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# **Dict Minimize Documentation**

**Dict Minimize Team**

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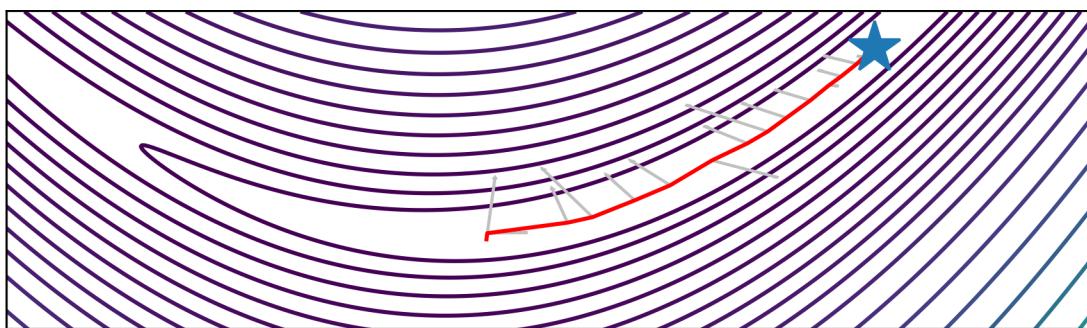


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## THE DICT MINIMIZE PACKAGE

Access `scipy` optimizers from your favorite deep learning framework.

### 1.1 Installation

Only Python`>=3.6` is officially supported, but older versions of Python likely work as well.

The core package itself can be installed with:

```
pip install dict_minimize
```

To also get the dependencies for all the supported frameworks (torch, JAX, tensorflow) in the README install with

```
pip install dict_minimize[framework]
```

See the [GitHub](#), [PyPI](#), and [Read the Docs](#).

### 1.2 Example Usage

In these examples we optimize a modified `Rosenbrock` function. However, the arguments have been split into two chunks and stored as two entries in a dictionary. This is to illustrate how this package optimizes *dictionaries* of (tensor) parameters rather than vectors. We also pass in an extra `shift` argument to demonstrate how `minimize` allows extra constant arguments to be passed into the objective.

#### 1.2.1 PyTorch

```
import torch
from dict_minimize.torch_api import minimize

def rosen_obj(params, shift):
    """Based on augmented Rosenbrock from botorch."""
    X, Y = params["x_half_a"], params["x_half_b"]
    X = X - shift
    Y = Y - shift
    obj = 100 * (X[1] - X[0] ** 2) ** 2 + 100 * (Y[1] - Y[0] ** 2) ** 2
    return obj

def d_rosen_obj(params, shift):
    obj = rosen_obj(params, shift=shift)
```

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```
da, db = torch.autograd.grad(obj, [params["x_half_a"], params["x_half_b"]])
d_obj = OrderedDict([('x_half_a', da), ('x_half_b', db)])
return obj, d_obj

torch.manual_seed(123)

n_a = 2
n_b = 2
shift = -1.0

params = OrderedDict([('x_half_a', torch.randn((n_a,))), ('x_half_b', torch.randn((n_b,))))]

params = minimize(d_rosen_obj, params, args=(shift,), method="L-BFGS-B", options={
    "disp": True})
```

## 1.2.2 TensorFlow

```
import tensorflow as tf
from dict_minimize.tensorflow_api import minimize

def rosen_obj(params, shift):
    """Based on augmented Rosenbrock from botorch."""
    X, Y = params["x_half_a"], params["x_half_b"]
    X = X - shift
    Y = Y - shift
    obj = 100 * (X[1] - X[0] ** 2) ** 2 + 100 * (Y[1] - Y[0] ** 2) ** 2
    return obj

def d_rosen_obj(params, shift):
    with tf.GradientTape(persistent=True) as t:
        t.watch(params["x_half_a"])
        t.watch(params["x_half_b"])

        obj = rosen_obj(params, shift=shift)

        da = t.gradient(obj, params["x_half_a"])
        db = t.gradient(obj, params["x_half_b"])
        d_obj = OrderedDict([('x_half_a', da), ('x_half_b', db)])
    del t # Explicitly drop the reference to the tape
    return obj, d_obj

tf.random.set_seed(123)

n_a = 2
n_b = 2
shift = -1.0

params = OrderedDict([('x_half_a', tf.random.normal((n_a,))), ('x_half_b', tf.random.normal((n_b,))))]

params = minimize(d_rosen_obj, params, args=(shift,), method="L-BFGS-B", options={
    "disp": True})
```

### 1.2.3 NumPy

