

pyTFA cheat sheet

Getting started:

```
import pytfa
```

```
from math import *
```

```
from pytfa.io import import_matlab_model, load_thermoDB
```

Loading data:

```
cobra_model = import_matlab_model('model.mat')
```

```
thermo_data = load_thermoDB('../data/thermo_data.thermodb')
```

Create Thermo model:

```
tfa_model = pytfa.ThermoModel(thermo_data, cobra_model)
```

Object: [ThermoModel](#)

Important Attributes		
reactions	DictList of reaction objects	FBA
metabolites	DictList of metabolite objects	
log_concentration	DictList of variables for logarithmic concentrations	TFA
deltaG	DictList of variables for free energies	
deltaGstd	DictList of variables for standard free energies	

Object: [Reaction](#)

Important Attributes	
forward_variable	Variable object for the forward flux
reverse_variable	Variable object for the backward flux
bounds	Bounds on the Net Flux
lower_bound	Lower bound of the net flux
upper_bound	Upper bound of the net flux

Object: [LogConcentration](#)

Important Attributes	
variable	Variable object

Object: [Variable](#)

Important Attributes	
bounds	Variable bounds

Object: [DictList](#)

Important Attributes	
<code>get_by_id()</code>	Method to get an entry by its id e.g rxn_id for deltaG or met_id for logarithmic concentration

ThermoModel

Methods defined here:

__deepcopy__(self, memo)

:param memo:
:return:

__init__(self, thermo_data, model, name=None, temperature=298.15, min_ph=3, max_ph=9)

:param float temperature: the temperature (K) at which to perform the calculations
:param dict thermo_data: The thermodynamic database
:type temperature: float

add_constraint(self, kind, hook, expr, **kwargs)

Add a new constraint to a COBRApy cobra_model

:param cobra.core.model.[Model](#) model: The COBRApy cobra_model
:param string, cobra.Reaction hook: Either a string representing the name of the variable to add to the cobra_model, or a reaction object if the kind allows it
:param sympy.core.expr.Expr expr: The expression of the constraint

:returns: The created constraint
:rtype: optlang.interface.Constraint

add_variable(self, kind, hook, **kwargs)

Add a new variable to a COBRApy cobra_model.

:param cobra.core.model.[Model](#) model: The COBRApy cobra_model
:param string, cobra.Reaction hook: Either a string representing the name of the variable to add to the cobra_model, or a reaction object if the kind allows it

:returns: The created variable
:rtype: optlang.interface.Variable

convert(self, add_potentials=False, add_displacement=False, verbose=True)

Converts a cobra_model into a tFBA ready cobra_model by adding the thermodynamic constraints required

.. warning::
This function requires you to have already called
:func:`~.pytfa.[ThermoModel](#).prepare`, otherwise it will raise an Exception !

copy(self)

get_constraints_of_type(self, constraint_type)

Convenience function that takes as input a constraint class and returns all its instances within the cobra_model

```

        :param constraint_type:
        :return:
get_primal(self, vartype, index_by_reactions=False)
    Returns the primal value of the cobra_model for variables of a given
    type

    :param vartype: Class of variable. Ex: pytfpa.optim.variables.ThermoDisplacement
    :return:
get_solution(self)
    Overrides the cobra.core.solution method, to also get the supplementary
    variables we added to the cobra_model
    :return:
get_variables_of_type(self, variable_type)
    Convenience function that takes as input a variable class and returns
    all its instances within the cobra_model

    :param variable_type:
    :return:
normalize_reactions(self)
    Find reactions with important stoichiometry and normalizes them
    :return:
optimize(self, objective_sense=None, **kwargs)
    Call the Model.optimize function (which really is but an interface to
    the
    solver's. Catches SolverError in the case of no solutions. Passes down
    supplementary keyword arguments (see cobra.core.Model.optimize)
    :type objective_sense: 'min' or 'max'
prepare(self)
    Prepares a COBRA toolbox cobra_model for TFBA analysis by doing the following:

    1. checks if a reaction is a transport reaction
    2. checks the ReactionDB for Gibbs energies of formation of metabolites
    3. computes the Gibbs energies of reactions
print_info(self)
    Print information and counts for the cobra_model
    :return:
regenerate_constraints(self)
    Generates references to the cobra_model's constraints in self._constraints
    as tab-searchable attributes of the thermo cobra_model
    :return:
regenerate_variables(self)
    Generates references to the cobra_model's constraints in self._variables
    as tab-searchable attributes of the thermo cobra_model
    :return:
remove_constraint(self, cons)
    Removes a constraint

    :param cons:
    :return:
remove_variable(self, var)

```

Removes a variable

:param var:
:return:

repair(self)

Updates references to variables and constraints
:return:

Methods inherited from [cobra.core.model.Model](#):

__add__(self, other_model)

Add the content of another model to this model (+).

The model is copied as a new object, with a new model identifier, and copies of all the reactions in the other model are added to this model. The objective is the sum of the objective expressions for the two models.

__enter__(self)

Record all future changes to the model, undoing them when a call to `__exit__` is received

__exit__(self, type, value, traceback)

Pop the top context manager and trigger the undo functions

__getstate__(self)

Get state for serialization.

Ensures that the context stack is cleared prior to serialization, since partial functions cannot be pickled reliably.

__iadd__(self, other_model)

Incrementally add the content of another model to this model (+=).

Copies of all the reactions in the other model are added to this model. The objective is the sum of the objective expressions for the two models.

__setstate__(self, state)

Make sure all cobra.Objects in the model point to the model.

add_boundary(self, metabolite, type='exchange', reaction_id=None, lb=None, ub=1000.0)

Add a boundary reaction for a given metabolite.

There are three different types of pre-defined boundary reactions: exchange, demand, and sink reactions.

An exchange reaction is a reversible, imbalanced reaction that adds to or removes an extracellular metabolite from the extracellular compartment.

A demand reaction is an irreversible reaction that consumes an intracellular metabolite.

A sink is similar to an exchange but specifically for intracellular metabolites.

If you set the reaction ``type`` to something else, you must specify the

desired identifier of the created reaction along with its upper and lower bound. The name will be given by the metabolite name and the given ``type``.

Parameters

metabolite : cobra.Metabolite

Any given metabolite. The compartment is not checked but you are encouraged to stick to the definition of exchanges and sinks.

type : str, {"exchange", "demand", "sink"}

Using one of the pre-defined reaction types is easiest. If you want to create your own kind of boundary reaction choose any other string, e.g., 'my-boundary'.

reaction_id : str, optional
The ID of the resulting reaction. Only used for custom reactions.

lb : float, optional
The lower bound of the resulting reaction. Only used for custom reactions.

ub : float, optional
The upper bound of the resulting reaction. For the pre-defined reactions this default value determines all bounds.

Returns

cobra.Reaction
The created boundary reaction.

Examples

```
>>> import cobra.test
>>> model = cobra.test.create_test_model("textbook")
>>> demand = model.add_boundary(model.metabolites.atp_c, type="demand")
>>> demand.id
'DM_atp_c'
>>> demand.name
'ATP demand'
>>> demand.bounds
(0, 1000.0)
>>> demand.build_reaction_string()
'atp_c --> '
```

add_cons_vars(self, what, **kwargs)

Add constraints and variables to the model's mathematical problem.

Useful for variables and constraints that can not be expressed with reactions and simple lower and upper bounds.

Additions are reversed upon exit if the model itself is used as context.

Parameters

what : list or tuple of optlang variables or constraints.
The variables or constraints to add to the model. Must be of class `optlang.interface.Variable` or `optlang.interface.Constraint`.

****kwargs** : keyword arguments
Passed to solver.add()

add_metabolites(self, metabolite_list)

Will add a list of metabolites to the model object and add new constraints accordingly.

The change is reverted upon exit when using the model as a context.

Parameters

metabolite_list : A list of `cobra.core.Metabolite` objects

add_reaction(self, reaction)

Will add a cobra.Reaction object to the model, if reaction.id is not in self.reactions.

Parameters

reaction : cobra.Reaction
The reaction to add

Deprecated (0.6). Use `~cobra.[Model](#).add\_reactions` instead

add_reactions(self, reaction_list)

Add reactions to the model.

Reactions with identifiers identical to a reaction already in the model are ignored.

The change is reverted upon exit when using the model as a context.

Parameters

reaction_list : list
A list of `cobra.Reaction` objects

get_metabolite_compartments(self)

Return all metabolites' compartments.

merge(self, right, prefix_existing=None, inplace=True, objective='left')

Merge two models to create a model with the reactions from both models.

Custom constraints and variables from right models are also copied to left model, however note that, constraints and variables are assumed to be the same if they have the same name.

right : cobra.[Model](#)

The model to add reactions from

prefix_existing : string

Prefix the reaction identifier in the right that already exist in the left model with this string.

inplace : bool

Add reactions from right directly to left model object.

Otherwise, create a new model leaving the left model untouched.

When done within the model as context, changes to the models are reverted upon exit.

objective : string

One of 'left', 'right' or 'sum' for setting the objective of the resulting model to that of the corresponding model or the sum of both.

remove_cons_vars(self, what)

Remove variables and constraints from the model's mathematical problem.

Remove variables and constraints that were added directly to the model's underlying mathematical problem. Removals are reversed upon exit if the model itself is used as context.

Parameters

what : list or tuple of optlang variables or constraints.

The variables or constraints to add to the model. Must be of class `optlang.interface.Variable` or `optlang.interface.Constraint`.

remove_metabolites(self, metabolite_list, destructive=False)

Remove a list of metabolites from the the object.

The change is reverted upon exit when using the model as a context.

Parameters

metabolite_list : list

A list with `cobra.Metabolite` objects as elements.

destructive : bool

If False then the metabolite is removed from all associated reactions. If True then all associated reactions are removed from the [Model](#).

remove_reactions(self, reactions, remove_orphans=False)

Remove reactions from the model.

The change is reverted upon exit when using the model as a context.

Parameters

reactions : list

A list with reactions (`cobra.Reaction`), or their id's, to remove

remove_orphans : bool

Remove orphaned genes and metabolites from the model as well

slim_optimize(self, error_value=nan, message=None)

Optimize model without creating a solution object.

Creating a full solution object implies fetching shadow prices and flux values for all reactions and metabolites from the solver object. This necessarily takes some time and in cases where only one or two values are of interest, it is recommended to instead use this function which does not create a solution object returning only the value of the objective. Note however that the [optimize\(\)](#) function uses efficient means to fetch values so if you need fluxes/shadow prices for more than say 4 reactions/metabolites, then the total speed increase of `slim_optimize` versus `optimize` is expected to be small or even negative depending on how you fetch the values after optimization.

Parameters

error_value : float, None

The value to return if optimization failed due to e.g. infeasibility. If None, raise `OptimizationError` if the optimization fails.

message : string

Error message to use if the model optimization did not succeed.

Returns

float

The objective value.

summary(self, solution=None, threshold=1e-06, fva=None, names=False, floatfmt='.3g')

Print a summary of the input and output fluxes of the model.

Parameters

solution: cobra.Solution, optional

A previously solved model solution to use for generating the summary. If none provided (default), the summary method will

```

        resolve the model. Note that the solution object must match the
        model, i.e., changes to the model such as changed bounds,
        added or removed reactions are not taken into account by this
        method.
    threshold : float, optional
        Threshold below which fluxes are not reported.
    fva : pandas.DataFrame, float or None, optional
        Whether or not to include flux variability analysis in the output
        .
        If given, fva should either be a previous FVA solution matching
        the model or a float between 0 and 1 representing the
        fraction of the optimum objective to be searched.
    names : bool, optional
        Emit reaction and metabolite names rather than identifiers (default
        is False).
    floatfmt : string, optional
        Format string for floats (default '.3g').

