

A package for combinatorial probability theory

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Abstract

We present a package for combinatorial manipulation of probability distributions, moments, cumulants, orthogonal polynomials and convolutions. The central object is the **FriCAS** domain **Distribution**, which represents probability distributions as streams of moments and cumulants.

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1 A few words about Axiom/FriCAS

The axiom clones (Axiom/FriCAS/OpenAxiom) are the only CAS supporting streams, i.e., lists of indefinite length, which allow to compute with the potential infinite (in the sense of Aristotle), and delayed computation (things are computed when needed). All this is necessary to compute with infinite sequences like moments and cumulants.

FriCAS can do most things other systems can do

1a $\langle \text{example1.input 1a} \rangle \equiv$ 1b
 $p := 1+x+x^2$
$$x^2 + x + 1 \tag{1}$$

Type: Polynomial(Integer)

like differentiation

1b $\langle \text{example1.input 1a} \rangle + \equiv$ <1a 1c>
 $D(p,x)$
$$2x + 1 \tag{2}$$

Type: Polynomial(Integer)

and integration (it has the most complete implementation of the Risch algorithm):

1c $\langle \text{example1.input 1a} \rangle + \equiv$ <1b 2a>
 $\text{integrate}(p,x)$

$$\frac{1}{3} x^3 + \frac{1}{2} x^2 + x \quad (3)$$

Type: Polynomial(Fraction(Integer))

The main difference to other CAS is *strong typing*, i.e., every object has a type. This is very handy when it comes to computing with non-standard things like finite fields or noncommutative rings:

2a `<example1.input 1a>+≡` `<1c 2b>`
`FG:= FreeGroup Symbol`
`a:FG := 'a`
`b:FG:= 'b`

`a*b*a^-1 *b^-1`

$$a b a^{(-1)} b^{(-1)} \quad (4)$$

Type: FreeGroup(Symbol)

Domains can be built on top of each other:

2b `<example1.input 1a>+≡` `<2a 2c>`
`FGA:=MonoidRing (Integer, FG)`
`a1:= a::FGA`
`a1i:= (a^-1)::FGA`
`b1:= b::FGA`
`b1i:= (b^-1)::FGA`
`T:=a1+a1i+b1+b1i`

$$b^{(-1)} + a^{(-1)} + a + b \quad (5)$$

Type: MonoidRing(Integer,FreeGroup(Symbol))

Through the type system, FriCAS knows which multiplication it is meant to perform:

2c `<example1.input 1a>+≡` `<2b 2d>`
`T^2`

$$b^2 + b a + b a^{(-1)} + a b + a^2 + a b^{(-1)} + a^{(-1)} b + a^{(-2)} + a^{(-1)} b^{(-1)} + 4 + b^{(-1)} a + b^{(-1)} a^{(-1)} + b^{(-2)} \quad (6)$$

Type: MonoidRing(Integer,FreeGroup(Symbol))

2d `<example1.input 1a>+≡` `<2c 2e>`
`A:Matrix FGA := matrix [[a1,a1i],[b1,b1i]]`

$$\begin{bmatrix} a & a^{(-1)} \\ b & b^{(-1)} \end{bmatrix} \quad (7)$$

Type: Matrix(MonoidRing(Integer,FreeGroup(Symbol)))

2e `<example1.input 1a>+≡` `<2d 2f>`
`A*A`

$$\begin{bmatrix} a^{(-1)} b + a^2 & a^{(-1)} b^{(-1)} + 1 \\ 1 + b a & b^{(-2)} + b a^{(-1)} \end{bmatrix} \quad (8)$$

Type: Matrix(MonoidRing(Integer,FreeGroup(Symbol)))

Types can be built on each other only if it makes sense:

2f `<example1.input 1a>+≡` `<2e 2g>`
`B:Matrix String := matrix [{"a"}, {"b"}], [{"c"}, {"d"}]`

Matrix(String) is not a valid type.

