

# Contributing to PyMC

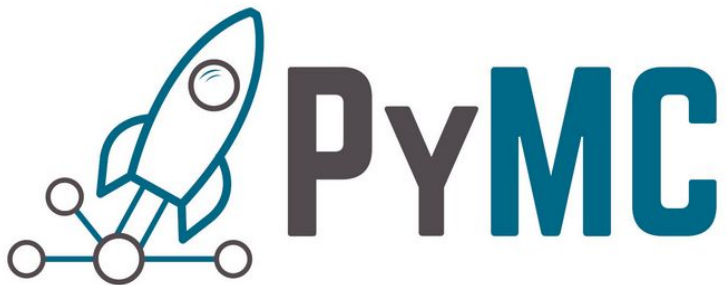
A testimonial by Ricardo Vieira

Slides at  
[tinyurl.com/contributing-pymc](https://tinyurl.com/contributing-pymc)



# My intentions

- Illustrate the Open Source collaboration experience
  - What it looks like when you are just starting
  - What it looks like for a regular contributor
- Share my biased views
  - It feels very random, you are unlikely to know where it leads
  - It is a great opportunity to learn
  - A lot of **fun** (hopefully)
- Invite you to give it a try!

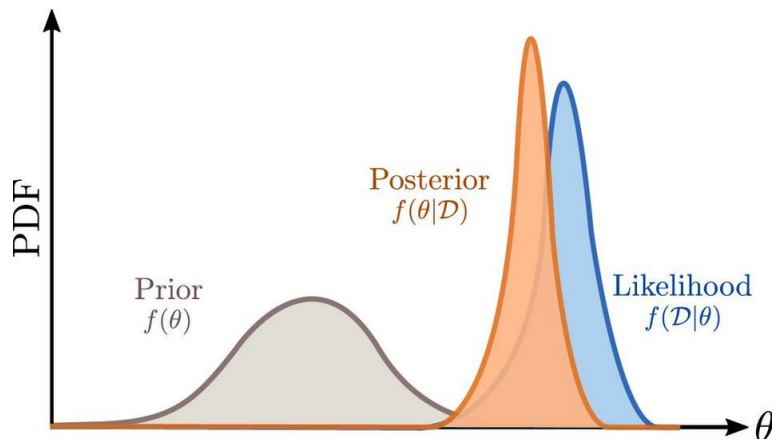


Probabilistic Programming in Python



# PYMC

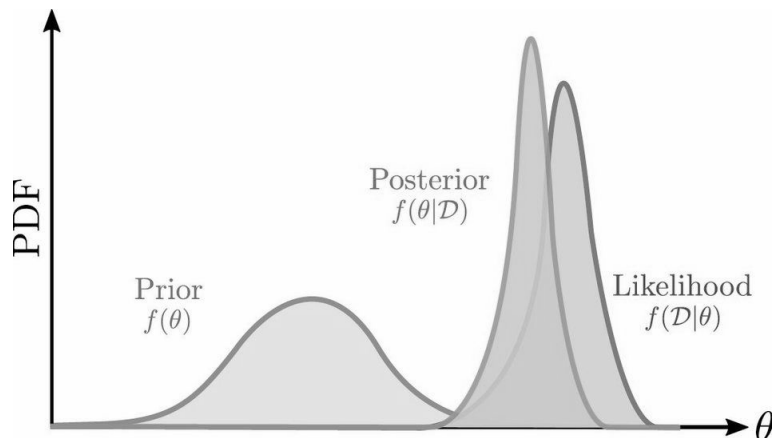
Probabilistic Programming in Python





# PYMC

Probabilistic Programming in Python

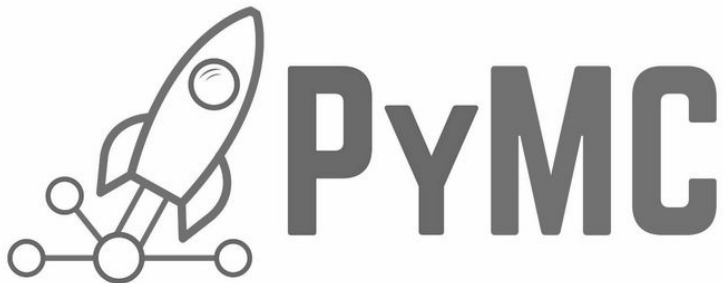


```
import pymc as pm
import pandas as pd
```

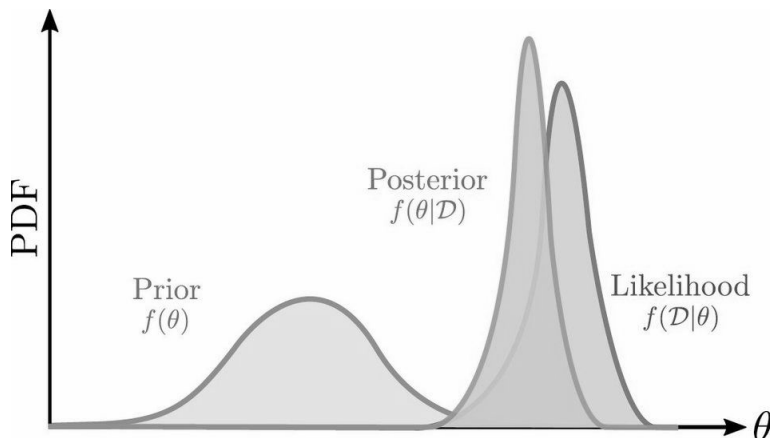
```
obs_data = pd.read_csv("my_observed_data.csv")
```

```
with pm.Model() as m:
    mean = pm.Normal("mean", 0, 1)
    noise = pm.HalfNormal("noise", 1)
    data = pm.Normal("data", mean, noise, observed=obs_data)

    posterior = pm.sample()
```



Probabilistic Programming in Python



```
import pymc as pm
import pandas as pd
```

```
obs_data = pd.read_csv("my_observed_data.csv")
```

```
with pm.Model() as m:
    mean = pm.Normal("mean", 0, 1)
    noise = pm.HalfNormal("noise", 1)
    data = pm.Normal("data", mean, noise, observed=obs_data)

    posterior = pm.sample()
```

Learn more at: [docs.pymc.io/en/latest](https://docs.pymc.io/en/latest)

[Austin Rochford: Intro to Probabilistic Programming with PyMC](#)

**PyMC timeline**

**Personal timeline**

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school



	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)
2013	PyMC v3 initial release	Bachelors in Psychology, learns some old fashioned statistics

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)
2013	PyMC v3 initial release	Bachelors in Psychology, learns some old fashioned statistics
2016	PyMC v3 paper published	Masters in Neuroscience, ports some Python2 code (err... adds parentheses to print statements)

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)
2013	PyMC v3 initial release	Bachelors in Psychology, learns some old fashioned statistics
2016	PyMC v3 paper published	Masters in Neuroscience, ports some Python2 code (err... adds parentheses to print statements)
2018	TensorFlow backend alternative explored	Starts PhD in Cognitive Science, learns some edgy statistics

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)
2013	PyMC v3 initial release	Bachelors in Psychology, learns some old fashioned statistics
2016	PyMC v3 paper published	Masters in Neuroscience, ports some Python2 code (err... adds parentheses to print statements)
2018	TensorFlow backend alternative explored	Starts PhD in Cognitive Science, learns some edgy statistics
2020	PyMC developers decide to maintain Theano	First contributions to PyMC

