# How to avoid NaN gradients when using `tf.where` to select finite results.

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## Abstract

So you have a NaN in your gradient and don't know why? Assuming you use `tf.where`, `tf.minimum`, `tf.maximum`, this note might help!

Tl;dr: Instead of this: tf.where(x\_ok, f(x), safe\_f(x))
Do this: tf.where(x\_ok, f(tf.where(x\_ok, x, safe\_x)), safe\_f(x))

Both give the same result. Only the latter gives the correct<sup>1</sup> gradient.

(Keep reading if the "tl;dr" doesn't make sense!)

### **Detailed Example**

Let's develop our intuition of the problem by considering a specific example. Suppose you wish to differentiate the following function:

$$f(x) = \begin{cases} \frac{1}{x} & x \ge 1\\ 0 & x < 1 \end{cases}$$

A naive implementation results in NaNs in the gradient, i.e.,

```
import tensorflow.compat.v2 as tf
import tensorflow_probability as tfp
tf.enable_v2_behavior()

f = lambda x: tf.where(x < 1., 0., 1. / x)
x = tf.constant(0.)
tfp.math.value_and_gradient(f, x)[1]
# ==> nan ...bah.
```

<sup>&</sup>lt;sup>1</sup> Arguably. Don't ask.

The basic pattern for avoiding NaN gradients when using `tf.where` is to call `tf.where` *twice*.<sup>2</sup> The innermost `tf.where` ensures that the result `f(x)` is always finite. The outermost `tf.where` ensures the correct result is chosen. For the running example, the trick plays out like this:

```
def safe_f(x):
    safe_x = tf.where(tf.equal(x, 0.), 1., x)  # inner tf.where
    return tf.where(x < 1., 0., 1. / safe_x)  # outer tf.where; just like f(x)</pre>
```

But did it work?

```
x = tf.constant(0.)
tfp.math.value_and_gradient(safe_f, x)[1]
# ==> 0.0 ...yay! double-where trick worked.
```

#### **General Recipe**

- Use an inner `tf.where` to ensure the function has no asymptote.
   I.e., alter the input to the inf generating function such that no inf can be created.
- Use a second `tf.where` to always select the valid code-path.
   I.e., implement the mathematical condition as you would "normally", i.e., the "naive" implementation.

In Python code, the recipe is:

```
Instead of this: tf.where(x_ok, f(x), alt_f(x))
Do this: tf.where(x_ok, f(tf.where(x_ok, x, safe_x)), alt_f(x))
```

#### Can we do better?

With luck, sometimes things work out even more cleanly. For example,

```
def cross_entropy(x, y, axis=-1):
    safe_y = tf.where(tf.equal(y, 0.), tf.ones_like(y), y)
    return -tf.reduce_sum(x * tf.math.log(safe_y), axis)
def entropy(x, axis=-1):
    return cross_entropy(x, x, axis)
```

Here we only needed one `tf.where` because the `x \*` acts like an outer `tf.where`.

But did it work?

<sup>&</sup>lt;sup>2</sup> The "double-tf.where" trick always works, assuming you have access to the data *before* it becomes +/- inf.

```
x = tf.constant([0.1, 0.2, 0., 0.7])
e = entropy(x)
# ==> 0.80181855
tfp.math.value_and_gradient(entropy, x)[1]
# ==> [1.30258512, 0.60943794, 0., -0.64332503] ...yay! no nan.
```

For additional discussion, see this StackOverflow post.