#### TopogrElow

TensorFlow API r1.4

tf.contrib.distributions.bijectors.Exp

```
Contents
Class Exp
Properties
dtype
event_ndims
```

# Class Exp

Inherits From: PowerTransform

Defined in tensorflow/contrib/distributions/python/ops/bijectors/exp\_impl.py.

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

```
Compute Y = g(X) = exp(X).
```

Example Use:



Note: the exp(.) is applied element-wise but the Jacobian is a reduction over the event space.

# **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**.

#### power

```
The c in: Y = g(X) = (1 + X * c)**(1 / c).
```

## validate\_args

Returns True if Tensor arguments will be validated.

## Methods

## \_\_init\_\_

```
__init__(
    event_ndims=0,
    validate_args=False,
    name='exp'
)
```

Instantiates the Exp bijector.

#### Args:

- event\_ndims: Scalar int32 Tensor indicating the number of dimensions associated with a particular draw from the distribution.
- validate\_args: Python bool indicating whether arguments should be checked for correctness.
- name: Python str name given to ops managed by this object.

#### 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.

Except as otherwise noted, the content of this page is licensed under the Creative Commons Attribution 3.0 License, and code samples are licensed under the Apache 2.0 License. For details, see our Site Policies. Java is a registered trademark of Oracle and/or its affiliates.

Last updated November 2, 2017.