```

import numpy as np
from scipy.optimize import rosen, rosen_der
from dict_minimize.numpy_api import minimize

def rosen_obj(params, shift):
    val = rosen(params["x_half_a"] - shift) + rosen(params["x_half_b"] - shift)

    dval = OrderedDict(
        [
            ("x_half_a", rosen_der(params["x_half_a"] - shift)),
            ("x_half_b", rosen_der(params["x_half_b"] - shift)),
        ]
    )
    return val, dval

np.random.seed(0)

n_a = 3
n_b = 5
shift = -1.0

params = OrderedDict([("x_half_a", np.random.randn(n_a)), ("x_half_b", np.random.
    ↪randn(n_b))])

params = minimize(rosen_obj, params, args=(shift,), method="L-BFGS-B", options={"disp
    ↪": True})

```

### 1.2.4 JAX

```

from jax import random, value_and_grad
import jax.numpy as np
from dict_minimize.jax_api import minimize

def rosen(x):
    r = np.sum(100.0 * (x[1:] - x[:-1] ** 2.0) ** 2.0 + (1 - x[:-1]) ** 2.0, axis=0)
    return r

def rosen_obj(params, shift):
    val = rosen(params["x_half_a"] - shift) + rosen(params["x_half_b"] - shift)
    return val

n_a = 3
n_b = 5
shift = -1.0

# Jax makes it this simple:
d_rosen_obj = value_and_grad(rosen_obj, argnums=0)

# Setup randomness in JAX
key = random.PRNGKey(0)
key, subkey_a = random.split(key)
key, subkey_b = random.split(key)

```

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```
params = OrderedDict(
    [ ("x_half_a", random.normal(subkey_a, shape=(n_a,))), ("x_half_b", random.
    ↪normal(subkey_b, shape=(n_b,)))]

params = minimize(d_rosen_obj, params, args=(shift,), method="L-BFGS-B", options={
    ↪"disp": True})
```

## 1.3 Contributing

The following instructions have been tested with Python 3.7.4 on Mac OS (10.14.6).

### 1.3.1 Install in editable mode

First, define the variables for the paths we will use:

```
GIT=/path/to/where/you/put/repos
ENVS=/path/to/where/you/put/virtualenvs
```

Then clone the repo in your git directory \$GIT:

```
cd $GIT
git clone https://github.com/twitter/dict_minimize.git
```

Inside your virtual environments folder \$ENVS, make the environment:

```
cd $ENVS
virtualenv dict_minimize --python=python3.7
source $ENVS/dict_minimize/bin/activate
```

Now we can install the pip dependencies. Move back into your git directory and run

```
cd $GIT/dict_minimize
pip install -r requirements/base.txt
pip install -e . # Install the package itself
```

### 1.3.2 Contributor tools

First, we need to setup some needed tools:

```
cd $ENVS
virtualenv dict_minimize_tools --python=python3.7
source $ENVS/dict_minimize_tools/bin/activate
pip install -r $GIT/dict_minimize/requirements/tools.txt
```

To install the pre-commit hooks for contributing run (in the dict\_minimize\_tools environment):

```
cd $GIT/dict_minimize
pre-commit install
```

To rebuild the requirements, we can run:

```
cd $GIT/dict_minimize

# Check if there any discrepancies in the .in files
pipreqs dict_minimize/core/ --diff requirements/base.in
pipreqs dict_minimize/ --diff requirements/frameworks.in
pipreqs tests/ --diff requirements/tests.in
pipreqs docs/ --diff requirements/docs.in

# Regenerate the .txt files from .in files
pip-compile-multi --no-upgrade
```

### 1.3.3 Generating the documentation

First setup the environment for building with Sphinx:

```
cd $ENVS
virtualenv dict_minimize_docs --python=python3.7
source $ENVS/dict_minimize_docs/bin/activate
pip install -r $GIT/dict_minimize/requirements/docs.txt
```

Then we can do the build:

```
cd $GIT/dict_minimize/docs
make all
open _build/html/index.html
```

Documentation will be available in all formats in `Makefile`. Use `make html` to only generate the HTML documentation.