	<b>PyMC timeline</b>	<b>Personal timeline</b>
2003	PyMC v1 development starts	Plays Pokemon in primary school
2010	PyMC v2 paper published	Dabbles in Javascript during high school (err... watches youtube tutorials)
2013	PyMC v3 initial release	Bachelors in Psychology, learns some old fashioned statistics
2016	PyMC v3 paper published	Masters in Neuroscience, ports some Python2 code (err... adds parentheses to print statements)
2018	TensorFlow backend alternative explored	Starts PhD in Cognitive Science, learns some edgy statistics
2020	PyMC developers decide to fork Theano	First contributions to PyMC
2021	PyMC v4 beta is released	Becomes core developer Participates in Google Summer of Code Quits PhD Gets financial support from Chan Zuckerberg Initiative (Via NumFocus)

# An innocent feature request

## Add alternative parameterization to negative binomial distribution

🔒 Closed ricardoV94 opened this issue on 23 Sep 2020 · 2 comments



ricardoV94 commented on 23 Sep 2020

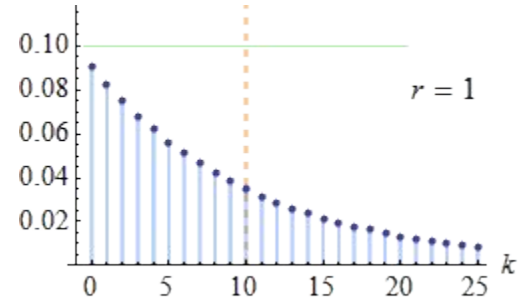
Member 😊 ...

The current parameterization in terms of  $\mu$  and  $\alpha$  is very useful in relation to the less flexible Poisson distribution.

However, the negative binomial distribution is often discussed in terms of the number of failures observed until a target number of successes is reached. In this case it is usually parameterized by a parameter  $p$  (the probability of observing a success in each trial), and a parameter  $n$  (the target number of successes). Would it make sense to add this alternative parameterization to the Negative Binomial distribution?

- $n$  is equivalent to the alpha parameter
- $p$  is equivalent to  $n / (n + \mu)$  or, equivalently,  $\mu = n(1-p)/p$




This parameterization is implemented and discussed in the [R implementation](#)



# An innocent feature request

## Add alternative parameterization to negative binomial distribution

 Closed **ricardoV94** opened this issue on 23 Sep 2020 · 2 comments

 **ricardoV94** commented on 23 Sep 2020 Member  

The current parameterization in terms of  $\mu$  and  $\alpha$  is very useful in relation to the less flexible Poisson distribution.

However, the negative binomial distribution is often discussed in terms of the number of failures observed until a target number of successes is reached. In this case it is usually parameterized by a parameter  $p$  (the probability of observing a success in each trial), and a parameter  $n$  (the target number of successes). Would it make sense to add this alternative parameterization to the Negative Binomial distribution?

- $n$  is equivalent to the alpha parameter
- $p$  is equivalent to  $n / (n + \mu)$  or, equivalently,  $\mu = n(1-p)/p$

This parameterization is implemented and discussed in the R implementation



**twlechl** commented on 23 Sep 2020


Great, want to do a PR [@ricardoV94?](#)



# An innocent feature request

## Add alternative parameterization to negative binomial distribution

🔒 Closed ricardoV94 opened this issue on 23 Sep 2020 · 2 comments

 ricardoV94 commented on 23 Sep 2020 Member 😊 ...

The current parameterization in terms of  $\mu$  and  $\alpha$  is very useful in relation to the less flexible Poisson distribution.

However, the negative binomial distribution is often discussed in terms of the number of failures observed until a target number of successes is reached. In this case it is usually parameterized by a parameter  $p$  (the probability of observing a success in each trial), and a parameter  $n$  (the target number of successes). Would it make sense to add this alternative parameterization to the Negative Binomial distribution?

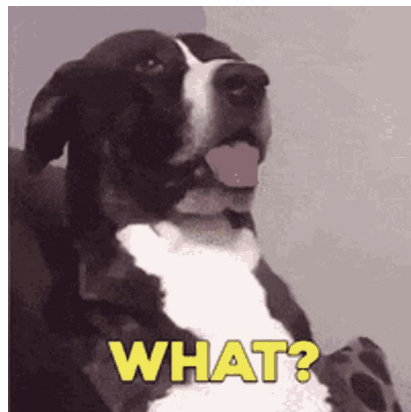
- $n$  is equivalent to the alpha parameter
- $p$  is equivalent to  $n / (n + \mu)$  or, equivalently,  $\mu = n(1-p)/p$

This parameterization is implemented and discussed in the R implementation



twleckl commented on 23 Sep 2020




Great, want to do a PR @ricardoV94?



# An innocent feature request

## Add alternative parameterization to negative binomial distribution

 Closed **ricardoV94** opened this issue on 23 Sep 2020 · 2 comments

 **ricardoV94** commented on 23 Sep 2020 Member  

The current parameterization in terms of  $\mu$  and  $\alpha$  is very useful in relation to the less flexible Poisson distribution.

However, the negative binomial distribution is often discussed in terms of the number of failures observed until a target number of successes is reached. In this case it is usually parameterized by a parameter  $p$  (the probability of observing a success in each trial), and a parameter  $n$  (the target number of successes). Would it make sense to add this alternative parameterization to the Negative Binomial distribution?

- $n$  is equivalent to the alpha parameter
- $p$  is equivalent to  $n / (n + \mu)$  or, equivalently,  $\mu = n(1-p)/p$

This parameterization is implemented and discussed in the R implementation



**twlechl** commented on 23 Sep 2020

Great, want to do a PR [@ricardoV94](#)?



**ricardoV94** commented on 23 Sep 2020

I can try :)

# What does a (first) PR look like?

## Add alternative parameterization to negative binomial distribution #4126 #4134

Merged twiecki merged 6 commits into `pymc-devs:master` from `ricardoV94:master` on 1 Oct 2020

Conversation 15 Commits 6 Checks 3 Files changed 3



ricardoV94 commented on 26 Sep 2020 • edited by twiecki

Member

Changes allow the specification of the NegativeBinomial distribution in terms of  $p$  (the probability of observing a success in each trial), and  $n$  (the target number of successes). This parametrization gives the likelihood for  $y$ , the number of failures observed until a target number of successes is reached.

I changed the docstrings and distribution initialization in line with other distribution that have multiple parametrizations.

I didn't figure out yet how the tests work, so I did not implement it (hence the WIP)! This should be straightforward, however, since the scipy negativebinomial distribution is implemented in terms of  $n$  and  $p$ . Any guidance in how to proceed is much appreciated!

Pinging @twiecki

Ref #4126

### Reviewers

MarcoGorelli  
twiecki

### Assignees

No one—assign yourself

### Labels

None yet

### Projects

None yet

# What does a (first) PR look like?

## Add alternative parameterization to negative binomial distribution #4126 #4134

Merged **tweiekl** merged 6 commits into `pymc-devs:master` from `ricardoV94:master` on 1 Oct 2020

Conversation 15 Commits 6 Checks 3 Files changed 3

**ricardoV94** commented on 26 Sep 2020 • edited by **tweiekl** Member

Changes allow the specification of the NegativeBinomial distribution in terms of  $p$  (the probability of observing a success in each trial), and  $n$  (the target number of successes). This parametrization gives the likelihood for  $y$ , the number of failures observed until a target number of successes is reached.

I changed the docstrings and distribution initialization in line with other distribution that have multiple parametrizations.

I didn't figure out yet how the tests work, so I did not implement it (hence the WIP)! This should be straightforward, however, since the `scipy` negative binomial distribution is implemented in terms of  $n$  and  $p$ . Any guidance in how to proceed is much appreciated!