```

Data descriptors inherited from [cobra.core.model.Model](#):

boundary

Boundary reactions in the model.
Reactions that either have no substrate or product.

compartments

constraints

The constraints in the cobra model.

In a cobra model, most constraints are metabolites and their stoichiometries. However, for specific use cases, it may also be useful to have other types of constraints. This property defines all constraints currently associated with the model's problem.

Returns

optlang.container.Container

A container with all associated constraints.

demands

Demand reactions in model.
Irreversible reactions that accumulate or consume a metabolite in the inside of the model.

description

exchanges

Exchange reactions in model.
Reactions that exchange mass with the exterior. Uses annotations and heuristics to exclude non-exchanges such as sink reactions.

medium

objective

Get or set the solver objective

Before introduction of the optlang based problems, this function returned the objective reactions as a list. With optlang, the objective is not limited a simple linear summation of individual reaction fluxes, making that return value ambiguous. Henceforth, use `cobra.util.solver.linear\_reaction\_coefficients` to get a dictionary of reactions with their linear coefficients (empty if there are none)

The set value can be dictionary (reactions as keys, linear coefficients as values), string (reaction identifier), int (reaction

index), Reaction or problem.Objective or sympy expression directly interpreted as objectives.

When using a `HistoryManager` context, this attribute can be set temporarily, reversed when exiting the context.

objective_direction

Get or set the objective direction.

When using a `HistoryManager` context, this attribute can be set temporarily, reversed when exiting the context.

problem

The interface to the model's underlying mathematical problem.

Solutions to cobra models are obtained by formulating a mathematical problem and solving it. Cobrapy uses the optlang package to accomplish that and with this property you can get access to the problem interface directly.

Returns

optlang.interface

The problem interface that defines methods for interacting with the problem and associated solver directly.

sinks

Sink reactions in model.

Reversible reactions that accumulate or consume a metabolite in the inside of the model.

solver

Get or set the attached solver instance.

The associated the solver object, which manages the interaction with the associated solver, e.g. glpk.

This property is useful for accessing the optimization problem directly and to define additional non-metabolic constraints.

Examples

```
>>> import cobra.test
>>> model = cobra.test.create_test_model("textbook")
>>> new = model.problem.Constraint(model.objective.expression,
>>> lb=0.99)
>>> model.solver.add(new)
```

variables

The mathematical variables in the cobra model.

In a cobra model, most variables are reactions. However, for specific use cases, it may also be useful to have other types of variables. This property defines all variables currently associated with the model's problem.

Returns

optlang.container.Container

A container with all associated variables.

Methods inherited from [cobra.core.object.Object](#):

`__repr__(self)`

`__str__(self)`

Data descriptors inherited from [cobra.core.object.Object](#):

`__dict__`
dictionary for instance variables (if defined)

`__weakref__`
list of weak references to the object (if defined)

`id`

Reaction

Method resolution order:

[Reaction](#)
[cobra.core.object.Object](#)
[builtin .object](#)

Methods defined here:

`__add__(self, other)`
Add two reactions

The stoichiometry will be the combined stoichiometry of the two reactions, and the gene reaction rule will be both rules combined by an and. All other attributes (i.e. reaction bounds) will match those of the first reaction

`__copy__(self)`

`__deepcopy__(self, memo)`

`__iadd__(self, other)`

`__imul__(self, coefficient)`
Scale coefficients in a reaction by a given value

E.g. A -> B becomes 2A -> 2B.

If coefficient is less than zero, the reaction is reversed and the bounds are swapped.

`__init__(self, id=None, name="", subsystem="", lower_bound=0.0, upper_bound=1000.0, objective_coefficient=0.0)`

`__isub__(self, other)`

`__mul__(self, coefficient)`

`__radd__ = __add__(self, other)`

`__setstate__(self, state)`
Probably not necessary to set `_model` as the `cobra.Model` that contains self sets the `_model` attribute for all metabolites and genes in the reaction.

However, to increase performance speed we do want to let the metabolite and gene know that they are employed in this reaction

`__str__(self)`

`__sub__(self, other)`

`add_metabolites(self, metabolites_to_add, combine=True, reversibly=True)`
Add metabolites and stoichiometric coefficients to the reaction.
If the final coefficient for a metabolite is 0 then it is removed

from the reaction.

The change is reverted upon exit when using the model as a context.

Parameters

metabolites_to_add : dict

Dictionary with metabolite objects or metabolite identifiers as keys and coefficients as values. If keys are strings (name of a metabolite) the reaction must already be part of a model and a metabolite with the given name must exist in the model.

combine : bool

Describes behavior a metabolite already exists in the reaction. True causes the coefficients to be added.

False causes the coefficient to be replaced.

reversibly : bool

Whether to add the change to the context to make the change reversibly or not (primarily intended for internal use).

build_reaction_from_string(self, reaction_str, verbose=True, fwd_arrow=None, rev_arrow=None, reversible_arrow=None, term_split='+')

Builds reaction from reaction equation reaction_str using parser

Takes a string and using the specifications supplied in the optional

arguments infers a set of metabolites, metabolite compartments and stoichiometries for the reaction. It also infers the reversibility of the reaction from the reaction arrow.

Changes to the associated model are reverted upon exit when using the model as a context.

Parameters

reaction_str : string

a string containing a reaction formula (equation)

verbose: bool

setting verbosity of function

fwd_arrow : re.compile

for forward irreversible reaction arrows

rev_arrow : re.compile

for backward irreversible reaction arrows

reversible_arrow : re.compile

for reversible reaction arrows

term_split : string

dividing individual metabolite entries

build_reaction_string(self, use_metabolite_names=False)

Generate a human readable reaction string

check_mass_balance(self)

Compute mass and charge balance for the reaction

returns a dict of {element: amount} for unbalanced elements.

"charge" is treated as an element in this dict

This should be empty for balanced reactions.

copy(self)

Copy a reaction

The referenced metabolites and genes are also copied.

delete(self, remove_orphans=False)

Removes the reaction from a model.

This removes all associations between a reaction the associated model, metabolites and genes.

The change is reverted upon exit when using the model as a context.

Deprecated, use ``reaction.remove_from_model`` instead.

Parameters

remove_orphans : bool

Remove orphaned genes and metabolites from the model as well

get_coefficient(self, metabolite_id)

Return the stoichiometric coefficient of a metabolite.

Parameters

metabolite_id : str or cobra.Metabolite

get_coefficients(self, metabolite_ids)

Return the stoichiometric coefficients for a list of metabolites.

Parameters

metabolite_ids : iterable

Containing ``str`` or ``cobra.Metabolite``s.

get_compartments(self)

lists compartments the metabolites are in

knock_out(self)

Knockout reaction by setting its bounds to zero.

remove_from_model(self, remove_orphans=False)

Removes the reaction from a model.

This removes all associations between a reaction the associated model, metabolites and genes.

The change is reverted upon exit when using the model as a context.

Parameters

remove_orphans : bool

Remove orphaned genes and metabolites from the model as well

subtract_metabolites(self, metabolites, combine=True, reversibly=True)

Subtract metabolites from a reaction.

That means add the metabolites with $-1 \times \text{coefficient}$. If the final coefficient for a metabolite is 0 then the metabolite is removed from the reaction.

Notes

* A final coefficient < 0 implies a reactant.

* The change is reverted upon exit when using the model as a context.

Parameters

metabolites : dict
 Dictionary where the keys are of class Metabolite and the values are the coefficients. These metabolites will be added to the reaction.

combine : bool
 Describes behavior a metabolite already exists in the reaction. True causes the coefficients to be added. False causes the coefficient to be replaced.

reversibly : bool
 Whether to add the change to the context to make the change reversibly or not (primarily intended for internal use).

Data descriptors defined here:

boundary

Whether or not this reaction is an exchange reaction.

Returns `True` if the reaction has either no products or reactants.

bounds

Get or set the bounds directly from a tuple

Convenience method for setting upper and lower bounds in one line using a tuple of lower and upper bound. Invalid bounds will raise a `AssertionError`.

When using a ``HistoryManager`` context, this attribute can be set temporarily, reversed when exiting the context.

compartments

lists compartments the metabolites are in

flux

The flux value in the most recent solution.

Flux is the primal value of the corresponding variable in the model.

Warnings

- * Accessing reaction fluxes through a ``Solution`` object is the safe, preferred, and only guaranteed to be correct way. You can see how to do so easily in the examples.
- * Reaction flux is retrieved from the currently defined ``self._model.solver``. The solver status is checked but there are no guarantees that the current solver state is the one you are looking for.
- * If you modify the underlying model after an optimization, you will retrieve the old optimization values.

Raises

RuntimeError

If the underlying model was never optimized beforehand or the

reaction is not part of a model.
OptimizationError
If the solver status is anything other than 'optimal'.
AssertionError
If the flux value is not within the bounds.

Examples

```
-----  
>>> import cobra.test  
>>> model = cobra.test.create_test_model("textbook")  
>>> solution = model.optimize()  
>>> model.reactions.PFK.flux  
7.477381962160283  
>>> solution.fluxes.PFK  
7.4773819621602833
```

flux_expression

Forward flux expression

Returns

sympy expression
The expression representing the the forward flux (if associated with model), otherwise None. Representing the net flux if model.reversible_encoding == 'unsplit' or None if reaction is not associated with a model

forward_variable

An optlang variable representing the forward flux

Returns

optlang.interface.Variable
An optlang variable for the forward flux or None if reaction is not associated with a model.

functional

All required enzymes for reaction are functional.

Returns

bool
True if the gene-protein-reaction (GPR) rule is fulfilled for this reaction, or if reaction is not associated to a model, otherwise False.

gene_name_reaction_rule

Display gene_reaction_rule with names instead.

Do NOT use this string for computation. It is intended to give a representation of the rule using more familiar gene names instead of the often cryptic ids.

gene_reaction_rule

genes

lower_bound

Get or set the lower bound

Setting the lower bound (float) will also adjust the associated optlang variables associated with the reaction. Infeasible combinations, such as a lower bound higher than the current upper bound will update the other bound.

When using a `HistoryManager` context, this attribute can be set temporarily, reversed when the exiting the context.

metabolites

model

returns the model the reaction is a part of

objective_coefficient

Get the coefficient for this reaction in a linear objective (float)

Assuming that the objective of the associated model is summation of fluxes from a set of reactions, the coefficient for each reaction can be obtained individually using this property. A more general way is to use the `model.objective` property directly.

products

Return a list of products for the reaction

reactants

Return a list of reactants for the reaction.

reaction

Human readable reaction string

reduced_cost

The reduced cost in the most recent solution.

Reduced cost is the dual value of the corresponding variable in the model.

Warnings

- * Accessing reduced costs through a `Solution` object is the safer, preferred, and only guaranteed to be correct way. You can see how to do so easily in the examples.
- * Reduced cost is retrieved from the currently defined `self.\_model.solver`. The solver status is checked but there are no guarantees that the current solver state is the one you are looking for.
- * If you modify the underlying model after an optimization, you will retrieve the old optimization values.