### 1.3.4 Running the tests

The tests for this package can be run with:

```
cd $GIT/dict_minimize
./local_test.sh
```

The script creates an environment using the requirements found in `requirements/test.txt`. A code coverage report will also be produced in `$GIT/dict_minimize/htmlcov/index.html`.

### 1.3.5 Deployment

The wheel (tar ball) for deployment as a pip installable package can be built using the script:

```
cd $GIT/dict_minimize/
./build_wheel.sh
```

This script will only run if the git repo is clean, i.e., first run `git clean -x -ff -d`.

## 1.4 Links

The [source](#) is hosted on GitHub.

The [documentation](#) is hosted at Read the Docs.

Installable from [PyPI](#).

## 1.5 License

This project is licensed under the Apache 2 License - see the LICENSE file for details.

## CODE OVERVIEW

### 2.1 PyTorch API

```
dict_minimize.torch_api.minimize(fun: Callable, x0_dict: collections.OrderedDict, *, lb_dict: Optional[collections.OrderedDict] = None, ub_dict: Optional[collections.OrderedDict] = None, args: Sequence = (), method: Optional[str] = None, tol: Optional[float] = None, callback: Optional[Callable] = None, options: Optional[dict] = None) → collections.OrderedDict
```

Minimization of a scalar function with a dictionary of variables as the input. It can interface to functions written for *torch*.

This is a wrapper around *scipy.optimize.minimize*.

#### Parameters

- **fun** (*callable*) – The objective function to be minimized, in the form of `fun(x, *args) -> (float, OrderedDict)` where *x* is an *OrderedDict* in the format of *x0\_dict*, and *args* is a tuple of the fixed parameters needed to completely specify the function. The second returned variable is the gradients. It should be an *OrderedDict* with the same keys and shapes as *x*. The values should be *torch Tensor*.
- **x0\_dict** (*OrderedDict*) – Initial guess. Dictionary of variables from variable name to *torch* variables.
- **lb\_dict** (*OrderedDict*) – Dictionary with same keys and shapes as *x0\_dict* with lower bounds for each variable. Set to *None* in an unconstrained problem.
- **ub\_dict** (*OrderedDict*) – Dictionary with same keys and shapes as *x0\_dict* with upper bounds for each variable. Set to *None* in an unconstrained problem.
- **args** (*tuple*) – Extra arguments passed to the objective function.
- **method** (*str*) – Type of solver. Should be one of: CG, BFGS, L-BFGS-B, TNC, SLSQP, or trust-constr. If not given, chosen to be one of BFGS, L-BFGS-B, SLSQP, depending if the problem has bounds. Note, only L-BFGS-B, TNC, SLSQP seem to strictly respect the bounds *lb\_dict* and *ub\_dict*.
- **tol** (*float*) – Tolerance for termination. For detailed control, use solver-specific options.
- **callback** (*callable*) – Called after each iteration. The signature is: `callback(xk)` where *xk* is the current parameter as an *OrderedDict* with the same form as the final solution *x*.
- **options** (*dict*) – A dictionary of solver options. All methods accept the following generic options: maxiter : int Maximum number of iterations to perform. Depending on

the method each iteration may use several function evaluations. `disp` : bool Set to `True` to print convergence messages.

