

# Groebner.jl

The Groebner.jl developers and contributors

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# Contents

<b>Contents</b>	<b>ii</b>
<b>I Home</b>	<b>1</b>
<b>1 Home</b>	<b>2</b>
1.1 Installation . . . . .	2
1.2 Features . . . . .	2
1.3 Contacts . . . . .	2
1.4 Citation . . . . .	2
1.5 Acknowledgement . . . . .	3
<b>II Examples</b>	<b>4</b>
<b>2 Examples</b>	<b>5</b>
2.1 Using AbstractAlgebra.jl . . . . .	5
2.2 Using DynamicPolynomials.jl . . . . .	6
2.3 Using Low-level interface . . . . .	6
2.4 Generic coefficients . . . . .	7
<b>III Interface</b>	<b>11</b>
<b>3 Interface</b>	<b>12</b>
3.1 Main functions . . . . .	12
3.2 Monomial orderings . . . . .	20
3.3 Learn and Apply . . . . .	23

**Part I**

**Home**

# Chapter 1

## Home

Groebner.jl is a package for computing Gröbner bases written in Julia.

### Note

This documentation is also available in PDF format: [Groebner.jl.pdf](#).

### 1.1 Installation

To install Groebner.jl, run the following in the Julia REPL:

```
using Pkg; Pkg.add("Groebner")
```

### 1.2 Features

Groebner.jl features:

- Gröbner basis over integers modulo a prime and over the rationals
- Gröbner trace algorithms
- Multi-threading

### 1.3 Contacts

This library is maintained by Alexander Demin ([asdemin\\_2@edu.hse.ru](mailto:asdemin_2@edu.hse.ru)).

### 1.4 Citation

```
@misc{demin2024groebnerjl,  
  title={Groebner.jl: A package for Gröbner bases computations in Julia},  
  author={Alexander Demin and Shashi Gowda},  
  year={2024},  
  eprint={2304.06935},  
  archivePrefix={arXiv},
```

```
primaryClass={cs.MS}  
}
```

## 1.5 Acknowledgement

We would like to acknowledge the developers of the msolve library (<https://msolve.lip6.fr/>), as several components of Groebner.jl were adapted from msolve. In our F4 implementation, we adapt and adjust the code of monomial hashtable, critical pair handling and symbolic preprocessing, and linear algebra from msolve. The source code of msolve is available at <https://github.com/algebraic-solving/msolve>.

We thank Vladimir Kuznetsov for helpful discussions and providing the sources of his F4 implementation.

We are grateful to The Max Planck Institute for Informatics, The MAX team at I'X, and the OURAGAN team at Inria for providing computational resources.

## **Part II**

### **Examples**

## Chapter 2

# Examples

Groebner.jl supports polynomials from the following frontends:

- AbstractAlgebra.jl
- Nemo.jl
- DynamicPolynomials.jl

Additionally, Groebner.jl has a low-level entry point that accepts raw polynomial data.

### 2.1 Using AbstractAlgebra.jl

First, we import AbstractAlgebra.jl. Then, we create an array of polynomials over a finite field

```
using AbstractAlgebra

R, (x, y, z) = polynomial_ring(GF(2^31 - 1), ["x", "y", "z"])
polys = [x^2 + y + z, x*y + z];
```

```
2-element Vector{AbstractAlgebra.Generic.MPoly{AbstractAlgebra.GFElem{Int64}}}:
 x^2 + y + z
 x*y + z
```

and compute a Gröbner basis with the groebner command

```
using Groebner

basis = groebner(polys)
```

```
4-element Vector{AbstractAlgebra.Generic.MPoly{AbstractAlgebra.GFElem{Int64}}}:
 y^3 + y^2*z + z^2
 x*z + 2147483646*y^2 + 2147483646*y*z
 x*y + z
 x^2 + y + z
```

We can check if a set of polynomials forms a Gröbner basis

```
isgroebner(basis)
```

```
true
```

Groebner.jl also provides several monomial orderings. For example, we can eliminate  $z$  from the above system:

```
ordering = Lex(z) * DegRevLex(x, y) # z > x, y
groebner(polys, ordering=ordering)
```

```
2-element Vector{AbstractAlgebra.Generic.MPoly{AbstractAlgebra.GFElem{Int64}}}:
 x^2 + 2147483646*x*y + y
 x*y + z
```

You can find more information on monomial orderings in Groebner.jl in [Monomial orderings](#).