Raises

RuntimeError

If the underlying model was never optimized beforehand or the reaction is not part of a model.

OptimizationError

If the solver status is anything other than 'optimal'.

Examples

```
>>> import cobra.test
>>> model = cobra.test.create_test_model("textbook")
>>> solution = model.optimize()
>>> model.reactions.PFK.reduced_cost
-8.673617379884035e-18
```

```
>>> solution.reduced_costs.PFK
-8.6736173798840355e-18
```

reverse_id

Generate the id of reverse_variable from the reaction's id.

reverse_variable

An optlang variable representing the reverse flux

Returns

optlang.interface.Variable

An optlang variable for the reverse flux or None if reaction is not associated with a model.

reversibility

Whether the reaction can proceed in both directions (reversible)

This is computed from the current upper and lower bounds.

upper_bound

Get or set the upper bound

Setting the upper bound (float) will also adjust the associated optlang variables associated with the reaction. Infeasible combinations, such as a upper bound lower than the current lower bound will update the other bound.

When using a `HistoryManager` context, this attribute can be set temporarily, reversed when the exiting the context.

x

The flux through the reaction in the most recent solution.

Flux values are computed from the primal values of the variables in the solution.

y

The reduced cost of the reaction in the most recent solution.

Reduced costs are computed from the dual values of the variables in the solution.

Methods inherited from [cobra.core.object.Object](#):

__getstate__(self)

To prevent excessive replication during deepcopy.

__repr__(self)

Data descriptors inherited from [cobra.core.object.Object](#):

__dict__

dictionary for instance variables (if defined)

__weakref__

list of weak references to the object (if defined)

id

Log_Concentration

Data and other attributes defined here:

prefix = 'LC_'

Methods inherited from [MetaboliteVariable](#):

__init__(self, metabolite, **kwargs)

Data descriptors inherited from [MetaboliteVariable](#):

id

model

Methods inherited from [GenericVariable](#):

__add__(self, other)

Adding either two variables together or a variable and a numeric results in a new variable
:param other:
:return: a new Generic Variable

__mul__(self, other)

Multiplying either two variables together or a variable and a numeric results in a new variable
:param other:
:return: a new Generic Variable

__radd__(self, other)

Take priority on symmetric arithmetic operation
:param other:
:return:

__repr__(self)

__rmul__(self, other)

Take priority on symmetric arithmetic operation
:param other:
:return:

__rsub__(self, other)

Take priority on symmetric arithmetic operation
:param other:
:return:

__rtruediv__(self, other)

Take priority on symmetric arithmetic operation
:param other:
:return:

__sub__(self, other)

Subtracting either two variables together or a variable and a numeric results in a new variable
:param other:
:return: a new Generic Variable

__truediv__(self, other)

Dividing either two variables together or a variable and a numeric results in a new variable
:param other:
:return: a new Generic Variable

get_interface(self)

Called upon completion of **__init__**, initializes the value of **self.var**, which is returned upon call, and stores the actual interfaced variable

e.

:return: instance of Variable from the problem

get_operand(self, other)

For operations, choose if the operand is a [GenericVariable](#), in which we return its optlang variable, or something else (presumably a numeric)

and we let optlang decide what to do

:param other:

:return:

make_name(self)

Needs to be overridden by the subclass, concatenates the id with a prefix

:return: None

make_result(self, new_variable)

Returns a SymPy expression

:param new_variable:

:return:

test_consistency(self, other)

Tests whether a candidate to an operation is of the right type and is from the same problem

:param other: an object

:return: None

Data descriptors inherited from [GenericVariable](#):

__attrname__

Name the attribute the instances will have

Example: GenericVariable -> generic_variable

:return:

name

variable

Variable

Methods defined here:

```
__getstate__(self)
__init__(self, name, lb=None, ub=None, type='continuous', problem=None, *args,
**kwargs)
__reduce__(self)
__repr__(self)
    Does exactly the same as __str__ for now.
__setstate__(self, state)
__str__(self)
    Print a string representation of variable.
```

Examples

```
-----
>>> Variable('x', lb=-10, ub=10)
'-10 <= x <= 10'
```

set_bounds(self, lb, ub)

Change the lower and upper bounds of a variable.

to_json(self)

Returns a json-compatible object from the Variable that can be saved using the json module.

Example

```
-----
>>> import json
>>> with open("path_to_file.json", "w") as outfile:
>>>     json.dump(var.to\_json(), outfile)
```

Class methods defined here:

clone(cls, variable, **kwargs) from [sympy.core.assumptions.ManagedProperties](#)

Make a copy of another variable. The variable being copied can be of the same type or belong to a different solver interface.

Example

```
-----
>>> var_copy = Variable.clone(old_var)
```

from_json(cls, json_obj) from [sympy.core.assumptions.ManagedProperties](#)

Constructs a Variable from the provided json-object.

Example

```
-----
>>> import json
>>> with open("path_to_file.json") as infile:
>>>     var = Variable.from\_json(json.load(infile))
```

Data descriptors defined here:

dual

The dual of variable (None if no solution exists).

lb

Lower bound of variable.

name

Name of variable.

primal

The primal of variable (None if no solution exists).

type Variable type ('either continuous, integer, or binary'.)
ub Upper bound of variable.

Data and other attributes defined here:

default_assumptions = {}

Static methods inherited from [optlang.symbolics.Symbol](#):

__new__(cls, name, **kwargs)

Data descriptors inherited from [optlang.symbolics.Symbol](#):

__dict__
dictionary for instance variables (if defined)

__weakref__
list of weak references to the object (if defined)

Methods inherited from [sympy.core.symbol.Symbol](#):

__call__(self, *args)

__getnewargs__(self)

__new_stage2__(cls, name, **assumptions)

as_dummy(self)

Return a Dummy having the same name and same assumptions as self.

as_real_imag(self, deep=True, **hints)

is_constant(self, *wrt, **flags)

sort_key(...)

Static methods inherited from [sympy.core.symbol.Symbol](#):

__xnew__ = **__new_stage2__**(cls, name, **assumptions)

Data descriptors inherited from [sympy.core.symbol.Symbol](#):

assumptions0

free_symbols

Data and other attributes inherited from [sympy.core.symbol.Symbol](#):

is_Symbol = True

is_comparable = False

is_symbol = True

Data and other attributes inherited from [sympy.core.expr.AtomicExpr](#):

is_Atom = True

is_number = False

Methods inherited from [sympy.core.basic.Atom](#):

doit(self, **hints)

matches(self, expr, repl_dict={}, old=False)

xreplace(self, rule, hack2=False)

Class methods inherited from [sympy.core.basic.Atom](#):

class_key(cls) from [sympy.core.assumptions.ManagedProperties](#)

Methods inherited from [sympy.core.expr.Expr](#):

```
__abs__(self)
__add__(a, b)
__complex__(self)
__div__(a, b)
__float__(self)
__floordiv__(a, b)
__ge__(self, other)
__gt__(self, other)
__int__(self)
__le__(self, other)
__long__ = __int__(self)
__lt__(self, other)
__mod__(a, b)
__mul__(a, b)
__neg__(self)
__pos__(self)
__pow__(a, b)
__radd__(a, b)
__rdiv__(a, b)
__rfloordiv__(a, b)
__rmod__(a, b)
__rmul__(a, b)
__rpow__(a, b)
__rsub__(a, b)
__rtruediv__ = __rdiv__(a, b)
__sub__(a, b)
__truediv__ = __div__(a, b)
```

adjoint(self)

apart(self, x=None, **args)

See the apart function in sympy.polys

args_cnc(self, cset=False, warn=True, split_1=True)

Return [commutative factors, non-commutative factors] of self.

self is treated as a Mul and the ordering of the factors is maintained.

If ``cset`` is True the commutative factors will be returned in a set

.

If there were repeated factors (as may happen with an unevaluated Mul)

then an error will be raised unless it is explicitly suppressed by setting ``warn`` to False.

Note: -1 is always separated from a Number unless split_1 is False.

```
>>> from sympy import symbols, oo
>>> A, B = symbols('A B', commutative=0)
>>> x, y = symbols('x y')
>>> (-2*x*y).args_cnc()
[[-1, 2, x, y], []]
>>> (-2.5*x).args_cnc()
[[-1, 2.5, x], []]
>>> (-2*x*A*B*y).args_cnc()
[[-1, 2, x, y], [A, B]]
```

```
>>> (-2*x*A*B*y).args_cnc(split_1=False)
[[-2, x, y], [A, B]]
>>> (-2*x*y).args_cnc(cset=True)
[[-1, 2, x, y], []]
```

The arg is always treated as a Mul:

```
>>> (-2 + x + A).args_cnc()
[[], [x - 2 + A]]
>>> (-oo).args_cnc() # -oo is a singleton
[[-1, oo], []]
```

as_base_exp(self)

as_coeff_Add(self, rational=False)

Efficiently extract the coefficient of a summation.

as_coeff_Mul(self, rational=False)

Efficiently extract the coefficient of a product.

as_coeff_add(self, *deps)

Return the tuple (c, args) where self is written as an Add, ``a``.

c should be a Rational added to any terms of the Add that are independent of deps.

args should be a tuple of all other terms of ``a``; args is empty if self is a Number or if self is independent of deps (when given).

This should be used when you don't know if self is an Add or not but you want to treat self as an Add or if you want to process the individual arguments of the tail of self as an Add.

- if you know self is an Add and want only the head, use self.args[0];
- if you don't want to process the arguments of the tail but need the tail then use self.as_two_terms() which gives the head and tail.
- if you want to split self into an independent and dependent parts use ``self.as\_independent(\*deps)``

```
>>> from sympy import S
>>> from sympy.abc import x, y
>>> (S(3)).as_coeff_add()
(3, ())
>>> (3 + x).as_coeff_add()
(3, (x,))
>>> (3 + x + y).as_coeff_add(x)
(y + 3, (x,))
>>> (3 + y).as_coeff_add(x)
(y + 3, ())
```

as_coeff_exponent(self, x)

``c\*x\*\*e -> c,e`` where x can be any symbolic expression.

as_coeff_mul(self, *deps, **kwargs)

Return the tuple (c, args) where self is written as a Mul, ``m``.

c should be a Rational multiplied by any factors of the Mul that are independent of deps.

args should be a tuple of all other factors of m; args is empty if self is a Number or if self is independent of deps (when given).

This should be used when you don't know if self is a Mul or not but you want to treat self as a Mul or if you want to process the individual arguments of the tail of self as a Mul.

- if you know self is a Mul and want only the head, use `self.args[0]`;
- if you don't want to process the arguments of the tail but need the tail then use `self.as_two_terms()` which gives the head and tail;
- if you want to split self into an independent and dependent parts use `self.as_independent(*deps)`

```
>>> from sympy import S
>>> from sympy.abc import x, y
>>> (S(3)).as_coeff_mul()
(3, ())
>>> (3*x*y).as_coeff_mul()
(3, (x, y))
>>> (3*x*y).as_coeff_mul(x)
(3*y, (x,))
>>> (3*y).as_coeff_mul(x)
(3*y, ())
```

as_coefficient(self, expr)

Extracts symbolic coefficient at the given expression. In other words, this function separates 'self' into the product of 'expr' and 'expr'-free coefficient. If such separation is not possible it will return None.