FriCAS allows infinite objects (potentially infinite in the sense of Aristoteles):

2g `<example1.input 1a>+≡` `<2f 3a>`
`nat:= [n for n in 1..]`

$$[1, 2, 3, 4, 5, 6, 7, 8, 9, 10, \dots] \quad (9)$$

Type: Stream(PositiveInteger)

3a `<example1.input 1a>+≡` `<2g 3b>`
`prim:=[n for n in nat | prime? n]`

$$[2, 3, 5, 7, 11, 13, 17, 19, 23, 29, \dots] \quad (10)$$

Type: Stream(PositiveInteger)

Streams are used for the representation of power series, this way we don't have to worry about the number of terms we want to compute

3b `<example1.input 1a>+≡` `<3a 3c>`
`exps := series(exp x, x=0)`

$$1+x+\frac{1}{2}x^2+\frac{1}{6}x^3+\frac{1}{24}x^4+\frac{1}{120}x^5+\frac{1}{720}x^6+\frac{1}{5040}x^7+\frac{1}{40320}x^8+\frac{1}{362880}x^9+\frac{1}{3628800}x^{10}+O(x^{11}) \quad (11)$$

Type: UnivariatePuisseuxSeries(Expression(Integer),x,0)

The number of terms computed by default is controlled with the command

3c `<example1.input 1a>+≡` `<3b 3d>`
`)set stream calculate 5`
`exps := series(exp x, x=0)`

$$1+x+\frac{1}{2}x^2+\frac{1}{6}x^3+\frac{1}{24}x^4+\frac{1}{120}x^5+O(x^6) \quad (12)$$

Type: UnivariatePuisseuxSeries(Expression(Integer),x,0)

Yet we can compute as far as we want

3d `<example1.input 1a>+≡` `<3c>`
`coefficient(exps,50)`

Type: Expression(Integer)

2 Usage of the Distribution package

FriCAS can be extended by the user with her own data types. In our case, a probability distribution is represented by its sequence of moments

$$[m_1, m_2, \dots]$$

over a ring. With this package, various kinds of cumulants can be computed (classical, free and boolean), as well as classical, free additive and multiplicative, boolean and monotone convolutions and orthogonal polynomials. You can see all available commands by typing

```
)sh Distribution
```

which currently shows the following:

```
Distribution(R: CommutativeRing) is a domain constructor
Abbreviation for Distribution is DISTR0
This constructor is exposed in this frame.
----- Operations -----
=? : (%,% ) -> Boolean          0 : () -> %
?^? : (% ,PositiveInteger) -> %  booleanConvolution : (% ,%) -> %
coerce : % -> OutputForm        freeConvolution : (% ,%) -> %
freeCumulants : % -> Sequence(R) hash : % -> SingleInteger
latex : % -> String             moments : % -> Sequence(R)
?~=? : (% ,%) -> Boolean
booleanCumulant : (% ,PositiveInteger) -> R
booleanCumulantFromJacobi : (Integer ,Sequence(R) ,Sequence(R)) -> R
booleanCumulants : % -> Sequence(R)
classicalConvolution : (% ,%) -> %
classicalCumulant : (% ,PositiveInteger) -> R
classicalCumulants : % -> Sequence(R)
construct : (Sequence(R) ,Sequence(R) ,Sequence(R) ,Sequence(R)) -> %
distributionByBooleanCumulants : Stream(R) -> %
distributionByBooleanCumulants : Sequence(R) -> %
distributionByClassicalCumulants : Stream(R) -> %
distributionByClassicalCumulants : Sequence(R) -> %
distributionByEvenMoments : Stream(R) -> %
distributionByEvenMoments : Sequence(R) -> %
distributionByFreeCumulants : Stream(R) -> %
distributionByFreeCumulants : Sequence(R) -> %
distributionByJacobiParameters : (Stream(R) ,Stream(R)) -> %
distributionByJacobiParameters : (Sequence(R) ,Sequence(R)) -> %
distributionByMoments : Stream(R) -> %
distributionByMoments : Sequence(R) -> %
distributionBySTransform : (Fraction(Integer) ,Fraction(Integer) ,Sequence(R)) -> % if R has ALGEBRA(FRAC(INT))
distributionBySTransform : Record(puiseux: Fraction(Integer) ,laurent: Fraction(Integer) ,coef: Sequence(R)) -> %
freeCumulant : (% ,PositiveInteger) -> R
freeMultiplicativeConvolution : (% ,%) -> % if R has ALGEBRA(FRAC(INT))
hankelDeterminants : % -> Stream(R)
jacobiParameters : % -> Record(an: Stream(Fraction(R)) ,bn: Stream(Fraction(R))) if R has INTDOMAIN
jacobiParameters : % -> Record(an: Stream(R) ,bn: Stream(R)) if R has FIELD
```

```

moment : (% , NonNegativeInteger) -> R
monotoneConvolution : (% , %) -> %
monotoneCumulants : % -> Sequence(R) if R has ALGEBRA(FRAC(INT))
orthogonalPolynomials : % -> Stream(SparseUnivariatePolynomial(Fraction(R))) if R has INTDOM and
orthogonalPolynomials : % -> Stream(SparseUnivariatePolynomial(R)) if R has FIELD