Pinging **@tweiekl**

Ref #4126

### Reviewers

- MarcoGorelli**
- tweiekl**

### Assignees

No one—assign yourself

### Labels

None yet

### Projects

None yet

```
00 000, 20 +000, 50 00 def NegativeBinomial(n, p, x):
001 001
002 002 """
003 003 The negative binomial distribution can be parametrized either in terms of mu or p,
004 004 and either in terms of alpha or n. The link between the parametrization is given by
005 005 .. math::
006 006 \mu = n * p / (1 - p)
007 007 \alpha = n
008 008 """
009 009 Parameters
010 010 -----
011 011 mu: float
012 012 Poisson distribution parameter (mu > 0).
013 013 alpha: float
014 014 Same distribution parameter (alpha > 0).
015 015 p: float
016 016 Alternative probability of success in each trial (0 <= p <= 1).
017 017 n: float
018 018 Alternative number of target success trials (n > 0)
019 019 """
020 020
021 021 def __init__(self, mu, alpha, *args, **kwargs):
022 022 super().__init__(args, **kwargs)
023 023 mu, alpha = self.get_mu_alpha(mu, alpha, p)
024 024 self.mu = mu + 1e-10 * tensor_like(float)(mu)
025 025 self.alpha = alpha + 1e-10 * tensor_like(float)(alpha)
026 026 self.mode = int(np.floor(mu))
027 027
028 028 def get_mu_alpha(self, mu=None, alpha=None, p=None, verbose):
029 029 if alpha is None:
030 030 if p is not None:
031 031 alpha = n
032 032 else:
033 033 raise ValueError
034 034 "Incompatible parametrization. Must specify either alpha or n."
035 035
036 036 elif p is not None:
037 037 raise ValueError
038 038 "Incompatible parametrization. Can't specify both alpha and n."
039 039
040 040
041 041 if mu is None:
042 042 if p is not None:
043 043 mu = alpha * (1 - p) / p
044 044 else:
045 045 raise ValueError
046 046 "Incompatible parametrization. Must specify either mu or p."
047 047
048 048 elif p is not None:
049 049 raise ValueError
050 050 "Incompatible parametrization. Can't specify both mu and p."
051 051
052 052
053 053 return mu, alpha
054 054
055 055 def random(self, point=None, size=None):
056 056 """
057 057 Draw random values from NegativeBinomial distribution.
```

# What does a (first) PR look like?

## Add alternative parameterization to negative binomial distribution #4126 #4134

Merged twiecki merged 6 commits into pymc-devs:master from ricardoV94:master on 1 Oct 2020

Conversation 15 Commits 6 Checks 3 Files changed 3



ricardoV94 commented on 26 Sep 2020 • edited by twiecki

Member

Changes allow the specification of the NegativeBinomial distribution in terms of  $p$  (the probability of observing a success in each trial), and  $n$  (the target number of successes). This parametrization gives the likelihood for  $y$ , the number of failures observed until a target number of successes is reached.

I changed the docstrings and distribution initialization in line with other distributions that have multiple parametrizations.

I didn't figure out yet how the tests work, so I did not implement it (hence the WIP)! This should be straightforward, however, since the scipy negative binomial distribution is implemented in terms of  $n$  and  $p$ . Any guidance in how to proceed is much appreciated!

Pinging @twiecki

Ref #4126



ricardoV94 added 2 commits 16 months ago



Add alternative parameters  $p$  and  $n$  0bb12ec



Update Release Notes 241b311

### Reviewers

- MarcoGorelli
- twiecki

### Assignees

No one—assign yourself

### Labels

None yet

### Projects

None yet

```
def __init__(self, mu, alpha, *args, **kwargs):
    def __init__(self, mu, alpha, p, n, *args, **kwargs):
        super().__init__(args, **kwargs)
        mu, alpha = self.get_mu_alpha(mu, alpha, p, n)
        self.mu = mu + 1e-10 * tensor_variable(float(mu))
        self.alpha = alpha + 1e-10 * tensor_variable(float(alpha))
        self.mode = int(np.floor(mu))

    def get_mu_alpha(self, mu=None, alpha=None, p=None, n=None):
        if alpha is None:
            if n is not None:
                alpha = n
            else:
                raise ValueError("Incompatible parametrization. Must specify either alpha or n.")
        elif p is not None:
            raise ValueError("Incompatible parametrization. Can't specify both alpha and n.")
        else:
            raise ValueError("Incompatible parametrization. Can't specify both alpha and n.")

        return mu, alpha

    def __call__(self, point=None, size=None):
        """
        Draw random values from NegativeBinomial distribution.
        """
```

# What does a (first) PR look like?

## Add alternative parameterization to negative binomial distribution #4126 #4134

Merged twiecki merged 6 commits into pymc-devs:master from ricardoV94:master on 1 Oct 2020

Conversation 15 Commits 6 Checks 3 Files changed 3



ricardoV94 commented on 26 Sep 2020 • edited by twiecki

Member

Changes allow the specification of the NegativeBinomial distribution in terms of  $p$  (the probability of observing a success in each trial), and  $n$  (the target number of successes). This parametrization gives the likelihood for  $y$ , the number of failures observed until a target number of successes is reached.

I changed the docstrings and distribution initialization in line with other distribution that have multiple parametrizations.

I didn't figure out the scipy negat

I didn't figure out yet how the tests work, so I did not implement it (hence the WIP)!

Pinging @twiecki

Ref #4126

### Reviewers

- MarcoGorelli
- twiecki

### Assignees

None yet

None yet

Projects

None yet

ricardoV94 added 2 commits 16 months ago

Add alternative parameters  $p$  and  $n$  0bb12ec

Update Release Notes X 241b311

```
def __init__(self, mu, alpha, *args, **kwargs):
    super().__init__(args, **kwargs)
    mu, alpha = self.get_mu_alpha(mu, alpha, *args, **kwargs)
    self.mu = mu = It.is_tensor_variable(float(mu))
    self.alpha = alpha = It.is_tensor_variable(float(alpha))
    self.name = self.__class__.__name__

    def get_mu_alpha(self, mu=None, alpha=None, p=None, n=None):
        if alpha is None:
            if n is not None:
                alpha = n
            else:
                raise ValueError(
                    "Incompatible parametrization. Must specify either alpha or n."
                )
        elif p is not None:
            if n is not None:
                raise ValueError(
                    "Incompatible parametrization. Can't specify both alpha and n."
                )
            else:
                alpha = n * p

        return mu, alpha

    def random(self, size=None, **kwargs):
        """Draw random values from NegativeBinomial distribution.

        Parameters
        -----
        mu: float
            Poisson distribution parameter ( $\mu > 0$ ).
        alpha: float
            Gamma distribution parameter ( $\alpha > 0$ ).
        p: float
            Alternative probability of success in each trial ( $0 < p < 1$ ).
        n: float
            Alternative number of target success trials ( $n > 0$ )
        """
        ...

    def __init__(self, mu, alpha, *args, **kwargs):
    def __init__(self, mu=None, alpha=None, p=None, n=None, *args, **kwargs):
    super().__init__(args, **kwargs)
    mu, alpha = self.get_mu_alpha(mu, alpha, *args, **kwargs)
    self.mu = mu = It.is_tensor_variable(float(mu))
    self.alpha = alpha = It.is_tensor_variable(float(alpha))
    self.name = self.__class__.__name__


    def get_mu_alpha(self, mu=None, alpha=None, p=None, n=None):
        if alpha is None:
            if n is not None:
                alpha = n
            else:
                raise ValueError(
                    "Incompatible parametrization. Must specify either alpha or n."
                )
        elif p is not None:
            if n is not None:
                raise ValueError(
                    "Incompatible parametrization. Can't specify both alpha and n."
                )
            else:
                alpha = n * p

        return mu, alpha


    def random(self, size=None, **kwargs):
        """Draw random values from NegativeBinomial distribution.

        Parameters
        -----
        mu: float
            Poisson distribution parameter ( $\mu > 0$ ).
        alpha: float
            Gamma distribution parameter ( $\alpha > 0$ ).
        p: float
            Alternative probability of success in each trial ( $0 < p < 1$ ).
        n: float
            Alternative number of target success trials ( $n > 0$ )
        """
        ...
```


# What does a (first) PR look like?

 **AlexAndorra** commented on 28 Sep 2020 Member ...


Thanks for the PR [@ricardoV94](#) ! For the tests, I think you can take inspiration from the tests of Negative Binomial done with the current parametrization. Doing them with the new one should already be a good sample of use-cases





# What does a (first) PR look like?

 AlexAndorra commented on 28 Sep 2020 Member ...