Examples
=====

```
>>> from sympy import E, pi, sin, I, Poly
>>> from sympy.abc import x

>>> E.as_coefficient(E)
1
>>> (2*E).as_coefficient(E)
2
>>> (2*sin(E)*E).as_coefficient(E)
```

Two terms have E in them so a sum is returned. (If one were desiring the coefficient of the term exactly matching E then the constant from the returned expression could be selected. Or, for greater precision, a method of Poly can be used to indicate the desired term from which the coefficient is desired.)

```
>>> (2*E + x*E).as_coefficient(E)
x + 2
>>> _.args[0] # just want the exact match
2
>>> p = Poly(2*E + x*E); p
Poly(x*E + 2*E, x, E, domain='ZZ')
>>> p.coeff_monomial(E)
2
>>> p.nth(0, 1)
2
```

Since the following cannot be written as a product containing E as a factor, None is returned. (If the coefficient `2*x` is desired then the `coeff` method should be used.)

```
>>> (2*E*x + x).as_coefficient(E)
None
>>> (2*E*x + x).coeff(E)
2*x
```

```
>>> (E*(x + 1) + x).as\_coefficient(E)

>>> (2*pi*I).as\_coefficient(pi*I)
2
>>> (2*I).as\_coefficient(pi*I)
```

See Also
=====

coeff: return sum of terms have a given factor
as_coeff_Add: separate the additive constant from an expression
as_coeff_Mul: separate the multiplicative constant from an expression
as_independent: separate x-dependent terms/factors from others
sympy.polys.polytools.coeff_monomial: efficiently find the single coefficient of a monomial in Poly
sympy.polys.polytools.nth: like coeff_monomial but powers of monomial terms are used

as_coefficients_dict(self)

Return a dictionary mapping terms to their Rational coefficient. Since the dictionary is a defaultdict, inquiries about terms which were not present will return a coefficient of 0. If an expression is not an Add it is considered to have a single term.

Examples
=====

```
>>> from sympy.abc import a, x
>>> (3*x + a*x + 4).as\_coefficients\_dict()
{1: 4, x: 3, a*x: 1}
>>> _[a]
0
>>> (3*a*x).as\_coefficients\_dict()
{a*x: 3}
```

as_content_primitive(self, radical=False, clear=True)

This method should recursively remove a Rational from all arguments and return that (content) and the new self (primitive). The content should always be positive and ``Mul(\*foo.[as\\_content\\_primitive](#)()) == foo``.

The primitive need no be in canonical form and should try to preserve the underlying structure if possible (i.e. expand_mul should not be applied to self).

Examples
=====

```
>>> from sympy import sqrt
>>> from sympy.abc import x, y, z

>>> eq = 2 + 2*x + 2*y*(3 + 3*y)
```

The as_content_primitive function is recursive and retains structure:

```
>>> eq.as\_content\_primitive()
(2, x + 3*y*(y + 1) + 1)
```

Integer powers will have Rationals extracted from the base:

```
>>> ((2 + 6*x)**2).as\_content\_primitive()
(4, (3*x + 1)**2)
>>> ((2 + 6*x)**(2*y)).as\_content\_primitive()
(1, (2*(3*x + 1))**(2*y))
```


Terms may end up joining once their `as_content_primitives` are added:

```
>>> ((5*(x*(1 + y)) + 2*x*(3 + 3*y))).as\_content\_primitive()
(11, x*(y + 1))
>>> ((3*(x*(1 + y)) + 2*x*(3 + 3*y))).as\_content\_primitive()
(9, x*(y + 1))
>>> ((3*(z*(1 + y)) + 2.0*x*(3 + 3*y))).as\_content\_primitive()
(1, 6.0*x*(y + 1) + 3*z*(y + 1))
>>> ((5*(x*(1 + y)) + 2*x*(3 + 3*y))**2).as\_content\_primitive()
(121, x**2*(y + 1)**2)
>>> ((5*(x*(1 + y)) + 2.0*x*(3 + 3*y))**2).as\_content\_primitive()
(1, 121.0*x**2*(y + 1)**2)
```

Radical content can also be factored out of the primitive:

```
>>> (2*sqrt(2) + 4*sqrt(10)).as\_content\_primitive(radical=True)
(2, sqrt(2)*(1 + 2*sqrt(5)))
```

If `clear=False` (default is `True`) then content will not be removed from an `Add` if it can be distributed to leave one or more terms with integer coefficients.

```
>>> (x/2 + y).as\_content\_primitive()
(1/2, x + 2*y)
>>> (x/2 + y).as\_content\_primitive(clear=False)
(1, x/2 + y)
```

`as_expr(self, *gens)`

Convert a polynomial to a SymPy expression.

Examples
=====

```
>>> from sympy import sin
>>> from sympy.abc import x, y

>>> f = (x**2 + x*y).as\_poly(x, y)
>>> f.as\_expr()
x**2 + x*y

>>> sin(x).as\_expr()
sin(x)
```

`as_independent(self, *deps, **hint)`

A mostly naive separation of a `Mul` or `Add` into arguments that are not are dependent on `deps`. To obtain as complete a separation of variable `s` as possible, use a separation method first, e.g.:

- * `separatevars()` to change `Mul`, `Add` and `Pow` (including `exp`) into `Mul`
- * `.expand(mul=True)` to change `Add` or `Mul` into `Add`
- * `.expand(log=True)` to change `log expr` into an `Add`

The only non-naive thing that is done here is to respect noncommutative ordering of variables and to always return `(0, 0)` for ``self`` of zero regardless of hints.

For nonzero ``self``, the returned tuple `(i, d)` has the following interpretation:

- * `i` will has no variable that appears in `deps`

```

* d will be 1 or else have terms that contain variables that are in d
eps
* if self is an Add then self = i + d
* if self is a Mul then self = i*d
* otherwise (self, S.One) or (S.One, self) is returned.

```

To force the expression to be treated as an Add, use the hint `as_Add=True`

Examples

=====

```
-- self is an Add
```

```

>>> from sympy import sin, cos, exp
>>> from sympy.abc import x, y, z

>>> (x + x*y).as_independent(x)
(0, x*y + x)
>>> (x + x*y).as_independent(y)
(x, x*y)
>>> (2*x*sin(x) + y + x + z).as_independent(x)
(y + z, 2*x*sin(x) + x)
>>> (2*x*sin(x) + y + x + z).as_independent(x, y)
(z, 2*x*sin(x) + x + y)

```

```
-- self is a Mul
```

```

>>> (x*sin(x)*cos(y)).as_independent(x)
(cos(y), x*sin(x))

```

non-

commutative terms cannot always be separated out when self is a Mul

```

>>> from sympy import symbols
>>> n1, n2, n3 = symbols('n1 n2 n3', commutative=False)
>>> (n1 + n1*n2).as_independent(n2)
(n1, n1*n2)
>>> (n2*n1 + n1*n2).as_independent(n2)
(0, n1*n2 + n2*n1)
>>> (n1*n2*n3).as_independent(n1)
(1, n1*n2*n3)
>>> (n1*n2*n3).as_independent(n2)
(n1, n2*n3)
>>> ((x-n1)*(x-y)).as_independent(x)
(1, (x - y)*(x - n1))

```

```
-- self is anything else:
```

```

>>> (sin(x)).as_independent(x)
(1, sin(x))
>>> (sin(x)).as_independent(y)
(sin(x), 1)
>>> exp(x+y).as_independent(x)
(1, exp(x + y))

```

```
-- force self to be treated as an Add:
```

```

>>> (3*x).as_independent(x, as_Add=True)
(0, 3*x)

```

```
-- force self to be treated as a Mul:
```

```
>>> (3+x).as\_independent(x, as_Add=False)
(1, x + 3)
>>> (-3+x).as\_independent(x, as_Add=False)
(1, x - 3)
```

Note how the below differs from the above in making the constant on the dep term positive.

```
>>> (y*(-3+x)).as\_independent(x)
(y, x - 3)
```

-- use [.as_independent\(\)](#) for true independence testing instead of [.has\(\)](#). The former considers only symbols in the free symbols while the latter considers all symbols

```
>>> from sympy import Integral
>>> I = Integral(x, (x, 1, 2))
>>> I.has(x)
True
>>> x in I.free_symbols
False
>>> I.as\_independent(x) == (I, 1)
True
>>> (I + x).as\_independent(x) == (I, x)
True
```

Note: when trying to get independent terms, a separation method might need to be used first. In this case, it is important to keep track of what you send to this routine so you know how to interpret the returned values

```
>>> from sympy import separatevars, log
>>> separatevars(exp(x+y)).as\_independent(x)
(exp(y), exp(x))
>>> (x + x*y).as\_independent(y)
(x, x*y)
>>> separatevars(x + x*y).as\_independent(y)
(x, y + 1)
>>> (x*(1 + y)).as\_independent(y)
(x, y + 1)
>>> (x*(1 + y)).expand(mul=True).as\_independent(y)
(x, x*y)
>>> a, b=symbols('a b', positive=True)
>>> (log(a*b)).expand(log=True).as\_independent(b)
(log(a), log(b))
```

See Also

=====

```
.separatevars(), .expand(log=True), Add.as_two_terms(),
Mul.as_two_terms(), .as\_coeff\_add(), .as\_coeff\_mul()
```

as_leading_term(...)

Returns the leading (nonzero) term of the series expansion of self.

The `_eval_as_leading_term` routines are used to do this, and they must always return a non-zero value.

Examples

=====

```
>>> from sympy.abc import x
>>> (1 + x + x**2).as\_leading\_term(x)
1
>>> (1/x**2 + x + x**2).as\_leading\_term(x)
x**(-2)
```

as_numer_denom(self)

expression -> a/b -> a, b

This is just a stub that should be defined by an object's class methods to get anything else.

See Also

=====

normal: return a/b instead of a, b

as_ordered_factors(self, order=None)

Return list of ordered factors (if Mul) else [self].

as_ordered_terms(self, order=None, data=False)

Transform an expression to an ordered list of terms.

Examples

=====

```
>>> from sympy import sin, cos
>>> from sympy.abc import x

>>> (sin(x)**2*cos(x) + sin(x)**2 + 1).as\_ordered\_terms()
[sin(x)**2*cos(x), sin(x)**2, 1]
```

as_powers_dict(self)

Return self as a dictionary of factors with each factor being treated as a power. The keys are the bases of the factors and the values, the corresponding exponents. The resulting dictionary should be used with caution if the expression is a Mul and contains non-commutative factors since the order that they appeared will be lost in the dictionary.

as_terms(self)

Transform an expression to a list of terms.

cancel(self, *gens, **args)

See the cancel function in sympy.polys

coeff(self, x, n=1, right=False)

Returns the coefficient from the term(s) containing ``x\*\*n``. If ``n`` is zero then all terms independent of ``x`` will be returned.

When ``x`` is noncommutative, the coefficient to the left (default) or right of ``x`` can be returned. The keyword 'right' is ignored when ``x`` is commutative.