## 2.2 Using DynamicPolynomials.jl

Computing the Gröbner basis of some system:

```
using DynamicPolynomials, Groebner

@polyvar x1 x2
system = [10*x1*x2^2 - 11*x1 + 10,
          10*x1^2*x2 - 11*x2 + 10]

groebner(system)
```

```
3-element
↳ Vector{DynamicPolynomials.Polynomial{DynamicPolynomials.Commutative{DynamicPolynomials.CreationOrder},
↳ MultivariatePolynomials.Graded{MultivariatePolynomials.LexOrder, Rational{BigInt}}}:
 10//11x2 - 10//11x1 - x2^2 + x1^2
 1//1 - 11//10x2 - 10//11x2^2 + 10//11x1x2 + x2^3
 1//1 - 11//10x1 + x1x2^2
```

## 2.3 Using Low-level interface

Some functions in the interface have a low-level entry point. Low-level functions accept and output "raw" exponent vectors and coefficients. This could be convenient when one does not want to depend on a frontend.

For example,



```

using Groebner

# define {x * y - 1, x^3 + 7 * y^2} modulo 65537 in DRL
ring = Groebner.PolyRing(2, Groebner.DegRevLex(), 65537)
monoms = [ [[1, 1], [0, 0]], [[3, 0], [0, 2]] ]
coeffs = [ [ 1, -1 ], [ 1, 7 ] ]

# compute a GB
gb_monoms, gb_coeffs = Groebner.groebner(ring, monoms, coeffs)

```

```

(Vector{Vector{UInt32}})[[[0x00000001, 0x00000001], [0x00000000, 0x00000000]], [[0x00000000,
↪ 0x00000003], [0x00000002, 0x00000000]], [[0x00000003, 0x00000000], [0x00000000, 0x00000002]]],
↪ Vector{UInt32}[[0x00000001, 0x00010000], [0x00000001, 0x00004925], [0x00000001, 0x00000007]])

```

The list of functions that provide a low-level entry point: `groebner`, `normalform`, `isgroebner`, `groebner_learn`, `groebner_apply`.

Low-level functions may be faster than their user-facing analogues since they bypass data conversions. Low-level functions do not make any specific assumptions on input polynomials, that is, all of these cases are correctly handled: unsorted monomials, non-normalized coefficients, duplicate terms, aliasing memory.

## 2.4 Generic coefficients

The implementation in `Groebner.jl` uses a generic type for coefficients. Hence, in theory, `Groebner.jl` can compute Gröbner bases over any type that behaves like a field.

For the following ground fields `Groebner.jl` runs an efficient native implementation:

- integers modulo a prime,
- rational numbers.

For other ground fields, a possibly slower generic fallback is used. In this case, coefficients of polynomials are treated as black-boxes which implement field operations: `zero`, `one`, `inv`, `==`, `+`, `*`, `-`.

For example, we can compute a Gröbner basis over a univariate rational function field over a finite field:

```

using Groebner, AbstractAlgebra

R, t = GF(101)["t"]
ff = fraction_field(R)
_, (x, y) = ff["x", "y"]

sys = [(t/(t+1)*x*y - t^3, y^2 + t]

gb = groebner(sys)

```

```
2-element
↳ Vector{AbstractAlgebra.Generic.MPoly{AbstractAlgebra.Generic.FracFieldElem{AbstractAlgebra.Generic.Poly{AbstractAlgebra.Generic.Integer}}}
y^2 + t
x + 51*t^2*y
```

Many functions reuse the core implementation, so they can also be used over generic fields:

```
@assert isgroebner(gb)
normalform(gb, x*y)
```

## Computing over floating point intervals

In the following example, we combine low-level interface and generic coefficients.