```

Other commands are in different packages

```

)sh STransformPackage
)sh DistributionPolynomialPackage
)sh DistributionContinuedFractionPackage
)sh DISTEX DistributionPackage

```

All parameters are modeled as **Streams**, which has the following advantages:

- Calculation of moments and cumulants is delayed until demand
- once calculated, they are cached.
- the user does not have to worry about a maximal order of moments beforehand, the system computes as far as time and RAM restrictions of the underlying LISP and hardware allow it.

Once an interesting sequence is found, the **GUESS** package by M. Rubey ([arXiv:math/0702086](https://arxiv.org/abs/math/0702086), also part of **FriCAS**) can be used to guess a recurrence or closed formula.

2.1 Example: Creating distributions via cumulant and moment sequences

Say we want to check whether the normal distribution is infinite divisible with respect to free convolution. There are several ways to construct the normal distribution.

1. The quickest way to get the normal distribution is via its classical cumulants, because only the second cumulant is nonzero. We need a declaration of an integer stream, because by default a nonnegative stream is generated and nonnegative integers do not form a ring.

```

5a <example2.input 5a>≡ 5b>
    gausscum:Stream Integer := concat([0,1], repeating [0])

                                [0, 1, 0̄] (14)
                                Type: Stream(Integer)

```

```

5b <example2.input 5a>+≡ <5a 5c>
    gauss:=distributionByClassicalCumulants gausscum

                                [0, 1, 0, 3, 0, 15, 0, 105, 0, 945, ...] (15)
                                Type: Distribution(Integer)

```

2. By the explicit formula for its moments $m_{2k} = \frac{2k!}{2^k k!}$:

```

5c <example2.input 5a>+≡ <5b 6a>
    gaussmom k ==
      odd? k => return 0
      k2:= exquo(k,2)
      return (factorial(k)/(2^k2*factorial k2))::Integer

    gauss:=distributionByMoments [gaussmom k for k in 1..]

```

$$[0, 1, 0, 3, 0, 15, 0, 105, 0, 945, \dots] \quad (16)$$

Type: Distribution(Integer)

3. By dissecting the Taylor series of the exponential moment generating function $e^{z^2/2}$:

6a <5c 6b>

```

<example2.input 5a>+≡
  gaussexp:= series(exp(z^2/2), z=0)

```

$$1 + \frac{1}{2} z^2 + \frac{1}{8} z^4 + \frac{1}{48} z^6 + \frac{1}{384} z^8 + \frac{1}{3840} z^{10} + O(z^{11}) \quad (17)$$

Type: UnivariatePuisseuxSeries(Expression(Integer),z,0)

Next we must extract the coefficients of this series

6b <6a 6c>

```

<example2.input 5a>+≡
  coefficients gaussexp

```

$$\left[1, 0, \frac{1}{2}, 0, \frac{1}{8}, 0, \frac{1}{48}, 0, \frac{1}{384}, 0, \dots \right] \quad (18)$$

Type: Stream(Expression(Integer))

and multiply term by term with $n!$:

6c <6b 6d>

```

<example2.input 5a>+≡
  gaussmom := [c*factorial n for n in 1.. for c in coefficients gaussexp]

```

$$[1, 0, 3, 0, 15, 0, 105, 0, 945, 0, \dots] \quad (19)$$

Type: Stream(Expression(Integer))

4. As moments of the position operator in the Weyl algebra of differential operators, which is already implemented in FricAS.

