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# XGBTune Documentation

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**R. Picard**

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## Contents:

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<b>1</b>	<b>Get Started</b>	<b>1</b>
<b>2</b>	<b>Parameters</b>	<b>3</b>
2.1	Configuring cross-validation . . . . .	3
2.2	Using a validation set . . . . .	3
2.3	Providing your own hyper-parameter search space . . . . .	4
<b>3</b>	<b>Reference</b>	<b>5</b>
<b>4</b>	<b>Indices and tables</b>	<b>7</b>
	<b>Python Module Index</b>	<b>9</b>
	<b>Index</b>	<b>11</b>



# CHAPTER 1

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## Get Started

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XGBTune uses a default configuration that should work in most cases. This allows to tune a model by using cross compilation‡ very easily:

```
from xgbtune import tune_xgb_model  
params, round_count = tune_xgb_model(params, x_train, y_train)
```

params contains your xgboost configuration and initial parameters. x\_train and y\_train are your datasets with labels. The result is a tuple with the tuned parameters and the optimal number of booster rounds.



# CHAPTER 2

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## Parameters

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### 2.1 Configuring cross-validation

The following parameters are directly forwarded to the xgboost.cv function call:

- nfold: number of folds
- stratified: Perform stratified sampling
- folds: Sklearn KFolds or StratifiedKFolds object
- shuffle: shuffle data

These parameters must be provided as keyword parameters. Here is an example of tuning with 5 folds and without shuffling:

```
from xgbtune import tune_xgb_model

params, round_count = tune_xgb_model(
    params,
    x_train, y_train,
    folds=5, shuffle=False)
```

### 2.2 Using a validation set

It is possible to use a validation set instead of cross-validation:

```
from xgbtune import tune_xgb_model

params, round_count = tune_xgb_model(
    params,
    x_train, y_train,
    x_val, y_val)
```

## 2.3 Providing your own hyper-parameter search space

The tune\_params is a dict that can be used to overload the default search space for each parameter. A search space is provided as a list, with the dict key name corresponding to the xgboost parameter: max\_depth, min\_child\_weight, gamma, subsample, colsample\_bytree, alpha, lambda, seed.

Here is an example of tuning with custom alpha search space:

```
from xgbtune import tune_xgb_model

params, round_count = tune_xgb_model(
    params,
    x_train, y_train,
    {"alpha": [(i/10.0,) for i in range(0,11)]})
```

# CHAPTER 3

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## Reference

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```
xgbtune.tune.tune_xgb_model(params, x_train, y_train, x_val=None, y_val=None, nfold=3,  
stratified=False, folds=None, shuffle=True, tune_params={},  
max_round_count=5000, loss_compare=<built-in function lt>,  
pass_count=2, verbose=True)
```

Tunes a XGBoost model

### Examples

```
>>> params, round_count = tune_xgb_model(x, y, x_val, y_val, model_params)
```

### Parameters

- **params** – A dictionary with the base xgboost parameters to use
- **x\_train** – Train set
- **y\_train** – Train labels
- **x\_val** – Validation set
- **y\_val** – Validation labels
- **nfold** – Number of folds for cv
- **stratified** – Perform stratified sampling
- **folds** – Sklearn KFolds or StratifiedKFolds object
- **shuffle** – shuffle data on cross validation
- **tune\_params** – dictionary containing list of values to test to each parameter
- **max\_round\_count** – Maximum number of rounds during training
- **pass\_count** – Number of tuning pass to do

**Returns** A tuple of tuned parameters and round count.



# CHAPTER 4

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## Indices and tables

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- genindex
- modindex
- search



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## Python Module Index

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X

xgbtune.tune, 5



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## Index

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### T

`tune_xgb_model()` (*in module `xgbtune.tune`*), 5

### X

`xgbtune.tune` (*module*), 5