We are going to compute a basis of the hexapod system over tuples  $(Z_p, \text{Interval})$ : each coefficient is treated as a pair, the first coordinate is a finite field element used for zero testing, and the second coordinate is a floating point interval with some fixed precision, the payload. For floating point arithmetic, we will be using MPFI.jl.

```

using Pkg;
Pkg.add(url="https://gitlab.inria.fr/ckatsama/mpfi.jl")

import Base: +, -, *, zero, iszero, one, isone, inv
using AbstractAlgebra, Groebner, MPFI

PRECISION = 1024 # For MPFI intervals

struct Zp_And_FloatInterval{Zp, FloatInterval}
    a::Zp
    b::FloatInterval
end

# Pretend it is a field and hakuna matata
+(x::Zp_And_FloatInterval, y::Zp_And_FloatInterval) = Zp_And_FloatInterval(x.a + y.a, x.b + y.b)
*(x::Zp_And_FloatInterval, y::Zp_And_FloatInterval) = Zp_And_FloatInterval(x.a * y.a, x.b * y.b)
-(x::Zp_And_FloatInterval, y::Zp_And_FloatInterval) = Zp_And_FloatInterval(x.a - y.a, x.b - y.b)
zero(x::Zp_And_FloatInterval) = Zp_And_FloatInterval(zero(x.a), zero(x.b))
one(x::Zp_And_FloatInterval) = Zp_And_FloatInterval(one(x.a), one(x.b))
inv(x::Zp_And_FloatInterval) = Zp_And_FloatInterval(inv(x.a), inv(x.b))
iszero(x::Zp_And_FloatInterval) = iszero(x.a)
isone(x::Zp_And_FloatInterval) = isone(x.a)

@info "Computing Hexapod over QQ"
c_zp = Groebner.Examples.hexapod(k=AbstractAlgebra.GF(2^30+3));
c_qq = Groebner.Examples.hexapod(k=AbstractAlgebra.QQ);
@time gb_truth = groebner(c_qq);
gbcoeffs_truth = map(f -> collect(coefficients(f)), gb_truth);
@info "Coefficient size (in bits): $(maximum(f -> maximum(c -> log2(abs(numerator(c))) +
-> log2(denominator(c))), f), gbcoeffs_truth)"

@info "Computing Hexapod over (Zp, Interval). Precision = $PRECISION bits"
ring = Groebner.PolyRing(nvars(parent(c_qq[1])), Groebner.DegRevLex(), 0, :generic); # Note
-> :generic

```

```

exps = map(f -> collect(exponent_vectors(f)), c_zp);
cfs_qq = map(f -> collect(coefficients(f)), c_qq);
cfs_zp = map(f -> collect(coefficients(f)), c_zp);
cfs = map(f -> map(c -> Groebner.CoeffGeneric(Zp_And_FloatInterval(c[1], BigInterval(c[2],
↪ precision=PRECISION))), zip(f...)), zip(cfs_zp, cfs_qq));
@time gbexps, gbcoeffs = groebner(ring, exps, cfs);

to_inspect = gbcoeffs[end][end]
@info "
    Inspect one coefficient in the basis:
    Zp          = $(to_inspect.data.a)
    Interval    = $(to_inspect.data.b)
    Diam        = $(diam(to_inspect.data.b))
    Diam (rel)  = $(diam_rel(to_inspect.data.b))"

# Sanity check
all_are_inside(x::Zp_And_FloatInterval, truth) = is_inside(BigInterval(truth; precision=PRECISION),
↪ x.b)
all_are_inside(x::Groebner.CoeffGeneric, truth) = all_are_inside(x.data, truth)
all_are_inside(x::AbstractVector, truth) = all(map(all_are_inside, x, truth))
@assert all_are_inside(gbcoeffs, gbcoeffs_truth)

# Max |midpoint - truth|
max_error(x::Zp_And_FloatInterval, y; rel=false) = abs(mid(x.b) - y) / ifelse(rel, max(abs(y), 0),
↪ 1)
max_error(x::Groebner.CoeffGeneric, y; rel=false) = max_error(x.data, y; rel=rel)
max_error(x::AbstractVector, y::AbstractVector; rel=false) = maximum(map(f -> max_error(f...;
↪ rel=rel), zip(x, y)))
@info "
    Max error      : $(max_error(gbcoeffs, gbcoeffs_truth))
    Max error (rel): $(max_error(gbcoeffs, gbcoeffs_truth; rel=true))"

# Max diameter
max_diam(x::Zp_And_FloatInterval; rel=false) = ifelse(rel, diam_rel(x.b), diam(x.b))
max_diam(x::Groebner.CoeffGeneric; rel=false) = max_diam(x.data; rel=rel)
max_diam(x::AbstractVector; rel=false) = maximum(map(f -> max_diam(f; rel=rel), x))
@info "
    Max diam      : $(max_diam(gbcoeffs))
    Max diam (rel): $(max_diam(gbcoeffs; rel=true))"

```

```

Updating git-repo `https://gitlab.inria.fr/ckatsama/mpfi.jl`
Resolving package versions...
No Changes to `~/work/Groebner.jl/Groebner.jl/docs/Project.toml`
No Changes to `~/work/Groebner.jl/Groebner.jl/docs/Manifest.toml`
[ Info: Computing Hexapod over QQ
3.523621 seconds (2.15 M allocations: 326.380 MiB, 6.96% gc time)
[ Info: Coefficient size (in bits):
↪ 26417.51221660290345963338930851770671711863223901550154467730075030408186732383
[ Info: Computing Hexapod over (Zp, Interval). Precision = 1024 bits
3.780835 seconds (11.60 M allocations: 1.143 GiB, 3.86% gc time, 58.65% compilation time)
└ Info:
|   Inspect one coefficient in the basis:
|   Zp          = 431552538
|   Interval    =
↪ [-558257.1799642934991032546658846510721335157750956599082293646475511502749327011931224642255369266525547349452219
↪ -558257.1799642934991032546658846510721335157750956599082293646475511502749327011931224642255369266525547349452219