**Returns** Final solution found by the optimizer. It has the same keys and shapes as `x0_dict`.

**Return type** `x` (`OrderedDict`)

## 2.2 TensorFlow API

```
dict_minimize.tensorflow_api.minimize(fun: Callable, x0_dict: collections.OrderedDict, *,  
    lb_dict: Optional[collections.OrderedDict] = None,  
    ub_dict: Optional[collections.OrderedDict] = None,  
    args: Sequence = (), method: Optional[str] =  
        None, tol: Optional[float] = None, callback: Optional[Callable] = None, options: Optional[dict] =  
        None) → collections.OrderedDict
```

Minimization of a scalar function with a dictionary of variables as the input. It can interface to functions written for `tensorflow`.

This is a wrapper around `scipy.optimize.minimize`.

### Parameters

- **fun** (`callable`) – The objective function to be minimized, in the form of `fun(x, *args) -> (float, OrderedDict)` where `x` is an `OrderedDict` in the format of `x0_dict`, and `args` is a tuple of the fixed parameters needed to completely specify the function. The second returned variable is the gradients. It should be an `OrderedDict` with the same keys and shapes as `x`. The values should be `tensorflow` variables.
- **x0\_dict** (`OrderedDict`) – Initial guess. Dictionary of variables from variable name to `tensorflow` variables.
- **lb\_dict** (`OrderedDict`) – Dictionary with same keys and shapes as `x0_dict` with lower bounds for each variable. Set to `None` in an unconstrained problem.
- **ub\_dict** (`OrderedDict`) – Dictionary with same keys and shapes as `x0_dict` with upper bounds for each variable. Set to `None` in an unconstrained problem.
- **args** (`tuple`) – Extra arguments passed to the objective function.
- **method** (`str`) – Type of solver. Should be one of: CG, BFGS, L-BFGS-B, TNC, SLSQP, or trust-constr. If not given, chosen to be one of BFGS, L-BFGS-B, SLSQP, depending if the problem has bounds. Note, only L-BFGS-B, TNC, SLSQP seem to strictly respect the bounds `lb_dict` and `ub_dict`.
- **tol** (`float`) – Tolerance for termination. For detailed control, use solver-specific options.
- **callback** (`callable`) – Called after each iteration. The signature is: `callback(xk)` where `xk` is the current parameter as an `OrderedDict` with the same form as the final solution `x`.
- **options** (`dict`) – A dictionary of solver options. All methods accept the following generic options: maxiter : int Maximum number of iterations to perform. Depending on the method each iteration may use several function evaluations. `disp` : bool Set to `True` to print convergence messages.

**Returns** Final solution found by the optimizer. It has the same keys and shapes as `x0_dict`.

**Return type** `x` (`OrderedDict`)

## 2.3 NumPy API

```
dict_minimize.numpy_api.minimize(fun: Callable, x0_dict: collections.OrderedDict, *, lb_dict: Optional[collections.OrderedDict] = None, ub_dict: Optional[collections.OrderedDict] = None, args: Sequence = (), method: Optional[str] = None, tol: Optional[float] = None, callback: Optional[Callable] = None, options: Optional[dict] = None) → collections.OrderedDict
```

Minimization of a scalar function with a dictionary of variables as the input.

This is a wrapper around `scipy.optimize.minimize`.

### Parameters

- **fun** (`callable`) – The objective function to be minimized, in the form of `fun(x, *args) -> (float, OrderedDict)` where `x` is an `OrderedDict` in the format of `x0_dict`, and `args` is a tuple of the fixed parameters needed to completely specify the function. The second returned variable is the gradients. It should be an `OrderedDict` with the same keys and shapes as `x`. The values should be `numpy ndarray` variables.
- **x0\_dict** (`OrderedDict`) – Initial guess. Dictionary of variables from variable name to `numpy` variables.
- **lb\_dict** (`OrderedDict`) – Dictionary with same keys and shapes as `x0_dict` with lower bounds for each variable. Set to `None` in an unconstrained problem.
- **ub\_dict** (`OrderedDict`) – Dictionary with same keys and shapes as `x0_dict` with upper bounds for each variable. Set to `None` in an unconstrained problem.
- **args** (`tuple`) – Extra arguments passed to the objective function.
- **method** (`str`) – Type of solver. Should be one of: CG, BFGS, L-BFGS-B, TNC, SLSQP, or trust-constr. If not given, chosen to be one of BFGS, L-BFGS-B, SLSQP, depending if the problem has bounds. Note, only L-BFGS-B, TNC, SLSQP seem to strictly respect the bounds `lb_dict` and `ub_dict`.
- **tol** (`float`) – Tolerance for termination. For detailed control, use solver-specific options.
- **callback** (`callable`) – Called after each iteration. The signature is: `callback(xk)` where `xk` is the current parameter as an `OrderedDict` with the same form as the final solution `x`.
- **options** (`dict`) – A dictionary of solver options. All methods accept the following generic options: maxiter : int Maximum number of iterations to perform. Depending on the method each iteration may use several function evaluations. disp : bool Set to `True` to print convergence messages.