See Also

=====

as_coefficient: separate the expression into a coefficient and factor
as_coeff_Add: separate the additive constant from an expression
as_coeff_Mul: separate the multiplicative constant from an expression
as_independent: separate x-dependent terms/factors from others
sympy.polys.polytools.coeff_monomial: efficiently find the single coefficient of a monomial in Poly
sympy.polys.polytools.nth: like coeff_monomial but powers of monomial terms are used

Examples

=====

```
>>> from sympy import symbols
>>> from sympy.abc import x, y, z
```

You can select terms that have an explicit negative in front of them:

```
>>> (-x + 2*y).coeff(-1)
x
>>> (x - 2*y).coeff(-1)
2*y
```

You can select terms with no Rational coefficient:

```
>>> (x + 2*y).coeff(1)
x
>>> (3 + 2*x + 4*x**2).coeff(1)
0
```

You can select terms independent of x by making $n=0$; in this case `expr.as_independent(x)[0]` is returned (and 0 will be returned instead of None):

```
>>> (3 + 2*x + 4*x**2).coeff(x, 0)
3
>>> eq = ((x + 1)**3).expand() + 1
>>> eq
x**3 + 3*x**2 + 3*x + 2
>>> [eq.coeff(x, i) for i in reversed(range(4))]
[1, 3, 3, 2]
>>> eq -= 2
>>> [eq.coeff(x, i) for i in reversed(range(4))]
[1, 3, 3, 0]
```

You can select terms that have a numerical term in front of them:

```
>>> (-x - 2*y).coeff(2)
-y
>>> from sympy import sqrt
>>> (x + sqrt(2)*x).coeff(sqrt(2))
x
```

The matching is exact:

```
>>> (3 + 2*x + 4*x**2).coeff(x)
2
>>> (3 + 2*x + 4*x**2).coeff(x**2)
4
>>> (3 + 2*x + 4*x**2).coeff(x**3)
0
>>> (z*(x + y)**2).coeff((x + y)**2)
z
>>> (z*(x + y)**2).coeff(x + y)
0
```

In addition, no factoring is done, so $1 + z*(1 + y)$ is not obtained from the following:

```
>>> (x + z*(x + x*y)).coeff(x)
```

1

If such factoring is desired, `factor_terms` can be used first:

```
>>> from sympy import factor_terms
>>> factor_terms(x + z*(x + x*y)).coeff(x)
z*(y + 1) + 1

>>> n, m, o = symbols('n m o', commutative=False)
>>> n.coeff(n)
1
>>> (3*n).coeff(n)
3
>>> (n*m + m*n*m).coeff(n) # = (1 + m)*n*m
1 + m
>>> (n*m + m*n*m).coeff(n, right=True) # = (1 + m)*n*m
m
```

If there is more than one possible coefficient 0 is returned:

```
>>> (n*m + m*n).coeff(n)
0
```

If there is only one possible coefficient, it is returned:

```
>>> (n*m + x*m*n).coeff(m*n)
x
>>> (n*m + x*m*n).coeff(m*n, right=1)
1
```

collect(self, syms, func=None, evaluate=True, exact=False, distribute_order_term=True)

See the `collect` function in `sympy.simplify`

combsimp(self)

See the `combsimp` function in `sympy.simplify`

compute_leading_term(self, x, logx=None)

`as_leading_term` is only allowed for results of `.series()`

This is a wrapper to compute a series first.

conjugate(self)

could_extract_minus_sign(self)

Canonical way to choose an element in the set $\{e, -e\}$ where e is any expression. If the canonical element is e , we have `e.could\_extract\_minus\_sign() == True`, else `e.could\_extract\_minus\_sign() == False`.

For any expression, the set `{e.could\_extract\_minus\_sign(), (-e).could\_extract\_minus\_sign()}` must be `{True, False}`.

```
>>> from sympy.abc import x, y
>>> (x-y).could\_extract\_minus\_sign() != (y-x).could\_extract\_minus\_sign()
True
```

count_ops(self, visual=None)

wrapper for `count_ops` that returns the operation count.

diff(self, *symbols, **assumptions)

equals(self, other, failing_expression=False)

Return True if `self == other`, False if it doesn't, or None. If `failing_expression` is True then the expression which did not simplify to a 0 will be returned instead of None.

If `self` is a Number (or complex number) that is not zero, then

the result is False.

If ``self`` is a number and has not evaluated to zero, evalf will be used to test whether the expression evaluates to zero. If it does so and the result has significance (i.e. the precision is either -1, for a Rational result, or is greater than 1) then the evalf value will be used to return True or False.

expand(...)

Expand an expression using hints.

See the docstring of the [expand\(\)](#) function in `sympy.core.function` for more information.

extract_additively(self, c)

Return `self - c` if it's possible to subtract `c` from `self` and make all matching coefficients move towards zero, else return `None`.

Examples

=====

```
>>> from sympy.abc import x, y
>>> e = 2*x + 3
>>> e.extract\_additively(x + 1)
x + 2
>>> e.extract\_additively(3*x)
>>> e.extract\_additively(4)
>>> (y*(x + 1)).extract\_additively(x + 1)
>>> ((x + 1)*(x + 2*y + 1) + 3).extract\_additively(x + 1)
(x + 1)*(x + 2*y) + 3
```

Sometimes auto-expansion will return a less simplified result than desired; `gcd_terms` might be used in such cases:

```
>>> from sympy import gcd_terms
>>> (4*x*(y + 1) + y).extract\_additively(x)
4*x*(y + 1) + x*(4*y + 3) - x*(4*y + 4) + y
>>> gcd_terms(_)
x*(4*y + 3) + y
```

See Also

=====

`extract_multiplicatively`
`coeff`
`as_coefficient`

extract_branch_factor(self, allow_half=False)

Try to write `self` as ```exp_polar(2*pi*I*n)*z``` in a nice way. Return `(z, n)`.

```
>>> from sympy import exp_polar, I, pi
>>> from sympy.abc import x, y
>>> exp_polar(I*pi).extract\_branch\_factor()
(exp_polar(I*pi), 0)
>>> exp_polar(2*I*pi).extract\_branch\_factor()
(1, 1)
>>> exp_polar(-pi*I).extract\_branch\_factor()
(exp_polar(I*pi), -1)
>>> exp_polar(3*pi*I + x).extract\_branch\_factor()
(exp_polar(x + I*pi), 1)
>>> (y*exp_polar(-
5*pi*I)*exp_polar(3*pi*I + 2*pi*x)).extract\_branch\_factor()
(y*exp_polar(2*pi*x), -1)
```

```
>>> exp_polar(-I*pi/2).extract\_branch\_factor()
(exp_polar(-I*pi/2), 0)
```

If `allow_half` is True, also extract `exp_polar(I*pi)`:

```
>>> exp_polar(I*pi).extract\_branch\_factor(allow_half=True)
(1, 1/2)
>>> exp_polar(2*I*pi).extract\_branch\_factor(allow_half=True)
(1, 1)
>>> exp_polar(3*I*pi).extract\_branch\_factor(allow_half=True)
(1, 3/2)
>>> exp_polar(-I*pi).extract\_branch\_factor(allow_half=True)
(1, -1/2)
```

`extract_multiplicatively(self, c)`

Return None if it's not possible to make self in the form `c * something` in a nice way, i.e. preserving the properties of arguments of self.

```
>>> from sympy import symbols, Rational

>>> x, y = symbols('x,y', real=True)

>>> ((x*y)**3).extract\_multiplicatively(x**2 * y)
x*y**2

>>> ((x*y)**3).extract\_multiplicatively(x**4 * y)

>>> (2*x).extract\_multiplicatively(2)
x

>>> (2*x).extract\_multiplicatively(3)

>>> (Rational(1, 2)*x).extract\_multiplicatively(3)
x/6
```

`factor(self, *gens, **args)`

See the [factor\(\)](#) function in `sympy.polys.polytools`

`fourier_series(self, limits=None)`

Compute fourier sine/cosine series of self.

See the docstring of the `:func:`fourier_series`` in `sympy.series.fourier` for more information.

`fps(self, x=None, x0=0, dir=1, hyper=True, order=4, rational=True, full=False)`

Compute formal power series of self.

See the docstring of the `:func:`fps`` function in `sympy.series.formal` for more information.

`getO(self)`

Returns the additive `O(..)` symbol if there is one, else None.

`getn(self)`

Returns the order of the expression.

The order is determined either from the `O(...)` term. If there is no `O(...)` term, it returns None.

Examples
=====

```
>>> from sympy import O
```



```
>>> from sympy.abc import x
>>> (1 + x + O(x**2)).getn()
2
>>> (1 + x).getn()
```

integrate(self, *args, **kwargs)

See the integrate function in sympy.integrals

invert(self, g, *gens, **args)

Return the multiplicative inverse of ``self`` mod ``g``
where ``self`` (and ``g``) may be symbolic expressions).

See Also

=====

[sympy.core.numbers.mod_inverse](#), [sympy.polys.polytools.invert](#)

is_algebraic_expr(self, *syms)

This tests whether a given expression is algebraic or not, in the given symbols, syms. When syms is not given, all free symbols will be used. The rational function does not have to be in expanded or in any kind of canonical form.

This function returns False for expressions that are "algebraic expressions" with symbolic exponents. This is a simple extension to the

[is_rational_function](#), including rational exponentiation.

Examples

=====

```
>>> from sympy import Symbol, sqrt
>>> x = Symbol('x', real=True)
>>> sqrt(1 + x).is\_rational\_function()
False
>>> sqrt(1 + x).is\_algebraic\_expr()
True
```

This function does not attempt any nontrivial simplifications that may result in an expression that does not appear to be an algebraic expression to become one.

```
>>> from sympy import exp, factor
>>> a = sqrt(exp(x)**2 + 2*exp(x) + 1)/(exp(x) + 1)
>>> a.is\_algebraic\_expr(x)
False
>>> factor(a).is\_algebraic\_expr()
True
```

See Also

=====

[is_rational_function](#)()

References

=====

- http://en.wikipedia.org/wiki/Algebraic_expression

is_polynomial(self, *syms)

Return True if self is a polynomial in syms and False otherwise.

This checks if self is an exact polynomial in syms. This function returns False for expressions that are "polynomials" with symbolic exponents. Thus, you should be able to apply polynomial algorithms to

o
 expressions for which this returns True, and `Poly(expr, \*syms)` should
 work if and only if `expr.is_polynomial(\*syms)` returns True. The
 polynomial does not have to be in expanded form. If no symbols are
 given, all free symbols in the expression will be used.

This is not part of the assumptions system. You cannot do
`Symbol('z', polynomial=True)`.

Examples
 =====

```
>>> from sympy import Symbol
>>> x = Symbol('x')
>>> ((x**2 + 1)**4).is_polynomial(x)
True
>>> ((x**2 + 1)**4).is_polynomial()
True
>>> (2**x + 1).is_polynomial(x)
False

>>> n = Symbol('n', nonnegative=True, integer=True)
>>> (x**n + 1).is_polynomial(x)
False
```

This function does not attempt any nontrivial simplifications that may
 result in an expression that does not appear to be a polynomial to
 become one.

```
>>> from sympy import sqrt, factor, cancel
>>> y = Symbol('y', positive=True)
>>> a = sqrt(y**2 + 2*y + 1)
>>> a.is_polynomial(y)
False
>>> factor(a)
y + 1
>>> factor(a).is_polynomial(y)
True

>>> b = (y**2 + 2*y + 1)/(y + 1)
>>> b.is_polynomial(y)
False
>>> cancel(b)
y + 1
>>> cancel(b).is_polynomial(y)
True
```

See also [`.is\_rational\_function\(\)`](#)

is_rational_function(self, *syms)

Test whether function is a ratio of two polynomials in the given
 symbols, syms. When syms is not given, all free symbols will be used.
 The rational function does not have to be in expanded or in any kind
 of
 canonical form.

This function returns False for expressions that are "rational
 functions" with symbolic exponents. Thus, you should be able to call
[`.as\_numer\_denom\(\)`](#) and apply polynomial algorithms to the result for

expressions for which this returns True.