```

```

|      Diam      =
↪  1.6720271589196519760068693101141224196639007734362045500004601516707958401504299142339073752692854254678099523958
L      Diam (rel) =
↪  1.6720271589196519760068693101141224196639007734362045500004601516707958401504299142339073752692854254678099523958
└ Info:
|      Max error      :
↪  1.228864326279742133434060342410214649753730878890848130615480485662809941472111e-51
L      Max error (rel):
↪  2.127849791509757036887182607381235177361098814152126217820831601897105585301116e-77
└ Info:
|      Max diam      :
↪  1.2024602113770529916842797302629892806951226766469337193331989384733382206662430059112089072544895998153899916056
L      Max diam (rel):
↪  1.2024602113770529916842797302629892806951226766469337193331989384733382206662430059112089072544895998153899916056

```

However, if we lower MPFI precision to 256 bits, some of the intervals become NaN.

## **Part III**

# **Interface**

## Chapter 3

# Interface

### 3.1 Main functions

Groebner.groebner – Function.

```
groebner(polynomials; options...)
```

Computes a Groebner basis of the ideal generated by polynomials.

#### Arguments

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`.

#### Returns

- `basis`: an array of polynomials, a Groebner basis.

#### Possible Options

- `reduced`: A bool, if the returned basis must be autoreduced and unique. Default is `true`.
- `ordering`: Specifies the monomial ordering. Available monomial orderings are:
  - `InputOrdering()` for inferring the ordering from the given polynomials (default),
  - `Lex(args...)` for lexicographic,
  - `DegLex(args...)` for degree lexicographic,
  - `DegRevLex(args...)` for degree reverse lexicographic,
  - `WeightedOrdering(args...)` for weighted ordering,
  - `ProductOrdering(args...)` for block ordering,
  - `MatrixOrdering(args...)` for matrix ordering.

For details and examples see the corresponding documentation page.

- `certify`: A bool, whether to certify the obtained basis. When this option is `false`, the algorithm is randomized and the result is correct with high probability. When this option is `true`, the result is guaranteed to be correct in case the ideal is homogeneous. Default is `false`.
- `linalg`: A symbol, linear algebra backend. Available options are:

- :auto for the automatic choice (default),
- :deterministic for deterministic sparse linear algebra,
- :randomized for probabilistic sparse linear algebra.
- threaded: The use of multi-threading. Available options are:
  - :auto for the automatic choice (default),
  - :no never use multi-threading,
  - :yes allow the use of multi-threading.

Additionally, it is possible to set the environment variable GROEBNER\_NO\_THREADED to 1 to disable all multi-threading in Groebner.jl. In this case, the environment variable takes precedence over the threaded option.

- monoms: Monomial representation used in the computations. Available options are:
  - :auto for the automatic choice (default),
  - :dense for classic dense exponent vectors,
  - :packed for packed representation.
- modular: Modular computation algorithm. Only has effect when computing basis over rational numbers. Available options are:
  - :auto for the automatic choice (default),
  - :classic\_modular for the classic multi-modular algorithm,
  - :learn\_and\_apply for the learn & apply algorithm.
- seed: The seed for randomization. Default is 42.
- homogenize: Controls the use of homogenization in the algorithm. Available options are:
  - :auto, for the automatic choice (default).
  - :yes, always homogenize the input ideal,
  - :no, never homogenize the input ideal,

### Example

Using DynamicPolynomials.jl:

```
using Groebner, DynamicPolynomials
@polyvar x y
groebner([x*y^2 + x, y*x^2 + y])
```

Using AbstractAlgebra.jl:

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"]
groebner([x*y^2 + x, y*x^2 + y])
```

Using Nemo.jl:

```
using Groebner, Nemo
R, (x, y) = GF(2^30+3)["x", "y"]
groebner([x*y^2 + x, y*x^2 + y])
```

Or, say, in another monomial ordering:

```
# lex with y > x
groebner([x*y^2 + x, y*x^2 + y], ordering=Lex(y, x))

# degree reverse lexicographic
groebner([x*y^2 + x, y*x^2 + y], ordering=DegRevLex())
```

### Notes

- The function is thread-safe.
- For `AbstractAlgebra.jl` and `Nemo.jl`, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and  $\mathbb{Q}\mathbb{Q}$ .
- The default algorithm is probabilistic (with `certify=false`). Results are correct with high probability, however, no precise bound on the probability is known.

### source

`Groebner.isgroebner` – Function.

