output_shape: Tensor, int32 vector indicating event-portion shape passed into inverse function.
- name: name to give to the op

Returns:

• inverse_event_shape_tensor: Tensor, int32 vector indicating event-portion shape after applying inverse.

inverse_log_det_jacobian

```
inverse_log_det_jacobian(
    y,
    name='inverse_log_det_jacobian'
)
```

Returns the (log o det o Jacobian o inverse)(y).

Mathematically, returns: log(det(dX/dY))(Y). (Recall that: $X=g^{-1}(Y)$.)

Note that $forward_log_det_jacobian$ is the negative of this function, evaluated at $g^{-1}(y)$.

Args:

- y: Tensor. The input to the "inverse" Jacobian evaluation.
- name: The name to give this op.

Returns:

Tensor, if this bijector is injective. If not injective, returns the tuple of local log det Jacobians, $log(det(Dg_i^{-1}_{-1}(y)))$, where g_i is the restriction of g to the g-injective, returns the tuple of local log det Jacobians, $log(det(Dg_i^{-1}_{-1}(y)))$,

Raises:

- TypeError: if self.dtype is specified and y.dtype is not self.dtype.
- NotImplementedError: if _inverse_log_det_jacobian is not implemented.

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