This is not part of the assumptions system. You cannot do `Symbol('z', rational_function=True)`.

Examples
=====

```
>>> from sympy import Symbol, sin
>>> from sympy.abc import x, y

>>> (x/y).is\_rational\_function()
True

>>> (x**2).is\_rational\_function()
True

>>> (x/sin(y)).is\_rational\_function(y)
False

>>> n = Symbol('n', integer=True)
>>> (x**n + 1).is\_rational\_function(x)
False
```

This function does not attempt any nontrivial simplifications that may result in an expression that does not appear to be a rational function to become one.

```
>>> from sympy import sqrt, factor
>>> y = Symbol('y', positive=True)
>>> a = sqrt(y**2 + 2*y + 1)/y
>>> a.is\_rational\_function(y)
False
>>> factor(a)
(y + 1)/y
>>> factor(a).is\_rational\_function(y)
True
```

See also [is_algebraic_expr\(\)](#).

leadterm(self, x)

Returns the leading term $a*x^b$ as a tuple (a, b) .

Examples
=====

```
>>> from sympy.abc import x
>>> (1+x+x**2).leadterm(x)
(1, 0)
>>> (1/x**2+x+x**2).leadterm(x)
(1, -2)
```

limit(self, x, xlim, dir='+')

Compute limit $x \rightarrow \text{xlim}$.

lseries(self, x=None, x0=0, dir='+', logx=None)

Wrapper for series yielding an iterator of the terms of the series.

Note: an infinite series will yield an infinite iterator. The following, for example, will never terminate. It will just keep printing terms

of the `sin(x)` series::

```
for term in sin(x).lseries(x):
    print term
```

The advantage of [lseries](#)() over [nseries](#)() is that many times you are just interested in the next term in the series (i.e. the first term for example), but you don't know how many you should ask for in [nseries](#)() using the "n" parameter.

See also [nseries](#)() .

normal(self)

nseries(self, x=None, x0=0, n=6, dir='+', logx=None)

Wrapper to `_eval_nseries` if assumptions allow, else to `series`.

If `x` is given, `x0` is 0, `dir`='+', and `self` has `x`, then `_eval_nseries` is called. This calculates "n" terms in the innermost expressions and then builds up the final series just by "cross-multiplying" everything out.

The optional ```logx``` parameter can be used to replace any `log(x)` in the returned series with a symbolic value to avoid evaluating `log(x)` at 0. A symbol to use in place of `log(x)` should be provided.

Advantage -- it's fast, because we don't have to determine how many terms we need to calculate in advance.

Disadvantage -- you may end up with less terms than you may have expected, but the $O(x^{**n})$ term appended will always be correct and so the result, though perhaps shorter, will also be correct.

If any of those assumptions is not met, this is treated like a wrapper to `series` which will try harder to return the correct number of terms.

See also [lseries](#)() .

Examples

=====

```
>>> from sympy import sin, log, Symbol
>>> from sympy.abc import x, y
>>> sin(x).nseries(x, 0, 6)
x - x**3/6 + x**5/120 + O(x**6)
>>> log(x+1).nseries(x, 0, 5)
x - x**2/2 + x**3/3 - x**4/4 + O(x**5)
```

Handling of the ```logx``` parameter --- in the following example the expansion fails since ```sin``` does not have an asymptotic expansion at $-\infty$ (the limit of `log(x)` as `x` approaches 0):

```
>>> e = sin(log(x))
>>> e.nseries(x, 0, 6)
Traceback (most recent call last):
...
PoleError: ...
```

```
...
>>> logx = Symbol('logx')
>>> e.nseries(x, 0, 6, logx=logx)
sin(logx)
```

In the following example, the expansion works but gives only an Order term unless the ``logx`` parameter is used:

```
>>> e = x**y
>>> e.nseries(x, 0, 2)
O(log(x)**2)
>>> e.nseries(x, 0, 2, logx=logx)
exp(logx*y)
```

nsimplify(self, constants=[], tolerance=None, full=False)

See the nsimplify function in sympy.simplify

powsimp(self, *args, **kwargs)

See the powsimp function in sympy.simplify

primitive(self)

Return the positive Rational that can be extracted non-recursively from every term of self (i.e., self is treated like an Add). This is like the [as_coeff_Mul\(\)](#) method but primitive always extracts a positive Rational (never a negative or a Float).

Examples
=====

```
>>> from sympy.abc import x
>>> (3*(x + 1)**2).primitive()
(3, (x + 1)**2)
>>> a = (6*x + 2); a.primitive()
(2, 3*x + 1)
>>> b = (x/2 + 3); b.primitive()
(1/2, x + 6)
>>> (a*b).primitive() == (1, a*b)
True
```

ratsimp(self, **kwargs)

See the ratsimp function in sympy.simplify

ratsimp(self)

See the ratsimp function in sympy.simplify

refine(self, assumption=True)

See the refine function in sympy.assumptions

removeO(self)

Removes the additive O(..) symbol if there is one

round(self, p=0)

Return x rounded to the given decimal place.

If a complex number would results, apply round to the real and imaginary components of the number.

Examples
=====

```
>>> from sympy import pi, E, I, S, Add, Mul, Number
>>> S(10.5).round()
11.
>>> pi.round()
3.
```

```
>>> pi.round(2)
3.14
>>> (2*pi + E*I).round()
6. + 3.*I
```

The round method has a chopping effect:

```
>>> (2*pi + I/10).round()
6.
>>> (pi/10 + 2*I).round()
2.*I
>>> (pi/10 + E*I).round(2)
0.31 + 2.72*I
```

Notes
=====

Do not confuse the Python builtin function, round, with the SymPy method of the same name. The former always returns a float (or raises an error if applied to a complex value) while the latter returns either a Number or a complex number:

```
>>> isinstance(round(S(123), -2), Number)
False
>>> isinstance(S(123).round(-2), Number)
True
>>> isinstance((3*I).round(), Mul)
True
>>> isinstance((1 + 3*I).round(), Add)
True
```

separate(self, deep=False, force=False)

See the separate function in sympy.simplify

series(self, x=None, x0=0, n=6, dir='+', logx=None)

Series expansion of "self" around ``x = x0`` yielding either terms of the series one by one (the lazy series given when n=None), else all the terms at once when n != None.

Returns the series expansion of "self" around the point ``x = x0`` with respect to ``x`` up to ``O((x - x0)\*\*n, x, x0)`` (default n is 6).

If ``x=None`` and ``self`` is univariate, the univariate symbol will be supplied, otherwise an error will be raised.

```
>>> from sympy import cos, exp
>>> from sympy.abc import x, y
>>> cos(x).series()
1 - x**2/2 + x**4/24 + O(x**6)
>>> cos(x).series(n=4)
1 - x**2/2 + O(x**4)
>>> cos(x).series(x, x0=1, n=2)
cos(1) - (x - 1)*sin(1) + O((x - 1)**2, (x, 1))
>>> e = cos(x + exp(y))
>>> e.series(y, n=2)
cos(x + 1) - y*sin(x + 1) + O(y**2)
>>> e.series(x, n=2)
cos(exp(y)) - x*sin(exp(y)) + O(x**2)
```

If ``n=None`` then a generator of the series terms will be returned.

```
>>> term=cos(x).series(n=None)
```

```
>>> [next(term) for i in range(2)]  
[1, -x**2/2]
```

For ``dir=+`` (default) the series is calculated from the right and for ``dir=-`` the series from the left. For smooth functions this flag will not alter the results.

```
>>> abs(x).series(dir="+")  
x  
>>> abs(x).series(dir="-")  
-x
```

simplify(self, ratio=1.7, measure=None)

See the simplify function in `sympy.simplify`

taylor_term(self, n, x, *previous_terms)

General method for the taylor term.

This method is slow, because it differentiates n -times. Subclasses can redefine it to make it faster by using the "previous_terms".

together(self, *args, **kwargs)

See the together function in `sympy.polys`

transpose(self)

trigsimp(self, **args)

See the trigsimp function in `sympy.simplify`

Methods inherited from [sympy.logic.boolalg.Boolean](#):

__and__(self, other)

Overloading for & operator

__invert__(self)

Overloading for ~

__lshift__(self, other)

Overloading for <<

__or__(self, other)

Overloading for |

__rand__ = __and__(self, other)

Overloading for & operator

__rshift__ = __rshift__(self, other)

Overloading for >>

__ror__ = __or__(self, other)

Overloading for |

__rrshift__ = __lshift__(self, other)

Overloading for <<

__rshift__(self, other)

Overloading for >>

__rxor__ = __xor__(self, other)

__xor__(self, other)

Methods inherited from [sympy.core.basic.Basic](#):

__eq__(self, other)

Return a boolean indicating whether $a == b$ on the basis of their symbolic trees.

This is the same as `a.compare(b) == 0` but faster.

Notes

=====

If a class that overrides `__eq__()` needs to retain the implementation of `__hash__()` from a parent class, the interpreter must be told this explicitly by setting `__hash__ = <ParentClass>.__hash__`. Otherwise the inheritance of `__hash__()` will be blocked, just as if `__hash__` had been explicitly set to `None`.

References
=====

from http://docs.python.org/dev/reference/datamodel.html#object.__hash__

__hash__(self)

__ne__(self, other)

`a != b` -

> Compare two symbolic trees and see whether they are different

this is the same as:

`a.compare(b) != 0`

but faster

__reduce_ex__(self, proto)

Pickling support.

as_poly(self, *gens, **args)

Converts ``self`` to a polynomial or returns ``None``.

```
>>> from sympy import sin
```

```
>>> from sympy.abc import x, y
```

```
>>> print((x**2 + x*y).as_poly())
```

```
Poly(x**2 + x*y, x, y, domain='ZZ')
```

```
>>> print((x**2 + x*y).as_poly(x, y))
```

```
Poly(x**2 + x*y, x, y, domain='ZZ')
```

```
>>> print((x**2 + sin(y)).as_poly(x, y))
```

```
None
```

atoms(self, *types)

Returns the atoms that form the current object.

By default, only objects that are truly atomic and can't be divided into smaller pieces are returned: symbols, numbers, and number symbols like `I` and `pi`. It is possible to request atoms of any type, however, as demonstrated below.

Examples
=====

```
>>> from sympy import I, pi, sin
```

```
>>> from sympy.abc import x, y
```

```
>>> (1 + x + 2*sin(y + I*pi)).atoms()
```

```
{1, 2, I, pi, x, y}
```

If one or more types are given, the results will contain only those types of atoms.

Examples
=====


```

>>> from sympy import Number, NumberSymbol, Symbol
>>> (1 + x + 2*sin(y + I*pi)).atoms(Symbol)
{x, y}

>>> (1 + x + 2*sin(y + I*pi)).atoms(Number)
{1, 2}

>>> (1 + x + 2*sin(y + I*pi)).atoms(Number, NumberSymbol)
{1, 2, pi}

>>> (1 + x + 2*sin(y + I*pi)).atoms(Number, NumberSymbol, I)
{1, 2, I, pi}

```

Note that I (imaginary unit) and ∞ (complex infinity) are special types of number symbols and are not part of the `NumberSymbol` class.