```
isgroebner(polynomials; options...)
```

Checks if `polynomials` forms a Groebner basis.

### Arguments

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`.

### Returns

- `flag`: a bool, whether `polynomials` is a Groebner basis of the ideal generated by `polynomials`.

### Possible Options

- `ordering`: Specifies the monomial ordering. Available monomial orderings are:
  - `InputOrdering()` for inferring the ordering from the given `polynomials` (default),
  - `Lex()` for lexicographic,
  - `DegLex()` for degree lexicographic,
  - `DegRevLex()` for degree reverse lexicographic,
  - `WeightedOrdering(weights)` for weighted ordering,
  - `ProductOrdering(args...)` for block ordering,
  - `MatrixOrdering(matrix)` for matrix ordering.

For details and examples see the corresponding documentation page.

- `certify`: a bool, whether to use a deterministic algorithm. Default is `false`.
- `seed`: The seed for randomization. Default value is 42.

### Example

Using `DynamicPolynomials`:



```
using Groebner, DynamicPolynomials
@polyvar x y;
isgroebner([x*y^2 + x, y*x^2 + y])
```

Using AbstractAlgebra:

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"]
isgroebner([x*y^2 + x, y*x^2 + y])
```

### Notes

- The function is thread-safe.
- For AbstractAlgebra.jl and Nemo.jl, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and `QQ`.
- The default algorithm is probabilistic (with `certify=false`). Results are correct with high probability, however, no precise bound on the probability is known.

### source

Groebner.normalform – Function.

```
normalform(basis, to_be_reduced; options...)
```

Computes the normal form of polynomials `to_be_reduced` with respect to a Groebner basis `basis`.

### Arguments

- `basis`: an array of polynomials, a Groebner basis. Supports polynomials from AbstractAlgebra.jl, Nemo.jl, and DynamicPolynomials.jl.
- `to_be_reduced`: either a single polynomial or an array of polynomials. Supports polynomials from AbstractAlgebra.jl, Nemo.jl, and DynamicPolynomials.jl.

### Returns

- `reduced`: either a single polynomial or an array of polynomials, the normal forms.

### Possible Options

- `check`: Check if basis forms a Groebner basis. Default is `true`.
- `ordering`: Specifies the monomial ordering. Available monomial orderings are:
  - `InputOrdering()` for inferring the ordering from the given polynomials (default),
  - `Lex()` for lexicographic,
  - `DegLex()` for degree lexicographic,
  - `DegRevLex()` for degree reverse lexicographic,
  - `WeightedOrdering(weights)` for weighted ordering,
  - `ProductOrdering(args...)` for block ordering,

- `MatrixOrdering(matrix)` for matrix ordering.

For details and examples see the corresponding documentation page.

### Example

Finding the normal form a single polynomial:

```
using Groebner, DynamicPolynomials
@polyvar x y;
normalform([y^2 + x, x^2 + y], x^2 + y^2 + 1)
```

Or, reducing two polynomials at a time:

```
using Groebner, DynamicPolynomials
@polyvar x y;
normalform([y^2 + x, x^2 + y], [x^2 + y^2 + 1, x^10*y^10])
```

### Notes

- The function is thread-safe.
- For `AbstractAlgebra.jl` and `Nemo.jl`, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and `QQ`.
- The default algorithm is probabilistic (with `certify=false`). Results are correct with high probability, however, no precise bound on the probability is known.

### source

`Groebner.leading_ideal` – Function.

```
leading_ideal(polynomials; options...)
```

Returns generators of the ideal of the leading terms.

If the input is not a Groebner basis, computes a Groebner basis.

### Arguments

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`.

### Returns

- `basis`: the basis of the ideal of the leading terms.

### Possible Options

- `ordering`: Specifies the monomial ordering. Available monomial orderings are:
  - `InputOrdering()` for inferring the ordering from the given `polynomials` (default),
  - `Lex()` for lexicographic,

- `DegLex()` for degree lexicographic,
- `DegRevLex()` for degree reverse lexicographic,
- `WeightedOrdering(weights)` for weighted ordering,
- `ProductOrdering(args...)` for block ordering,
- `MatrixOrdering(matrix)` for matrix ordering.

For details and examples see the corresponding documentation page.

### Example

Using `AbstractAlgebra.jl`:

```
using Groebner, Nemo
R, (x, y) = QQ["x", "y"]
leading_ideal([x*y^2 + x, y*x^2 + y])
```

### Notes

- The function is thread-safe.
- For `AbstractAlgebra.jl` and `Nemo.jl`, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and `QQ`.