Thanks for the PR @ricardoV94 ! For the tests, I think you can take inspiration from the tests of Negative Binomial done with the current parametrization. Doing them with the new one should already be a good sample of use-cases



  Add test ✖ edc5cf4


```

v ↕ 6 ■■■■■■ pymc3/tests/test_distributions.py 📄
...
795 795         return sp.nbinom.logpmf(value, alpha, 1 - mu / (mu + alpha))
796 796
797 797         self.pymc3_matches_scipy(NegativeBinomial, Nat, {"mu": Rplus, "alpha": Rplus}, test_fun)
798 +         self.pymc3_matches_scipy(
799 +             NegativeBinomial,
800 +             Nat,
801 +             {"p": Unit, "n": Rplus},
802 +             lambda value, p, n: sp.nbinom.logpmf(value, n, p),
803 +         )
798 804
799 805     def test_laplace(self):
800 806         self.pymc3_matches_scipy(
...
+

```




# What does a (first) PR look like?

 **AlexAndorra** commented on 28 Sep 2020 Member ...

Thanks for the PR [@ricardoV94](#) ! For the tests, I think you can take inspiration from the tests of Negative Binomial done with the current parametrization. Doing them with the new one should already be a good sample of use-cases

  Add test

× edc5cf4

 **ricardoV94** commented on 29 Sep 2020 Member Author ...


[@twiecki](#) [@AlexAndorra](#) would the added test be enough?

```

v ↕ 6 ██████ pymc3/tests/test_distributions.py ↗
...      @@ -795,6 +795,12 @@ def test_fun(value, mu, alpha):
795 795         return sp.nbinom.logpmf(value, alpha, 1 - mu / (mu + alpha))
796 796
797 797         self.pymc3_matches_scipy(NegativeBinomial, Nat, {"mu": Rplus, "alpha": Rplus}, test_fun)
798 +         self.pymc3_matches_scipy(
799 +             NegativeBinomial,
800 +             Nat,
801 +             {"p": Unit, "n": Rplus},
802 +             lambda value, p, n: sp.nbinom.logpmf(value, n, p),
803 +         )
798 804
799 805     def test_laplace(self):
800 806         self.pymc3_matches_scipy(
...
+


```

# What does a (first) PR look like?

 **AlexAndorra** commented on 28 Sep 2020 Member ...

Thanks for the PR @ricardoV94 ! For the tests, I think you can take inspiration from the tests of Negative Binomial done with the current parametrization. Doing them with the new one should already be a good sample of use-cases

 Add test × edc5cf4


 **ricardoV94** commented on 29 Sep 2020 Member Author ...

@twecki @AlexAndorra would the added test be enough?

```
┆ 6 pymc3/tests/test_distributions.py
┆ @ -795,6 +795,12 @@ def test_fun(value, mu, alpha):
795 795     return sp.nbinom.logpmf(value, alpha, 1 - mu / (mu + alpha))
796 796
797 797     self.pymc3_matches_scipy(NegativeBinomial, Nat, {"mu": Rplus, "alpha": Rplus}, test_fun)
798 +     self.pymc3_matches_scipy(
799 +         NegativeBinomial,
800 +         Nat,
801 +         {"p": Unit, "n": Rplus},
802 +         lambda value, p, n: sp.nbinom.logpmf(value, n, p),
803 +     )
798 804
799 805     def test_laplace(self):
800 806         self.pymc3_matches_scipy(
```




# What does a (first) PR look like?


 **AlexAndorra** commented on 28 Sep 2020 Member ...

Thanks for the PR @ricardoV94 ! For the tests, I think you can take inspiration from the tests of Negative Binomial done with the current parametrization. Doing them with the new one should already be a good sample of use-cases


  Add test × edc5cf4

 **ricardoV94** commented on 29 Sep 2020 Member Author ...

@twiecki @AlexAndorra would the added test be enough?

 **MarcoGorelli** commented on 29 Sep 2020 Member ...

IMO there should also be a test which hits the various error messages you've added, e.g. when you set both `alpha` and `n` - is that something you'd be happy to add? Feel free to ping if you want/need help

 1

```
pymc3/tests/test_distributions.py
...
795 795         return sp.nbinom.logpmf(value, alpha, 1 - mu / (mu + alpha))
796 796
797 797         self.pymc3_matches_scipy(NegativeBinomial, Nat, {"mu": Rplus, "alpha": Rplus}, test_fun)
798 +         self.pymc3_matches_scipy(
799 +             NegativeBinomial,
800 +             Nat,
801 +             {"p": Unit, "n": Rplus},
802 +             lambda value, p, n: sp.nbinom.logpmf(value, n, p),
803 +         )
798 804
799 805     def test_laplace(self):
800 806         self.pymc3_matches_scipy(
...
+
```



# What does a (first) PR look like?



**ricardoV94** commented on 29 Sep 2020

Member

Author



@**MarcoGorelli** That makes sense.

# What does a (first) PR look like?



ricardoV94 commented on 29 Sep 2020

Member

Author



@MarcoGorelli That makes sense.

I am not sure how to do it. Could you direct me to a snippet where this type of test is implemented? For example, the Beta distribution also raises this type of ValueErrors, but I couldn't find if it is being tested anywhere.

# What does a (first) PR look like?



ricardoV94 commented on 29 Sep 2020

Member

Author



@MarcoGorelli That makes sense.

I am not sure how to do it. Could you direct me to a snippet where this type of test is implemented? For example, the Beta distribution also raises this type of ValueErrors, but I couldn't find if it is being tested anywhere.



MarcoGorelli commented on 29 Sep 2020 • edited

Member



They do something similar in `pymc3/tests/test_starting.py`:

```
with raises(ValueError, match=r"Some variables not in the model: \['x2', 'y2'\]"):
    starting.allinmodel([[x2, y2], model1])
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([[x2, y1], model1])
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([x2], model1)
```

where `raises` was imported from `pytest` (see <https://docs.pytest.org/en/stable/assert.html> for more on this)

# What does a (first) PR look like?



ricardoV94 commented on 29 Sep 2020

Member Author 😊 ...

@MarcoGorelli That makes sense.

I am not sure how to do it. Could you direct me to a snippet where this type of test is implemented? For example, the distribution also raises this type of ValueError, but I couldn't find if it is being tested anywhere.