The type can be given implicitly, too:

```

>>> (1 + x + 2*sin(y + I*pi)).atoms(x) # x is a Symbol
{x, y}

```

Be careful to check your assumptions when using the implicit option since `S(1).is_Integer = True` but `type(S(1))` is `One`, a special type of sympy atom, while `type(S(2))` is type `Integer` and will find all integers in an expression:

```

>>> from sympy import S
>>> (1 + x + 2*sin(y + I*pi)).atoms(S(1))
{1}

>>> (1 + x + 2*sin(y + I*pi)).atoms(S(2))
{1, 2}

```

Finally, arguments to `atoms()` can select more than atomic atoms: any sympy type (loaded in `core/__init__.py`) can be listed as an argument and those types of "atoms" as found in scanning the arguments of the expression recursively:

```

>>> from sympy import Function, Mul
>>> from sympy.core.function import AppliedUndef
>>> f = Function('f')
>>> (1 + f(x) + 2*sin(y + I*pi)).atoms(Function)
{f(x), sin(y + I*pi)}
>>> (1 + f(x) + 2*sin(y + I*pi)).atoms(AppliedUndef)
{f(x)}

>>> (1 + x + 2*sin(y + I*pi)).atoms(Mul)
{I*pi, 2*sin(y + I*pi)}

```

compare(self, other)

Return -1, 0, 1 if the object is smaller, equal, or greater than other.

Not in the mathematical sense. If the object is of a different type from the "other" then their classes are ordered according to the `sorted_classes` list.

Examples
=====

```

>>> from sympy.abc import x, y
>>> x.compare(y)
-1
>>> x.compare(x)
0
>>> y.compare(x)
1

```

copy(self)

count(self, query)

Count the number of matching subexpressions.

dummy_eq(self, other, symbol=None)

Compare two expressions and handle dummy symbols.

Examples

=====

```

>>> from sympy import Dummy
>>> from sympy.abc import x, y

>>> u = Dummy('u')

>>> (u**2 + 1).dummy\_eq(x**2 + 1)
True
>>> (u**2 + 1) == (x**2 + 1)
False

>>> (u**2 + y).dummy\_eq(x**2 + y, x)
True
>>> (u**2 + y).dummy\_eq(x**2 + y, y)
False

```

find(self, query, group=False)

Find all subexpressions matching a query.

has(...)

Test whether any subexpression matches any of the patterns.

Examples

=====

```

>>> from sympy import sin
>>> from sympy.abc import x, y, z
>>> (x**2 + sin(x*y)).has(z)
False
>>> (x**2 + sin(x*y)).has(x, y, z)
True
>>> x.has(x)
True

```

Note ``has`` is a structural algorithm with no knowledge of mathematics. Consider the following half-open interval:

```

>>> from sympy.sets import Interval
>>> i = Interval.Lopen(0, 5); i
Interval.Lopen(0, 5)
>>> i.args
(0, 5, True, False)
>>> i.has(4) # there is no "4" in the arguments
False
>>> i.has(0) # there *is* a "0" in the arguments
True

```

Instead, use ``contains`` to determine whether a number is in the interval or not:

```
>>> i.contains(4)
True
>>> i.contains(0)
False
```

Note that ``expr.[has](#)(\*patterns)`` is exactly equivalent to ``any(expr.[has](#)(p) for p in patterns)``. In particular, ``False`` is returned when the list of patterns is empty.

```
>>> x.has()
False
```

is_hypergeometric(self, k)
match(self, pattern, old=False)
Pattern matching.

Wild symbols match all.

Return ``None`` when expression (self) does not match with pattern. Otherwise return a dictionary such that::

```
pattern.xreplace(self.match(pattern)) == self
```

Examples
=====

```
>>> from sympy import Wild
>>> from sympy.abc import x, y
>>> p = Wild("p")
>>> q = Wild("q")
>>> r = Wild("r")
>>> e = (x+y)**(x+y)
>>> e.match(p**p)
{p_: x + y}
>>> e.match(p**q)
{p_: x + y, q_: x + y}
>>> e = (2*x)**2
>>> e.match(p*q**r)
{p_: 4, q_: x, r_: 2}
>>> (p*q**r).xreplace(e.match(p*q**r))
4*x**2
```

The ``old`` flag will give the old-style pattern matching where expressions and patterns are essentially solved to give the match. Both of the following give None unless ``old=True``:

```
>>> (x - 2).match(p - x, old=True)
{p_: 2*x - 2}
>>> (2/x).match(p*x, old=True)
{p_: 2/x**2}
```

rcall(self, *args)

Apply on the argument recursively through the expression tree.

This method is used to simulate a common abuse of notation for operators. For instance in SymPy the the following will not work:

```
``(x+Lambda(y, 2*y))(z) == x+2*z``,
```

however you can use

```
>>> from sympy import Lambda
>>> from sympy.abc import x, y, z
>>> (x + Lambda(y, 2*y)).rcall(z)
x + 2*z
```

`replace(self, query, value, map=False, simultaneous=True, exact=False)`

Replace matching subexpressions of ``self`` with ``value``.

If ``map = True`` then also return the mapping {old: new} where ``old``

was a sub-expression found with query and ``new`` is the replacement value for it. If the expression itself doesn't match the query, then the returned value will be ``self.[xreplace](#)(map)`` otherwise it should be ``self.[subs](#)(ordered(map.items()))``.

Traverses an expression tree and performs replacement of matching subexpressions from the bottom to the top of the tree. The default approach is to do the replacement in a simultaneous fashion so changes made are targeted only once. If this is not desired or causes problems, ``simultaneous`` can be set to False. In addition, if an expression containing more than one Wild symbol is being used to match subexpressions and the ``exact`` flag is True, then the match will only succeed if non-zero values are received for each Wild that appears in the match pattern.

The list of possible combinations of queries and replacement values is listed below:

Examples
=====

Initial setup

```
>>> from sympy import log, sin, cos, tan, Wild, Mul, Add
>>> from sympy.abc import x, y
>>> f = log(sin(x)) + tan(sin(x**2))
```

1.1. type -> type
obj.[replace](#)(type, newtype)

When object of type ``type`` is found, replace it with the result of passing its argument(s) to ``newtype``.

```
>>> f.replace(sin, cos)
log(cos(x)) + tan(cos(x**2))
>>> sin(x).replace(sin, cos, map=True)
(cos(x), {sin(x): cos(x)})
>>> (x*y).replace(Mul, Add)
x + y
```

1.2. type -> func
obj.[replace](#)(type, func)

When object of type ``type`` is found, apply ``func`` to its argument(s). ``func`` must be written to handle the number of arguments of ``type``.

```
>>> f.replace(sin, lambda arg: sin(2*arg))
log(sin(2*x)) + tan(sin(2*x**2))
>>> (x*y).replace(Mul, lambda *args: sin(2*Mul(*args)))
sin(2*x*y)
```

2.1. pattern -> expr

```
obj.replace(pattern(wild), expr(wild))
```

Replace subexpressions matching ``pattern`` with the expression written in terms of the Wild symbols in ``pattern``.

```
>>> a = Wild('a')
>>> f.replace(sin(a), tan(a))
log(tan(x)) + tan(tan(x**2))
>>> f.replace(sin(a), tan(a/2))
log(tan(x/2)) + tan(tan(x**2/2))
>>> f.replace(sin(a), a)
log(x) + tan(x**2)
>>> (x*y).replace(a*x, a)
y
```

When the default value of False is used with patterns that have more than one Wild symbol, non-intuitive results may be obtained:

```
>>> b = Wild('b')
>>> (2*x).replace(a*x + b, b - a)
2/x
```

For this reason, the ``exact`` option can be used to make the replacement only when the match gives non-zero values for all Wild symbols:

```
>>> (2*x + y).replace(a*x + b, b - a, exact=True)
y - 2
>>> (2*x).replace(a*x + b, b - a, exact=True)
2*x
```

2.2. pattern -> func

```
obj.replace(pattern(wild), lambda wild: expr(wild))
```

All behavior is the same as in 2.1 but now a function in terms of pattern variables is used rather than an expression:

```
>>> f.replace(sin(a), lambda a: sin(2*a))
log(sin(2*x)) + tan(sin(2*x**2))
```

3.1. func -> func

```
obj.replace(filter, func)
```

Replace subexpression ``e`` with ``[func](#)(e)`` if ``filter(e)`` is True.

```
>>> g = 2*sin(x**3)
>>> g.replace(lambda expr: expr.is_Number, lambda expr: expr**2)
4*sin(x**9)
```

The expression itself is also targeted by the query but is done in such a fashion that changes are not made twice.

```
>>> e = x*(x*y + 1)
>>> e.replace(lambda x: x.is_Mul, lambda x: 2*x)
```

$2*x*(2*x*y + 1)$

See Also

=====

subs: substitution of subexpressions as defined by the objects themselves.

xreplace: exact node replacement in expr tree; also capable of using matching rules

rewrite(self, *args, **hints)

Rewrite functions in terms of other functions.

Rewrites expression containing applications of functions of one kind in terms of functions of different kind. For example you can rewrite trigonometric functions as complex exponentials or combinatorial functions as gamma function.

As a pattern this function accepts a list of functions to rewrite (instances of DefinedFunction class). As rule you can use string or a destination function instance (in this case [rewrite\(\)](#) will use the `str()` function).

There is also the possibility to pass hints on how to rewrite the given expressions. For now there is only one such hint defined called 'deep'. When 'deep' is set to False it will forbid functions to rewrite their contents.

Examples

=====

```
>>> from sympy import sin, exp
>>> from sympy.abc import x
```

Unspecified pattern:

```
>>> sin(x).rewrite(exp)
-I*(exp(I*x) - exp(-I*x))/2
```

Pattern as a single function:

```
>>> sin(x).rewrite(sin, exp)
-I*(exp(I*x) - exp(-I*x))/2
```

Pattern as a list of functions:

```
>>> sin(x).rewrite([sin, ], exp)
-I*(exp(I*x) - exp(-I*x))/2
```

subs(self, *args, **kwargs)

Substitutes old for new in an expression after sympifying args.

`args` is either:

- two arguments, e.g. `foo.subs(old, new)`
 - one iterable argument, e.g. `foo.subs(iterable)`. The iterable may be
 - o an iterable container with (old, new) pairs. In this case the replacements are processed in the order given with successive patterns possibly affecting replacements already made.
 - o a dict or set whose key/value items correspond to old/new pairs.
- s.
- In this case the old/new pairs will be sorted by op count and in case of a tie, by number of args and the `default_sort_key`. The

resulting sorted list is then processed as an iterable container
(see previous).

If the keyword ``simultaneous`` is True, the subexpressions will not be evaluated until all the substitutions have been made.