[source](#)

`Groebner.dimension` – Function.

```
dimension(polynomials; options...)
```

Computes the (Krull) dimension of the ideal generated by polynomials.

If input is not a Groebner basis, computes a Groebner basis.

### Arguments

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`.

### Returns

- `dimension`: an integer, the dimension.

### Example

Using `AbstractAlgebra.jl`:

```
using Groebner, Nemo
R, (x, y) = QQ["x", "y"]
dimension([x*y^2 + x, y*x^2 + y])
```

### Notes

- The function is thread-safe.
- For `AbstractAlgebra.jl` and `Nemo.jl`, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and `QQ`.

#### source

`Groebner.quotient_basis` – Function.

```
quotient_basis(polynomials; options...)
```

Returns a monomial basis of the quotient algebra of a zero-dimensional ideal.

If the input is not a Groebner basis, computes a Groebner basis. If the input is not a zero-dimensional ideal, an error is raised.

#### Arguments

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`.

#### Returns

- `basis`: an array of monomials, a quotient basis.

#### Possible Options

- `ordering`: Specifies the monomial ordering. Available monomial orderings are:
  - `InputOrdering()` for inferring the ordering from the given polynomials (default),
  - `Lex()` for lexicographic,
  - `DegLex()` for degree lexicographic,
  - `DegRevLex()` for degree reverse lexicographic,
  - `WeightedOrdering(weights)` for weighted ordering,
  - `ProductOrdering(args...)` for block ordering,
  - `MatrixOrdering(matrix)` for matrix ordering.

For details and examples see the corresponding documentation page.

#### Example

Using `AbstractAlgebra.jl`:

```
using Groebner, Nemo
R, (x, y) = QQ["x", "y"]
quotient_basis([x*y^2 + x, y*x^2 + y])
```

#### Notes

- The function is thread-safe.
- For `AbstractAlgebra.jl` and `Nemo.jl`, the function is most efficient for polynomials over  $\text{GF}(p)$ , `Native.GF(p)`, and `QQ`.

[source](#)

Groebner.groebner\_with\_change\_matrix – Function.

```
groebner_with_change_matrix(polynomials; options...)
```

Computes a Groebner basis of the ideal generated by `polynomials` and emits a change matrix, that is, a map from the original generators to basis elements.

**Arguments**

- `polynomials`: an array of polynomials. Supports polynomials from `AbstractAlgebra.jl`, `Nemo.jl`, and `DynamicPolynomials.jl`. For `AbstractAlgebra.jl` and `Nemo.jl`, coefficients of polynomials must belong to `GF(p)`, `Native.GF(p)`, or `QQ`.

**Returns**

Returns a tuple (`basis`, `matrix`).

- `basis`: an array of polynomials, a Groebner basis.
- `matrix`: a matrix, so that `matrix * polynomials == basis`.

**Possible Options**

Same as for `groebner`.

**Example**

Using `AbstractAlgebra.jl`:

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"]
f = [x*y^2 + x, y*x^2 + y]

g, m = groebner_with_change_matrix(f, ordering=DegRevLex())

@assert isgroebner(g, ordering=DegRevLex())
@assert m * f == g
```

**Notes**

- Only `DegRevLex` ordering is supported.
- The function is thread-safe.
- The default algorithm is probabilistic (with `certify=false`). Results are correct with high probability, however, no precise bound on the probability is known.

[source](#)

### 3.2 Monomial orderings

#### Note

Some frontends, for example, AbstractAlgebra.jl, may not support weighted/product/matrix orderings from Groebner.jl. In such cases, the basis is computed in the ordering requested by user, but the terms of polynomials in the output are ordered w.r.t. some other ordering that is supported by the frontend.

Groebner.Lex – Type.

```
Lex()
Lex(variables)
Lex(variables...)
```

Lexicographical monomial ordering.

We use the definition from Chapter 1, Computational Commutative Algebra 1, by Martin Kreuzer, Lorenzo Robbiano. DOI: <https://doi.org/10.1007/978-3-540-70628-1>.

Dura Lex, sed Lex.

#### Example

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"];

# Lexicographical ordering with x > y
groebner([x*y + x, x + y^2], ordering=Lex())

# Lexicographical ordering with y > x
groebner([x*y + x, x + y^2], ordering=Lex([y, x]))

# Lexicographical ordering with x > y
# Both syntax are allowed -- Lex([...]) and Lex(...)
groebner([x*y + x, x + y^2], ordering=Lex(x, y))
```

[source](#)

Groebner.DegLex – Type.

```
DegLex()
DegLex(variables)
DegLex(variables...)
```

Degree lexicographical monomial ordering.