MarcoGorelli commented on 29 Sep 2020 • edited

Mem

They do something similar in `pymc3/tests/test_starting.py`:

```
with raises(ValueError, match=r"Some variables not in the model: \['x2', 'y2'\]"):
    starting.allinmodel([[x2, y2], model1])
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([[x2, y1], model1])
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([[x2], model1])
```

where `raises` was imported from `pytest` (see <https://docs.pytest.org/en/stable/assert.html> for more on this)



Add test for invalid initializations

```
pymc3/tests/test_distributions.py
...
def test_fun(value, mu, alpha):
    lambda value, p, n: sp.nbinom.logpmf(value, n, p),
)
def test_negative_binomial_init_fail(self):
    with Model():
        with pytest.raises(ValueError) as err:
            x = NegativeBinomial("x", mu=5)
        err.match("Incompatible parametrization. Must specify either alpha or n.")
        with pytest.raises(ValueError) as err:
            x = NegativeBinomial("x", n=2, alpha=2)
        err.match("Incompatible parametrization Can't specify both alpha and n.")
        with pytest.raises(ValueError) as err:
            x = NegativeBinomial("x", n=2)
        err.match("Incompatible parametrization. Must specify either mu or p.")
        with pytest.raises(ValueError) as err:
            x = NegativeBinomial("x", n=2, mu=2, p=5)
        err.match("Incompatible parametrization. Can't specify both mu and p.")
def test_laplace(self):
    self.pymc3_matches_scipy(
        Laplace,
```

✓ 58bf829

# What does a (first) PR look like?



ricardoV94 commented on 29 Sep 2020

Member

Author



@MarcoGorelli That makes sense.

I am not sure how to do it. Could you direct me to a snippet where this type of test is implemented? For example, the distribution also raises this type of ValueError, but I couldn't find if it is being tested anywhere.



MarcoGorelli commented on 29 Sep 2020 • edited

Mem

They do something similar in `pymc3/tests/test_starting.py`:

```
with raises(ValueError, match=r"Some variables not in the model: \['x2', 'y2'\]"):
    starting.allinmodel([x2, y2], model1)
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([x2, y1], model1)
with raises(ValueError, match=r"Some variables not in the model: \['x2'\]"):
    starting.allinmodel([x2], model1)
```

where `raises` was imported from `pytest` (see <https://docs.pytest.org/en/stable/assert.html> for more on this)



1



Add test for invalid initializations

```
pymc3/tests/test_distributions.py
-802,6 +802,24 def test_fun(value, mu, alpha):
802 802     lambda value, p, n: sp.nbinom.logpmf(value, n, p),
803 803     )
804 804
805 + def test_negative_binomial_init_fail(self):
806 +     with Model():
807 +         with pytest.raises(ValueError) as err:
808 +             x = NegativeBinomial("x", mu=5)
809 +             err.match("Incompatible parametrization. Must specify either alpha or n.")
810 +
811 +         with pytest.raises(ValueError) as err:
812 +             x = NegativeBinomial("x", n=2, alpha=2)
813 +             err.match("Incompatible parametrization Can't specify both alpha and n.")
814 +
815 +         with pytest.raises(ValueError) as err:
816 +             x = NegativeBinomial("x", n=2)
817 +             err.match("Incompatible parametrization. Must specify either mu or p.")
818 +
819 +         with pytest.raises(ValueError) as err:
820 +             x = NegativeBinomial("x", n=2, mu=2, p=5)
821 +             err.match("Incompatible parametrization. Can't specify both mu and p.")
822 +
805 823     def test_laplace(self):
806 824         self.pymc3_matches_scipy(
807 825             Laplace,
```

✓ 58bf829



What does a (first) PR look like?



What does a (first) PR look like?



# What does a (first) PR look like?



MarcoGorelli commented on 1 Oct 2020 • edited ▾

Member



@ricardoV94 nice work!

I think this can be simplified a bit using `pytest.mark.parametrize`, something like this (I've only written the first two test cases, but it should be easy to extend):

```
@pytest.mark.parametrize(
    "mu, n, alpha, expected",
    [
        (5, None, None, "Incompatible parametrization. Must specify either alpha or n."),
        (None, 2, 2, "Incompatible parametrization Can't specify both alpha and n."),
        # other two test cases go here
    ]
)
def test_negative_binomial_init_fail(self, mu, n, alpha, expected):
    with Model():
        with pytest.raises(ValueError, match=expected):
            x = NegativeBinomial("x", mu=mu, n=n, alpha=alpha)
```



# What does a (first) PR look like?

```
805 - def test_negative_binomial_init_fail(self):
806 +     @pytest.mark.parametrize(
807 +         "mu, p, alpha, n, expected",
808 +         [
809 +             (5, None, None, None, "Incompatible parametrization. Must specify either alpha or n."),
810 +             (None, .5, None, None, "Incompatible parametrization. Must specify either alpha or n."),
811 +             (None, None, None, None, "Incompatible parametrization. Must specify either alpha or n."),
812 +             (5, None, 2, 2, "Incompatible parametrization. Can't specify both alpha and n."),
813 +             (None, .5, 2, 2, "Incompatible parametrization. Can't specify both alpha and n."),
814 +             (5, .5, 2, 2, "Incompatible parametrization. Can't specify both alpha and n."),
815 +             (None, None, 2, None, "Incompatible parametrization. Must specify either mu or p."),
816 +             (None, None, None, 2, "Incompatible parametrization. Must specify either mu or p."),
817 +             (5, .5, 2, None, "Incompatible parametrization. Can't specify both mu and p."),
818 +             (5, .5, None, 2, "Incompatible parametrization. Can't specify both mu and p."),
819 +         ]
820 +     )
821 +     def test_negative_binomial_init_fail(self, mu, p, alpha, n, expected):
822 +         with Model():
823 +             with pytest.raises(ValueError) as err:
824 +                 x = NegativeBinomial("x", mu=5)
825 +                 err.match("Incompatible parametrization. Must specify either alpha or n.")
826 +
827 +             with pytest.raises(ValueError) as err:
828 +                 x = NegativeBinomial("x", n=2, alpha=2)
829 +                 err.match("Incompatible parametrization Can't specify both alpha and n.")
830 +
831 +             with pytest.raises(ValueError) as err:
832 +                 x = NegativeBinomial("x", n=2)
833 +                 err.match("Incompatible parametrization. Must specify either mu or p.")
834 +
835 +             with pytest.raises(ValueError) as err:
836 +                 x = NegativeBinomial("x", n=2, mu=2, p=.5)
837 +                 err.match("Incompatible parametrization. Can't specify both mu and p.")
838 +
839 +             with pytest.raises(ValueError, match=expected):
840 +                 NegativeBinomial("x", mu=mu, p=p, alpha=alpha, n=n)
```

# What does a (first) PR look like?



ricardoV94 commented on 1 Oct 2020

Member

Author



@MarcoGorelli Thanks for the suggestion. I refactored the code with the `pytest.mark.parametrize` as you suggested.

I also added a couple of tests for the different permutations of missing/over-specified parameters, do you think it is too much?



Minor change

✓ 5cbb09b

# What does a (first) PR look like?

The screenshot displays a vertical timeline of pull request activity. At the top, a comment from user **ricardoV94** is shown, dated 1 Oct 2020. The comment text reads: "@MarcoGorelli Thanks for the suggestion. I refactored the code with the `pytest.mark.parametrize` as you suggested. I also added a couple of tests for the different permutations of missing/over-specified parameters, do you think it is too much?". Below the comment is a commit icon and the text "Minor change" with a commit hash "5cbb09b". Next is a green checkmark icon followed by "twiecki approved these changes on 1 Oct 2020" and a "View changes" button. Below that is an eye icon followed by "twiecki marked this pull request as ready for review 16 months ago". At the bottom is a merge icon followed by "twiecki merged commit 6f3193d into pymc-devs:master on 1 Oct 2020" and "4 checks passed", with a "View details" button.

**ricardoV94** commented on 1 Oct 2020 Member Author

@**MarcoGorelli** Thanks for the suggestion. I refactored the code with the `pytest.mark.parametrize` as you suggested.

I also added a couple of tests for the different permutations of missing/over-specified parameters, do you think it is too much?

