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# **XGBTune Documentation**

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# 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<sup>‡</sup> 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.



## 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)]})
```



```
xgbtune.tune.tune_xgb_model(params, x_train, y_train, x_val=None, y_val=None, ifold=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
- **ifold** – Number of folds for cv
- **stratified** – Perform stratified sampling
- **folds** – Sklearn KFold or StratifiedKFold 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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