

FEDOT Cheat Sheet

Quick Start

```
# Installation
pip install fedot

# Basic usage example
from fedot.api.main import Fedot
model = Fedot(problem='classification')
model.fit(features=X_train, target=y_train)
prediction = model.predict(features=X_test)
```

Core Concepts

- **Pipelines:** Computational graphs that combine data operations and models
- **Nodes:** Individual operations within a pipeline (preprocessing or models)
- **Edges:** Connections between nodes that define data flow
- **Composer:** Algorithm that creates pipeline structures
- **Tuner:** Optimizes hyperparameters of models in a pipeline

Common Methods

Automatic Pipeline Construction

Method	Description	Example
<code>Fedot()</code>	Initialize the framework	<code>model = Fedot(problem='regression')</code>
<code>fit()</code>	Train the model	<code>model.fit(features=X, target=y)</code>
<code>predict()</code>	Make predictions	<code>predictions = model.predict(features=X_test)</code>
<code>save()</code>	Save the model	<code>model.save('model.pkl')</code>
<code>load()</code>	Load a saved model	<code>model = Fedot.load('model.pkl')</code>

Manual Pipeline Construction

Method	Description	Example
<code>Pipeline()</code>	Create empty pipeline	<code>pipeline = Pipeline()</code>
<code>Node()</code>	Create operation node	<code>node = Node(operation_type='xgboost')</code>

Method	Description	Example
add_node()	Add node to pipeline	pipeline.add_node(node)
add_edge()	Connect nodes	pipeline.add_edge(node_from, node_to)
fit()	Train pipeline	pipeline.fit(input_data=train_data)
predict()	Get predictions	predictions = pipeline.predict(input_data=test_data)

Configuration Options

```
# Configuring FEDOT
model = Fedot(
    problem='classification',
    preset='best_quality', # Options: 'fast', 'stable', 'best_quality'
    timeout=5, # Minutes for optimization
    n_jobs=-1, # CPU cores to use (-1 = all)
    cv_folds=5, # Cross-validation folds
    random_state=42, # Random seed
    metric=['accuracy']) # Set metrics to optimize
)
```

Available Models and Operations

Primary Nodes (Data Processing)

scaling • normalization • pca • kernel_pca • fast_ica • polynomial_features
• lagged • smoothing

Secondary Nodes (Models)

linear • ridge • lasso • rf • xgboost • lgbm • catboost • knn • dt •
mlp

Common Patterns

Time Series Forecasting

```
# Time series forecasting
model = Fedot(problem='ts_forecasting', forecast_length=10)
train_input, test_input = train_test_data_setup(dataframe)
model.fit(train_input)
forecast = model.predict(test_input)
```

Troubleshooting Tips

- **Memory Error:** Reduce `pop_size` and `num_of_generations` parameters
- **Poor time series performance:** Add lagged features with lagged operation nodes
- **Slow execution:** Set `preset='fast'` or reduce optimization parameters
- **Overfitting:** Use cross-validation or simpler pipelines
- **Model not improving:** Try different metrics or increase optimization time

Resources

- Documentation: fedot.readthedocs.io
- GitHub: github.com/aimclib/FEDOT