Examples
=====

```
>>> from sympy import pi, exp, limit, oo
>>> from sympy.abc import x, y
>>> (1 + x*y).subs(x, pi)
pi*y + 1
>>> (1 + x*y).subs({x:pi, y:2})
1 + 2*pi
>>> (1 + x*y).subs([(x, pi), (y, 2)])
1 + 2*pi
>>> reps = [(y, x**2), (x, 2)]
>>> (x + y).subs(reps)
6
>>> (x + y).subs(reversed(reps))
x**2 + 2
>>> (x**2 + x**4).subs(x**2, y)
y**2 + y
```

To replace only the x^2 but not the x^4 , use `xreplace`:

```
>>> (x**2 + x**4).xreplace({x**2: y})
x**4 + y
```

To delay evaluation until all substitutions have been made, set the keyword ``simultaneous`` to True:

```
>>> (x/y).subs([(x, 0), (y, 0)])
0
>>> (x/y).subs([(x, 0), (y, 0)], simultaneous=True)
nan
```

This has the added feature of not allowing subsequent substitutions to affect those already made:

```
>>> ((x + y)/y).subs({x + y: y, y: x + y})
1
>>> ((x + y)/y).subs({x + y: y, y: x + y}, simultaneous=True)
y/(x + y)
```

In order to obtain a canonical result, unordered iterables are sorted by count, length, number of arguments and by the default `sort_key` to break any ties. All other iterables are left unsorted.

```
>>> from sympy import sqrt, sin, cos
>>> from sympy.abc import a, b, c, d, e

>>> A = (sqrt(sin(2*x)), a)
>>> B = (sin(2*x), b)
>>> C = (cos(2*x), c)
>>> D = (x, d)
```

```
>>> E = (exp(x), e)

>>> expr = sqrt(sin(2*x))*sin(exp(x)*x)*cos(2*x) + sin(2*x)

>>> expr.subs(dict([A, B, C, D, E]))
a*c*sin(d*e) + b
```

The resulting expression represents a literal replacement of the old arguments with the new arguments. This may not reflect the limiting behavior of the expression:

```
>>> (x**3 - 3*x).subs({x: oo})
nan

>>> limit(x**3 - 3*x, x, oo)
oo
```

If the substitution will be followed by numerical evaluation, it is better to pass the substitution to `evalf` as

```
>>> (1/x).evalf(subs={x: 3.0}, n=21)
0.33333333333333333333
```

rather than

```
>>> (1/x).subs({x: 3.0}).evalf(21)
0.3333333333333333333314830
```

as the former will ensure that the desired level of precision is obtained.

See Also
=====

`replace`: replacement capable of doing wildcard-like matching, parsing of match, and conditional replacements
`xreplace`: exact node replacement in expr tree; also capable of using matching rules
`evalf`: calculates the given formula to a desired level of precision

Class methods inherited from [sympy.core.basic.Basic](#):

fromiter(cls, args, **assumptions) from [sympy.core.assumptions.ManagedProperties](#)
 Create a new object from an iterable.

This is a convenience function that allows one to create objects from any iterable, without having to convert to a list or tuple first.

Examples
=====

```
>>> from sympy import Tuple
>>> Tuple.fromiter(i for i in range(5))
(0, 1, 2, 3, 4)
```

Data descriptors inherited from [sympy.core.basic.Basic](#):

args

Returns a tuple of arguments of 'self'.

Examples
=====


```
>>> from sympy import cot
>>> from sympy.abc import x, y
```

```
>>> cot(x).args
(x,)
```

```
>>> cot(x).args[0]
x
```

```
>>> (x*y).args
(x, y)
```

```
>>> (x*y).args[1]
y
```

Notes
=====

Never use `self._args`, always use `self.args`.
Only use `_args` in `__new__` when creating a new function.
Don't override `.args()` from Basic (so that it's easy to change the interface in the future if needed).

canonical_variables

Return a dictionary mapping any variable defined in `self.variables` as underscore-suffixed numbers corresponding to their position in `self.variables`. Enough underscores are added to ensure that there will be no clash with existing free symbols.

Examples
=====

```
>>> from sympy import Lambda
>>> from sympy.abc import x
>>> Lambda(x, 2*x).canonical_variables
{x: 0_}
```

func

The top-level function in an expression.

The following should hold for all objects::

```
>> x == x.func(*x.args)
```

Examples
=====

```
>>> from sympy.abc import x
>>> a = 2*x
>>> a.func
<class 'sympy.core.mul.Mul'>
>>> a.args
(2, x)
>>> a.func(*a.args)
2*x
>>> a == a.func(*a.args)
True
```

is_algebraic
is_antihermitian
is_commutative

is_complex
is_composite
is_even
is_finite
is_hermitian
is_imaginary
is_infinite
is_integer
is_irrational
is_negative
is_noninteger
is_nonnegative
is_nonpositive
is_nonzero
is_odd
is_polar
is_positive
is_prime
is_rational
is_real
is_transcendental
is_zero

Data and other attributes inherited from [sympy.core.basic.Basic](#):

is_Add = False
is_AlgebraicNumber = False
is_Boolean = False
is_Derivative = False
is_Dummy = False
is_Equality = False
is_Float = False
is_Function = False
is_Indexed = False
is_Integer = False
is_Matrix = False
is_Mul = False
is_Not = False
is_Number = False
is_NumberSymbol = False
is_Order = False
is_Piecewise = False
is_Point = False
is_Poly = False
is_Pow = False
is_Rational = False
is_Relational = False
is_Vector = False
is_Wild = False

Methods inherited from [sympy.core.evalf.EvalfMixin](#):

evalf(self, n=15, subs=None, maxn=100, chop=False, strict=False, quad=None, verbose=False)

Evaluate the given formula to an accuracy of n digits.

Optional keyword arguments:

subs=<dict>

Substitute numerical values for symbols, e.g.

subs={x:3, y:1+pi}. The substitutions must be given as a dictionary.

maxn=<integer>

Allow a maximum temporary working precision of maxn digits (default=100)

chop=<bool>

Replace tiny real or imaginary parts in subresults by exact zeros (default=False)

strict=<bool>

Raise PrecisionExhausted if any subresult fails to evaluate to full accuracy, given the available maxprec (default=False)

quad=<str>

Choose algorithm for numerical quadrature. By default, tanh-sinh quadrature is used. For oscillatory integrals on an infinite interval, try quad='osc'.

verbose=<bool>

Print debug information (default=False)

n = evalf(self, n=15, subs=None, maxn=100, chop=False, strict=False, quad=None, verbose=False)

Evaluate the given formula to an accuracy of n digits.

Optional keyword arguments:

subs=<dict>

Substitute numerical values for symbols, e.g.

subs={x:3, y:1+pi}. The substitutions must be given as a dictionary.

maxn=<integer>

Allow a maximum temporary working precision of maxn digits (default=100)

chop=<bool>

Replace tiny real or imaginary parts in subresults by exact zeros (default=False)

strict=<bool>

Raise PrecisionExhausted if any subresult fails to evaluate to full accuracy, given the available maxprec (default=False)

quad=<str>

Choose algorithm for numerical quadrature. By default, tanh-sinh quadrature is used. For oscillatory integrals on an infinite interval, try quad='osc'.

verbose=<bool>

Print debug information (default=False)

DictList

Method resolution order:

[DictList](#)
[__builtin__.list](#)
[__builtin__.object](#)

Methods defined here:

```
__add__(self, other)
    x. \_\_add\_\_ (y) <==> x + y

    Parameters
    -----
    other : iterable
        other must contain only unique id's which do not intersect
        with self

__contains__(self, object)
    DictList.\_\_contains\_\_ (object) <==> object in DictList

    object: str or :class:`~cobra.core.Object.Object`

__copy__(self)
__delitem__(self, index)
__delslice__(self, i, j)
__dir__(self)
__getattr__(self, attr)
__getitem__(self, i)
__getslice__(self, i, j)
__getstate__(self)
    gets internal state

    This is only provided for backwards compatibility so older
    versions of cobrapy can load pickles generated with cobrapy. In
    reality, the "_dict" state is ignored when loading a pickle

__iadd__(self, other)
    x. \_\_iadd\_\_ (y) <==> x += y

    Parameters
    -----
    other : iterable
        other must contain only unique id's whcih do not intersect
        with self

__init__(self, *args)
    Instantiate a combined dict and list.

    Parameters
    -----
    args : iterable
        iterable as single argument to create new DictList from

__isub__(self, other)
    x. \_\_sub\_\_ (y) <==> x -= y

    Parameters
    -----
    other : iterable
        other must contain only unique id's present in the list

__reduce__(self)
```

```

__setitem__(self, i, y)
__setslice__(self, i, j, y)
__setstate__(self, state)
    sets internal state

    Ignore the passed in state and recalculate it. This is only for
    compatibility with older pickles which did not correctly specify
    the initialization class

__sub__(self, other)
    x. sub (y) <==> x - y

    Parameters
    -----
    other : iterable
        other must contain only unique id's present in the list

add(self, x)
    Opposite of `remove`. Mirrors set.add

append(self, object)
    append object to end

extend(self, iterable)
    extend list by appending elements from the iterable

get_by_any(self, iterable)
    Get a list of members using several different ways of indexing

    Parameters
    -----
    iterable : list (if not, turned into single element list)
        list where each element is either int (referring to an index in
        in this DictList), string (a id of a member in this DictList) or
        member of this DictList for pass-through

    Returns
    -----
    list
        a list of members

get_by_id(self, id)
    return the element with a matching id

has_id(self, id)

index(self, id, *args)
    Determine the position in the list

    id: A string or a :class:`~cobra.core.Object.Object`

insert(self, index, object)
    insert object before index

list_attr(self, attribute)
    return a list of the given attribute for every object

pop(self, *args)
    remove and return item at index (default last).

query(self, search_function, attribute=None)
    Query the list

    Parameters
    -----
    search_function : a string, regular expression or function
        Used to find the matching elements in the list.
        - a regular expression (possibly compiled), in which case the
        given attribute of the object should match the regular expression
    .

```

- a function which takes one argument and returns True for desired values

attribute : string or None
the name attribute of the object to passed as argument to the `search\_function`. If this is None, the object itself is used.

Returns

[DictList](#)

a new [list](#) of objects which match the query

Examples

```
>>> import cobra.test
>>> model = cobra.test.create_test_model('textbook')
>>> model.reactions.query(lambda x: x.boundary)
>>> import re
>>> regex = re.compile('^g', flags=re.IGNORECASE)
>>> model.metabolites.query(regex, attribute='name')
```

remove(self, x)

.. warning :: Internal use only

reverse(self)

reverse *IN PLACE*

sort(self, cmp=None, key=None, reverse=False)

stable sort *IN PLACE*

cmp(x, y) -> -1, 0, 1

union(self, iterable)

adds elements with id's not already in the model

Data descriptors defined here:

__dict__

dictionary for instance variables (if defined)

__weakref__

list of weak references to the object (if defined)

Methods inherited from [builtin .list](#):

__eq__(...)

x.[__eq__](#)(y) <==> x==y

__ge__(...)

x.[__ge__](#)(y) <==> x>=y

__getattr__(...)

x.[__getattr__](#)('name') <==> x.name

__gt__(...)

x.[__gt__](#)(y) <==> x>y

__imul__(...)

x.[__imul__](#)(y) <==> x*=y

__iter__(...)

x.[__iter__](#)() <==> iter(x)

__le__(...)

x.[__le__](#)(y) <==> x<=y

__len__(...)

x.[__len__](#)() <==> len(x)

__lt__(...)

x.[__lt__](#)(y) <==> x<y

__mul__(...)

```

    x. mul (n) <==> x*n
__ne__(...)
    x. ne (y) <==> x!=y
__repr__(...)
    x. repr () <==> repr(x)
__reversed__(...)
    L. reversed () -- return a reverse iterator over the list
__rmul__(...)
    x. rmul (n) <==> n*x
__sizeof__(...)
    L. sizeof () -- size of L in memory, in bytes
count(...)
    L. count (value) -> integer -- return number of occurrences of value

```

Data and other attributes inherited from [builtin .list](#):

```

__hash__ = None
__new__ = <built-in method __new__ of type object>
    T. new (S, ...) -> a new object with type S, a subtype of T

```