**Minor change** ✓ 5cbb09b

**twiecki** approved these changes on 1 Oct 2020 [View changes](#)

**twiecki** marked this pull request as ready for review 16 months ago

**twiecki** merged commit **6f3193d** into `pymc-devs:master` on 1 Oct 2020 [View details](#)  
4 checks passed

# What was it like?

- More work than expected
- Learned many new concepts:
  - unittesting
  - parametrizing
  - monkey-patching
  - code style checks
- There is a whole community to support **and challenge** you
- It was fun?





# I came back for more

26 Sep - 01 Oct

 **Add alternative parameterization to negative binomial distribution #4126** ✓  
#4134 by ricardoV94 was merged on 1 Oct 2020 • Approved

# I came back for more

16 Nov - 27 Feb

 **WIP: Add tt.nnet.softmax to math (#4226) ×**

#4229 by ricardoV94 was closed on 27 Feb 2021 • Changes requested

26 Sep - 01 Oct

 **Add alternative parameterization to negative binomial distribution #4126 ✓**

#4134 by ricardoV94 was merged on 1 Oct 2020 • Approved

# I came back for more

2 Dec - 5 Dec

 - **Fix regression caused by #4211** ✓ defects  
#4285 by ricardoV94 was merged on 5 Dec 2020 • Approved  3.10

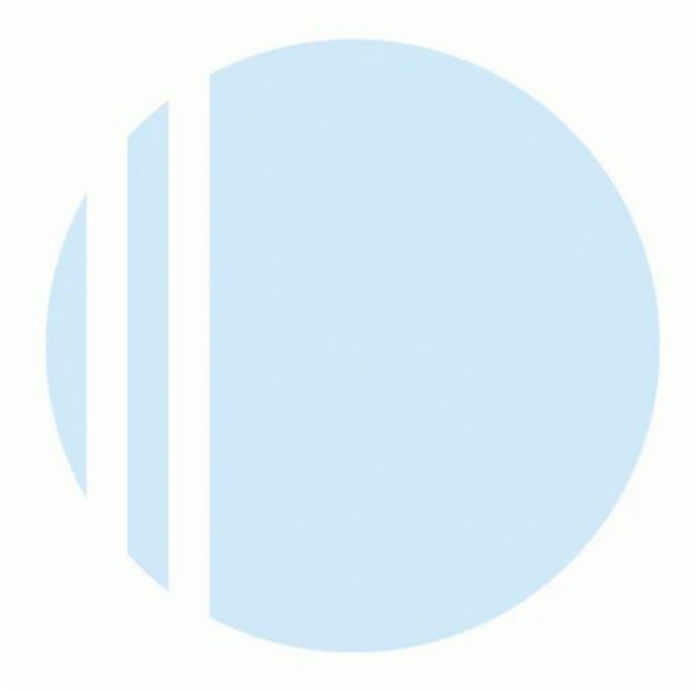
16 Nov - 27 Feb

 **WIP: Add tt.nnet.softmax to math (#4226)** ✗  
#4229 by ricardoV94 was closed on 27 Feb 2021 • Changes requested

26 Sep - 01 Oct

 **Add alternative parameterization to negative binomial distribution #4126** ✓  
#4134 by ricardoV94 was merged on 1 Oct 2020 • Approved

One year later they gave me the keys



One year later they gave me the keys



# One year later they gave me the keys



**Organizations**



# One year later they gave me the keys



Organizations



Organizations



# Reviewing a first time contributor PR



# Reviewing a first time contributor PR

Create helper `pm.draw()` to take draws from a given variable #5311

🔒 Closed

ricardoV94 opened this issue 22 days ago · 3 comments ·



ricardoV94 commented 22 days ago ·

edited ▾

Member



`eval()` should be aliased to something more intuitive like `sample()` or `draw()` where appropriate.

We have been using `eval` in our examples as a substitute to the old random method. This is just the standard Aesara debug feature that exists for any node, and shouldn't be used for more than that.

We can have a function wrapper that compiles a "proper" aesara function and takes a given number of draws. For analogy with V3 it could be named `pm.random`, but I like the `pm.draw` name better.

```
with pm.Model() as m:
    x = pm.Normal("x")

x_draws = pm.draw(x, draws=100)
```

# Reviewing a first time contributor PR

Create helper `pm.draw()` to take draws from a given variable #5311

🔒 Closed ricardoV94 opened this issue 22 days ago · 3 comments ·



ricardoV94 commented 22 days ago ·  
edited ▾

Member 😊 ⋮

`eval()` should be aliased to something more intuitive like `sample()` or `draw()` where appropriate.

We have been using `eval` in our examples as a substitute to the old random method. This is just the standard Aesara debug feature that exists for any node, and shouldn't be used for more than that.

We can have a function wrapper that compiles a "proper" aesara function and takes a given number of draws. For analogy with V3 it could be named `pm.random`, but I like the `pm.draw` name better.

```
with pm.Model() as m:
    x = pm.Normal("x")

x_draws = pm.draw(x, draws=100)
```



danhphan commented 19 days ago

Contributor 😊 ⋮

Hi @ricardoV94, I would like to work on this task. Could you please give some more detailed suggestions (like which file should we put the `pm.draw` function wrapper, and is there a similar wrapper function to follow). Thank you.



# Reviewing a first time contributor PR

```
2096 + def draw(  
2097 +     vars,  
2098 +     draws=500,  
2099 +     mode=None,  
2100 +     **kwargs  
2101 + ) -> Dict[str, np.ndarray]:  
2102 +     """Draw samples for one variable or a list of variables
```

```
2119 +     if vars is None:  
2120 +         raise AssertionError("Must include at least one variable")
```

This conversation was marked as resolved by **danhphan** [Show conversation](#)

```
2121 +  
2122 +     if isinstance(vars, tuple):  
2123 +         vars = list(vars)  
2124 +     elif not isinstance(vars, list):  
2125 +         vars = [vars]
```

This conversation was marked as resolved by **ricardoV94** [Show conversation](#)

```
2126 +  
2127 +     draw_fn = compile_pymc(inputs=[], outputs=vars, mode=mode, **kwargs)  
2128 +  
2129 +     values = zip(*(draw_fn() for _ in range(draws)))  
2130 +  
2131 +     names = [var.name for var in vars]  
2132 +     drawn_data = {k: np.stack(v) for k, v in zip(names, values)}  
2133 +  
2134 +     if drawn_data is None:  
2135 +         raise AssertionError("No variables drawn")
```

This conversation was marked as resolved by **ricardoV94** [Show conversation](#)

```
2136 +  
2137 +     return drawn_data  
2138 +
```

# Reviewing a first time contributor PR

```
2096 + def draw(  
2097 +     vars,  
2098 +     draws=500,  
2099 +     mode=None,  
2100 +     **kwargs  
2101 + ) -> Dict[str, np.ndarray]:  
2102 +     """Draw samples for one variable or a list of variables
```

```
2119 +     if vars is None:  
2120 +         raise AssertionError("Must include at least one variable")
```

This conversation was marked as resolved by **danhphan**

Show conversation

```
2121 +  
2122 +     if isinstance(vars, tuple):  
2123 +         vars = list(vars)  
2124 +     elif not isinstance(vars, list):  
2125 +         vars = [vars]
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2126 +  
2127 +     draw_fn = compile_pymc(inputs=[], outputs=vars, mode=mode, **kwargs)  
2128 +  
2129 +     values = zip(*(draw_fn() for _ in range(draws)))  
2130 +  
2131 +     names = [var.name for var in vars]  
2132 +     drawn_data = {k: np.stack(v) for k, v in zip(names, values)}  
2133 +  
2134 +     if drawn_data is None:  
2135 +         raise AssertionError("No variables drawn")
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2136 +  
2137 +     return drawn_data  
2138 +
```