We use the definition from Chapter 1, Computational Commutative Algebra 1, by Martin Kreuzer, Lorenzo Robbiano. DOI: <https://doi.org/10.1007/978-3-540-70628-1>.

#### Example

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"];

# Degree lexicographical ordering with x > y
groebner([x*y + x, x + y^2], ordering=DegLex())

# Degree lexicographical ordering with y > x
groebner([x*y + x, x + y^2], ordering=DegLex([y, x]))
```

[source](#)

Groebner.DegRevLex – Type.

```
DegRevLex()
DegRevLex(variables)
DegRevLex(variables...)
```

Degree reverse lexicographical monomial ordering.

We use the definition from Chapter 1, Computational Commutative Algebra 1, by Martin Kreuzer, Lorenzo Robbiano. DOI: <https://doi.org/10.1007/978-3-540-70628-1>.

### Example

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"];

# Degree reverse lexicographical ordering with x > y
groebner([x*y + x, x + y^2], ordering=DegRevLex())

# Degree reverse lexicographical ordering with y > x
groebner([x*y + x, x + y^2], ordering=DegRevLex(y, x))
```

[source](#)

Groebner.InputOrdering – Type.

```
InputOrdering()
```

Preserves the monomial ordering defined on the input polynomials.

This is the default value for the ordering keyword argument.

### Example

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"]

# Uses the ordering `InputOrdering`, which, in this case,
# defaults to the lexicographical ordering with x > y
groebner([x*y + x, x + y^2])
```

[source](#)

Groebner.WeightedOrdering - Type.

```
WeightedOrdering(weights)
```

Weighted monomial ordering.

Only positive weights are supported.

**Example**

```
using Groebner, AbstractAlgebra
R, (x, y) = QQ["x", "y"];

# x has weight 3, y has weight 1
ord = WeightedOrdering(x => 3, y => 1)
groebner([x*y + x, x + y^2], ordering=ord)
```

[source](#)

Groebner.ProductOrdering - Type.

```
ProductOrdering(ord1, ord2)
```

Product monomial ordering. Compares by ord1, breaks ties by ord2.

Can also be constructed with \*.

**Example**

```
using Groebner, AbstractAlgebra
R, (x, y, z, w) = QQ["x", "y", "z", "w"];

# Ordering with x, y > w, z
ord = ProductOrdering(DegRevLex(x, y), DegRevLex(w, z))
groebner([x*y + w, y*z - w], ordering=ord)
```

It is possible to use the \* operator:

```
using Groebner, AbstractAlgebra
R, (x, y, z, t) = QQ["x", "y", "z", "t"];

ord1 = Lex(t)
ord2 = DegRevLex(x, y, z)
# t >> x, y, z
ord = ord1 * ord2
groebner([x*y*z + z, t * z - 1], ordering=ord)
```

[source](#)



Groebner.MatrixOrdering – Type.

```
MatrixOrdering(matrix)
MatrixOrdering(Vector{Vector})
```

Matrix monomial ordering.

### Example

```
using Groebner, AbstractAlgebra
R, (x, y, z, w) = QQ["x", "y", "z", "w"];

# the number of columns equal to the number of variables
ord = MatrixOrdering(
    [x,y,z,w],
    [
        1 0 0 2;
        0 0 1 2;
        0 1 1 1;
    ])
groebner([x*y + w, y*z - w], ordering=ord)
```

[source](#)

## 3.3 Learn and Apply

Groebner.groebner\_learn – Function.

```
groebner_learn(polynomials; options...)
```

Computes a Groebner basis of the ideal generated by polynomials and emits a trace.

The trace can be used to speed up the computation of Groebner bases of specializations of the same ideal as the one groebner\_learn had been applied to.

See also groebner\_apply!.

### Arguments

- polynomials: an array of polynomials. Must be polynomials from AbstractAlgebra.jl or Nemo.jl over GF(p) or Native.GF(p).

### Returns

Returns a tuple (trace, basis).

- trace: an object, a trace. Can be used in groebner\_apply!.
- basis: an array of polynomials, a Groebner basis.

**Possible Options**

Same as for groebner.

**Example**

Using groebner\_learn and groebner\_apply! over the same ground field:

```
using Groebner, AbstractAlgebra
R, (x, y) = GF(2^31-1)["x", "y"]

# Learn
trace, gb_1 = groebner_learn([x*y^2 + x, y*x^2 + y])

# Apply (same support, different coefficients)
flag, gb_2 = groebner_apply!(trace, [2x*y^2 + 3x, 4y*x^2 + 5y])

@assert flag
```

Using groebner\_learn and groebner\_apply! over different ground fields:

```
using Groebner, AbstractAlgebra
R, (x, y) = GF(2^31-1)["x", "y"]

# Learn
trace, gb_1 = groebner_learn([x*y^2 + x, y*x^2 + y], ordering=DegRevLex())

# Create a ring with a different modulo
R2, (x2, y2) = GF(2^30+3)["x", "y"]

# Apply (different modulo)
flag, gb_2 = groebner_apply!(
    trace,
    [2x2*y2^2 + 3x2, 4y2*x2^2 + 5y2],
    ordering=DegRevLex()
)

@assert flag
@assert gb_2 == groebner([2x2*y2^2 + 3x2, 4y2*x2^2 + 5y2], ordering=DegRevLex())
```

Using groebner\_apply! in batches:

```
using Groebner, AbstractAlgebra
R, (x, y) = polynomial_ring(GF(2^31-1), ["x", "y"], internal_ordering=:degrevlex)

# Learn
trace, gb_1 = groebner_learn([x*y^2 + x, y*x^2 + y])

# Create rings with some other moduli
R2, (x2, y2) = polynomial_ring(GF(2^30+3), ["x", "y"], internal_ordering=:degrevlex)
R3, (x3, y3) = polynomial_ring(GF(2^27+29), ["x", "y"], internal_ordering=:degrevlex)

# Two specializations of the same ideal
batch = ([2x2*y2^2 + 3x2, 4y2*x2^2 + 5y2], [4x3*y3^2 + 4x3, 5y3*x3^2 + 7y3])
```

```
# Apply for two sets of polynomials at once
flag, (gb_2, gb_3) = groebner_apply!(trace, batch)

@assert flag
@assert (gb_2, gb_3) == map(groebner, batch)
```

Perhaps, in a more involved example, we will compute Groebner bases of the Katsura-9 system:

```
using Groebner, AbstractAlgebra, BenchmarkTools

# Create the system
kat = Groebner.Examples.katsuran(9, k=ZZ, internal_ordering=:degrevlex)

# Reduce the coefficients modulo 5 different primes
kat_0 = map(f -> map_coefficients(c -> GF(2^30 + 3)(c), f), kat)
kat_1 = map(f -> map_coefficients(c -> GF(2^30 + 7)(c), f), kat)
kat_2 = map(f -> map_coefficients(c -> GF(2^30 + 9)(c), f), kat)
kat_3 = map(f -> map_coefficients(c -> GF(2^30 + 15)(c), f), kat)
kat_4 = map(f -> map_coefficients(c -> GF(2^30 + 19)(c), f), kat)

# Learn the trace
trace, gb_0 = groebner_learn(kat_0);

# Compare the performance of applying with 1 input and with 4 different inputs:

# Apply for one system
@btime groebner_apply!($trace, $kat_1);
# 46.824 ms (19260 allocations: 24.48 MiB)

# Apply for a batch of four systems
@btime groebner_apply!($trace, $(kat_1, kat_2, kat_3, kat_4));
# 72.813 ms (23722 allocations: 59.44 MiB)
```

Observe the better amortized performance of the composite `groebner_apply!`.

### Notes

- The function is thread-safe.

### source

`Groebner.groebner_apply!` – Function.

```
groebner_apply!(trace, polynomials; options...)
groebner_apply!(trace, batch::NTuple{N, Vector}; options...)
```

Computes a Groebner basis of the ideal generated by polynomials following the given trace.

See also `groebner_learn`.

### Arguments

- `trace`: a trace produced by `groebner_learn`.

- `polynomials`: an array of polynomials. Must be polynomials from `AbstractAlgebra.jl` or `Nemo.jl` over  $\text{GF}(p)$  or `Nemo.GF(p)`. It is possible to supply a tuple of  $N$  arrays of polynomials to compute  $N$  Groebner bases simultaneously. This could be more efficient overall than computing them in separate.

**Returns**

Returns a tuple `(success, basis)`.

- `success`: a bool, whether the call succeeded.
- `basis`: an array of polynomials, a Groebner basis.

**Possible Options**

The `groebner_apply!` function automatically inherits most parameters from the given `trace`.

**Example**

For examples, see the documentation of `groebner_learn`.

**Notes**

- In general, `success` may be a false positive. The probability of a false positive is considered to be low enough in some practical applications.
- This function is **not** thread-safe since it mutates `trace`.
- This function is **not** safe against program interruptions. For example, pressing `Ctrl + C` while `groebner_apply!(trace, ...)` is running may leave `trace` corrupted.

[source](#)