**Returns** Final solution found by the optimizer. It has the same keys and shapes as `x0_dict`.

**Return type** `x` (`OrderedDict`)

## 2.4 JAX API

```
dict_minimize.jax_api.minimize(fun: Callable, x0_dict: collections.OrderedDict, *, lb_dict: Optional[collections.OrderedDict] = None, ub_dict: Optional[collections.OrderedDict] = None, args: Sequence = (), method: Optional[str] = None, tol: Optional[float] = None, callback: Optional[Callable] = None, options: Optional[dict] = None) → collections.OrderedDict
```

Minimization of a scalar function with a dictionary of variables as the input. It can interface to functions written for *jax*.

This is a wrapper around *scipy.optimize.minimize*.

### Parameters

- **fun** (*callable*) – The objective function to be minimized, in the form of `fun(x, *args) -> (float, OrderedDict)` where *x* is an *OrderedDict* in the format of *x0\_dict*, and *args* is a tuple of the fixed parameters needed to completely specify the function. The second returned variable is the gradients. It should be an *OrderedDict* with the same keys and shapes as *x*. The values should be *jax ndarray* variables.
- **x0\_dict** (*OrderedDict*) – Initial guess. Dictionary of variables from variable name to *jax ndarray*.
- **lb\_dict** (*OrderedDict*) – Dictionary with same keys and shapes as *x0\_dict* with lower bounds for each variable. Set to *None* in an unconstrained problem.
- **ub\_dict** (*OrderedDict*) – Dictionary with same keys and shapes as *x0\_dict* with upper bounds for each variable. Set to *None* in an unconstrained problem.
- **args** (*tuple*) – Extra arguments passed to the objective function.
- **method** (*str*) – Type of solver. Should be one of: CG, BFGS, L-BFGS-B, TNC, SLSQP, or trust-constr. If not given, chosen to be one of BFGS, L-BFGS-B, SLSQP, depending if the problem has bounds. Note, only L-BFGS-B, TNC, SLSQP seem to strictly respect the bounds *lb\_dict* and *ub\_dict*.
- **tol** (*float*) – Tolerance for termination. For detailed control, use solver-specific options.
- **callback** (*callable*) – Called after each iteration. The signature is: `callback(xk)` where *xk* is the current parameter as an *OrderedDict* with the same form as the final solution *x*.
- **options** (*dict*) – A dictionary of solver options. All methods accept the following generic options: maxiter : int Maximum number of iterations to perform. Depending on the method each iteration may use several function evaluations. disp : bool Set to *True* to print convergence messages.

**Returns** Final solution found by the optimizer. It has the same keys and shapes as *x0\_dict*.

**Return type** *x* (*OrderedDict*)

## 2.5 Internals

The dict minimize core routines.

Utility for minimization of a scalar function with a dictionary of variables as the input. It can interface to functions written outside of *numpy* (e.g., *tensorflow*, *torch* or *jax*).

This is a wrapper around *scipy.optimize.minimize*.



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**CREDITS**

### **3.1 Development lead**

Ryan Turner - Twitter (rdturnermtl)

### **3.2 Contributors**

- Gerard Casas Saez - Twitter (casassg)



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```