2144 2144

```
2145 -     if vars is None:  
2146 -         raise AssertionError("Must include at least one variable")  
2147 -
```

```
2148 2145     if not isinstance(vars, (list, tuple)):  
2149 2146         vars = [vars]
```

2150 2147



# Reviewing a first time contributor PR

```
2096 + def draw(  
2097 +     vars,  
2098 +     draws=500,  
2099 +     mode=None,  
2100 +     **kwargs  
2101 + ) -> Dict[str, np.ndarray]:  
2102 +     """Draw samples for one variable or a list of variables
```

```
2119 +     if vars is None:  
2120 +         raise AssertionError("Must include at least one variable")
```

This conversation was marked as resolved by **danhphan**

Show conversation

```
2121 +  
2122 +     if isinstance(vars, tuple):  
2123 +         vars = list(vars)  
2124 +     elif not isinstance(vars, list):  
2125 +         vars = [vars]
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2126 +  
2127 +     draw_fn = compile_pymc(inputs=[], outputs=vars, mode=mode, **kwargs)  
2128 +  
2129 +     values = zip(*(draw_fn() for _ in range(draws)))  
2130 +  
2131 +     names = [var.name for var in vars]  
2132 +     drawn_data = {k: np.stack(v) for k, v in zip(names, values)}  
2133 +  
2134 +     if drawn_data is None:  
2135 +         raise AssertionError("No variables drawn")
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2136 +  
2137 +     return drawn_data  
2138 +
```

2144 2144

```
2145 -     if vars is None:  
2146 -         raise AssertionError("Must include at least one variable")  
2147 -  
2148 2145     if not isinstance(vars, (list, tuple)):  
2149 2146         vars = [vars]  
2150 2147
```

```
1266 -     npt.assert_raises(AssertionError, assert_array_equal, x_draws_1, x_draws_2)  
1258 +     assert not np.all(np.isclose(x_draws_1, x_draws_2))
```

# Reviewing a first time contributor PR

```
2096 + def draw(  
2097 +     vars,  
2098 +     draws=500,  
2099 +     mode=None,  
2100 +     **kwargs  
2101 + ) -> Dict[str, np.ndarray]:  
2102 +     """Draw samples for one variable or a list of variables
```

```
2119 +     if vars is None:  
2120 +         raise AssertionError("Must include at least one variable")
```

This conversation was marked as resolved by **danhphan**

Show conversation

```
2121 +  
2122 +     if isinstance(vars, tuple):  
2123 +         vars = list(vars)  
2124 +     elif not isinstance(vars, list):  
2125 +         vars = [vars]
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2126 +  
2127 +     draw_fn = compile_pymc(inputs=[], outputs=vars, mode=mode, **kwargs)  
2128 +  
2129 +     values = zip(*(draw_fn() for _ in range(draws)))  
2130 +  
2131 +     names = [var.name for var in vars]  
2132 +     drawn_data = {k: np.stack(v) for k, v in zip(names, values)}  
2133 +  
2134 +     if drawn_data is None:  
2135 +         raise AssertionError("No variables drawn")
```

This conversation was marked as resolved by **ricardoV94**

Show conversation

```
2136 +  
2137 +     return drawn_data  
2138 +
```

```
2144 2144  
2145 -     if vars is None:  
2146 -         raise AssertionError("Must include at least one variable")  
2147 -  
2148 2145     if not isinstance(vars, (list, tuple)):  
2149 2146         vars = [vars]  
2150 2147
```

```
1266 -     npt.assert_raises(AssertionError, assert_array_equal, x_draws_1, x_draws_2)  
1258 +     assert not np.all(np.isclose(x_draws_1, x_draws_2))
```



**ricardoV94** commented 14 days ago • edited ▾ Member 😊 ⋮

@**OriolAbril** It seems like the code block in the function docstrings is not rendering in the docs preview. Is there a formatting issue or is this expected?

<https://pymc--5340.readthedocs.build/en/5340/api/samplers.html#pymc.sampling.draw>

# Reviewing a first time contributor PR



OriolAbril reviewed 14 days ago

[View changes](#)

pymc/sampling.py Outdated Hide resolved

```
2120 +  
2121 + Examples  
2122 + -----  
2123 + .. code-block:: python
```



OriolAbril 14 days ago

Member 😊 ...

You need a blank line between the directive definition and the content



danhphan 14 days ago

Contributor Author 😊 ...

Hi, can you be more specific? Thanks



danhphan 14 days ago

Contributor Author 😊 ...

Just updated the code-block, hope that it should be alright now :)

pymc/sampling.py Outdated Hide resolved

```
2104 +  
2105 + Parameters  
2106 + -----  
2107 + vars : Union[Variable, Sequence[Variab
```



OriolAbril 14 days ago

Member 😊 ...

You shouldn't use type hints in the docstring but numpydoc formatting: <https://numpydoc.readthedocs.io/en/latest/format.html#parameters>



1

# Reviewing a first time contributor PR

```
pymc.sampling.draw(vars, draws=1, mode=None, **kwargs)
```

Draw samples for one variable or a list of variables

**Parameters:** vars

A variable or a list of variables for which to draw samples.

**draws :** *int*

Number of samples needed to draw. Defaults to 500.

**mode**

The mode used by `aesara.function` to compile the graph.

**\*\*kwargs**

Keyword arguments for `pymc.aesara.compile_pymc()`

**Returns:** `List[np.ndarray]`

A list of numpy arrays.

## Examples

```
import pymc as pm

# Draw samples for one variable
with pm.Model():
    x = pm.Normal("x")
    x_draws = pm.draw(x, draws=100)
    print(x_draws.shape)

# Draw 1000 samples for several variables
with pm.Model():
    x = pm.Normal("x")
    y = pm.Normal("y", shape=10)
    z = pm.Uniform("z", shape=5)
    num_draws = 1000
# Draw samples of a list variables
draws = pm.draw([x, y, z], draws=num_draws)
assert draws[0].shape == (num_draws,)
assert draws[1].shape == (num_draws, 10)
assert draws[2].shape == (num_draws, 5)
```



# Reviewing a first time contributor PR

```
pymc.sampling.draw(vars, draws=1, mode=None, **kwargs)
```

Draw samples for one variable or a list of variables

**Parameters:** vars

A variable or a list of variables for which to draw samples.

**draws :** *int*

Number of samples needed to draw. Defaults to 500.

**mode**

The mode used by `aesara.function` to compile the graph.

**\*\*kwargs**

Keyword arguments for `pymc.aesara.compile_pymc()`

**Returns:** `List[np.ndarray]`

A list of numpy arrays.

## Examples

```
import pymc as pm


# Draw samples for one variable
with pm.Model():
    x = pm.Normal("x")
    x_draws = pm.draw(x, draws=100)
    print(x_draws.shape)

# Draw 1000 samples for several variables
with pm.Model():
    x = pm.Normal("x")
    y = pm.Normal("y", shape=10)
    z = pm.Uniform("z", shape=5)
    num_draws = 1000
# Draw samples of a list variables
draws = pm.draw([x, y, z], draws=num_draws)
assert draws[0].shape == (num_draws,)
assert draws[1].shape == (num_draws, 10)
assert draws[2].shape == (num_draws, 5)
```

# Reviewing a first time contributor PR

pymc/sampling.py Outdated Hide resolved

```
2141 + if isinstance(vars, tuple):
2142 +     vars = list(vars)
2143 + elif not isinstance(vars, list):
2144 +     vars = [vars]
```

 **ricardoV94** 16 days ago Member 😊 ⋮

Does this work?

Suggested change ⓘ

```
- if isinstance(vars, tuple):
-     vars = list(vars)
- elif not isinstance(vars, list):
-     vars = [vars]
+ if not isinstance(vars, (list, tuple)):
+     vars = [vars]
```

# Reviewing a first time contributor PR

pymc/sampling.py Outdated Hide resolved

```
2141 + if isinstance(vars, tuple):
2142 +     vars = list(vars)
2143 + elif not isinstance(vars, list):
2144 +     vars = [vars]
```

 **ricardoV94** 16 days ago Member  

Does this work?

Suggested change 

```
- if isinstance(vars, tuple):
-     vars = list(vars)
- elif not isinstance(vars, list):
-     vars = [vars]
+ if not isinstance(vars, (list, tuple)):
+     vars = [vars]
```

 **ricardoV94** 16 days ago Member  

Actually we don't even need this. The outputs do not need to be in a list when we have a single variable

  1

# Reviewing a first time contributor PR

pymc/sampling.py Outdated Hide resolved

```
2141 +   if isinstance(vars, tuple):
2142 +       vars = list(vars)
2143 +   elif not isinstance(vars, list):
2144 +       vars = [vars]
```

 **ricardoV94** 16 days ago Member 😊 ⋮

Does this work?

Suggested change ?

```
-   if isinstance(vars, tuple):
-       vars = list(vars)
-   elif not isinstance(vars, list):
-       vars = [vars]
+   if not isinstance(vars, (list, tuple)):
+       vars = [vars]
```

 **ricardoV94** 16 days ago Member 😊 ⋮

Actually we don't even need this. The outputs do not need to be in a list when we have a single variable

😊 👍 1

 **danhphan** 16 days ago • edited Contributor Author 😊 ⋮

Hi yes, I have tested this code.

```
if not isinstance(vars, (list, tuple)):
    vars = [vars]
```

It will fail to the following test if `vars` is a tuple.

# Reviewing a first time contributor PR

pymc/sampling.py Outdated Hide resolved

```
2141 + if isinstance(vars, tuple):
2142 +     vars = list(vars)
2143 + elif not isinstance(vars, list):
2144 +     vars = [vars]
```

**ricardoV94** 16 days ago Member 😊 ⋮

Does this work?

Suggested change ⓘ

```
- if isinstance(vars, tuple):
-     vars = list(vars)
- elif not isinstance(vars, list):
-     vars = [vars]
+ if not isinstance(vars, (list, tuple)):
+     vars = [vars]
```

**ricardoV94** 16 days ago Member 😊 ⋮

Actually we don't even need this. The outputs do not need to be in a list when we have a single variable

😊 👍 1

**danhphan** 16 days ago • edited 👍

Hi yes, I have tested this code.

```
if not isinstance(vars, (list, tuple)):
    vars = [vars]
```

It will fail to the following test if `vars` is a tuple.

This due to the `pymc.aesara.compile_pymc(inputs, outputs)` function expect outputs as a list.  
`output_to_list = outputs if isinstance(outputs, list) else [outputs]`

# Reviewing a first time contributor PR

```
2 pymc/aesaraf.py
@@ -967,7 +967,7 @@ def compile_pymc(inputs, outputs, mode=None, **kwargs):
967 967
968 968     # Set the default update of a NoDistribution RNG so that it is automatically
969 969     # updated after every function call
970 -   output_to_list = outputs if isinstance(outputs, list) else [outputs]
970 +   output_to_list = outputs if isinstance(outputs, (list, tuple)) else [outputs]
971 971     for rv in (
972 972         node
973 973         for node in walk_model(output_to_list, walk_past_rvs=True)
```

```
2141 -   if isinstance(vars, tuple):
2142 -       vars = list(vars)
2143 -   elif not isinstance(vars, list):
2148 +   if not isinstance(vars, (list, tuple)):
2144 2149       vars = [vars]
```

# Reviewing a first time contributor PR

---

## Change internal variable names



**ricardoV94** committed 14 days ago

---

## Fix type hints



**ricardoV94** committed 14 days ago

## Change internal variable names

 ricardoV94 committed 14 days ago

## Fix type hints

 ricardoV94 committed 14 days ago

## Return scalar variables instead of 1D

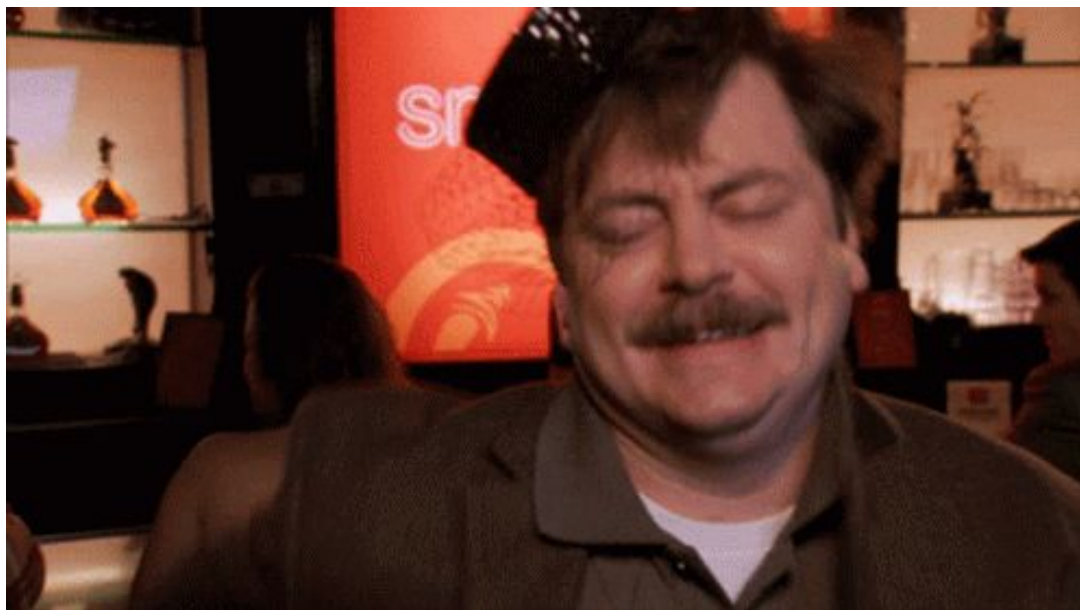
 ricardoV94 committed 14 days ago

```
-----
2145 2145
2146 -   if not isinstance(vars, (list, tuple)):
2147 -       vars = [vars]
2148 -
2149 2146     draw_fn = compile_pymc(inputs=[], outputs=vars, mode=mode, **kwargs)
2150 -     drawn_values = zip(*(draw_fn() for _ in range(draws)))
2151 -     drawn_values = [np.stack(v) for v in drawn_values]
2152 2147
2153 -     # If only one variable, return the numpy array instead of a list of numpy arrays
2154 2148     if draws == 1:
2155 -         return drawn_values[0]
2156 -         return drawn_values
2149 +         return draw_fn()
2150 +
2151 +     # Single variable output
2152 +     if not isinstance(vars, (list, tuple)):
2153 +         drawn_values = (draw_fn() for _ in range(draws))
2154 +         return np.stack(drawn_values)
2155 +
2156 +     # Multiple variable output
2157 +     drawn_values = zip(*(draw_fn() for _ in range(draws)))
2158 +     return [np.stack(v) for v in drawn_values]
2157 2159
```



What is it like?

What is it like?



What is it like?

**Still figuring it out**

# So you want to be an OSS contributor?

- Find a project you care about
- Engage with the community
- Be open to challenges
- Be ready to
  - learn
  - be wrong
  - make your case
- Take responsibility seriously
  - Your code will be used by many others
- Be patient
- Be polite

Thank you!