6d <6c 6e>

```

<example2.input 5a>+≡
  R:= UP(z,Integer)
  W:=LOD01 R

```

LinearOrdinaryDifferentialOperator1(Fraction(UnivariatePolynomial(z,Integer)))
(20)

Now z and d verify the canonical commutation relation:

6e <6d 6f>

```

<example2.input 5a>+≡
  z:R:= 'z
  d:W:= D()
  d*z-z*d

```

$$1 \quad (21)$$

Type:

LinearOrdinaryDifferentialOperator1(Fraction(UnivariatePolynomial(z,Integer)))

we can compute the moments in this algebra by extracting the constant coefficient:

6f <6e 6g>

```

<example2.input 5a>+≡
  g6:=(d+z)^6

```

(10)

$$D^6 + 6z D^5 + (15z^2 + 15) D^4 + (20z^3 + 60z) D^3 + (15z^4 + 90z^2 + 45) D^2 + (6z^5 + 60z^3 + 90z) D + z^6 + 15$$

Type: LinearOrdinaryDifferentialOperator1(UnivariatePolynomial(z,Integer))

The Weyl algebra is generated by D with coefficient ring $\mathbf{C}[z]$:

6g <6f 7a>

```

<example2.input 5a>+≡
  coefficient(g6, 0)

```

$$(12) \quad z^6 + 15 z^4 + 45 z^2 + 15$$

Type: UnivariatePolynomial(z,Integer)

Therefore we must extract the constant coefficient of this polynomial:

7a `<example2.input 5a>+≡` <6g 7b>
`coefficient(coefficient(g6, 0), 0)`

and finally we obtain the gaussian moments

7b `<example2.input 5a>+≡` <7a 7c>
`g:= distributionByMoments([coefficient(coefficient((d+z)^k, 0), 0) for k in 1..])`

$$(13) \quad [0, 1, 0, 3, 0, 15, 0, 105, 0, 945, \dots]$$

Type: Distribution(Integer)

5. By its jacobi parameters:

7c `<example2.input 5a>+≡` <7b 7d>
`an:Stream Integer := repeating [0]`

$$[0] \tag{22}$$

Type: Stream(Integer)

7d `<example2.input 5a>+≡` <7c 7e>
`bn:Stream Integer := [k for k in 1..]`

$$[1, 2, 3, 4, 5, 6, 7, 8, 9, 10, \dots] \tag{23}$$

Type: Stream(Integer)

7e `<example2.input 5a>+≡` <7d 7f>
`gauss:=distributionByJacobiParameters(an, bn)`

$$[0, 1, 0, 3, 0, 15, 0, 105, 0, 945, \dots] \tag{24}$$

Type: Distribution(Integer)

6. In the case of the normal distribution we can also use the built-in function

7f `<example2.input 5a>+≡` <7e 7h>
`gauss:=gaussianDistribution 1`

$$[0, 1, 0, 3, 0, 15, 0, 105, 0, 945, \dots] \tag{25}$$

Type: Distribution(Integer)

Infinite divisibility means that the sequence of free cumulants,

7g `<example.input 7g>≡`
`freeCumulants gauss`

$$[0, 1, 0, 1, 0, 4, 0, 27, 0, 248, \dots] \tag{26}$$

Type: Sequence(Integer)

starting with the third element, is positive definite and thus forms the moment sequence of the Levy measure:

7h `<example2.input 5a>+≡` <7f 8a>
`levymom:= rest rest freeCumulants gauss;`
`levy:= distributionByMoments levymom`

$$[0, 1, 0, 4, 0, 27, 0, 248, 0, 2830, \dots] \quad (27)$$

Type: Distribution(Integer)

Now we need to check whether the Hankel determinants are nonnegative:

8a `<example2.input 5a>+≡` `<7h 8b>`
`hankelDeterminants(levy)`
 $[1, 3, 33, 1837, 586337, 1319746279, 23073023457505, 3650976798436751385, 5654872627743587044800033, 9678677$
(28)

Type: Stream(Integer)

Equivalently, we can check the positivity of the Jacobi parameters:

8b `<example2.input 5a>+≡` `<8a`
`jacobiParameters levy`

$$an = [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, \dots],$$

$$bn = \left[1, 3, \frac{11}{3}, \frac{167}{33}, \frac{10533}{1837}, \frac{4134779}{586337}, \frac{10250885015}{1319746279}, \frac{208831020924783}{23073023457505}, \frac{11912337984453599843}{1216992266145583795}, \frac{69432012544677}{62831918086039}, \dots \right]$$

Type: Record(an: Stream(Fraction(Integer)),bn: Stream(Fraction(Integer)))

2.2 Example: Convolutions

Computing convolutions is easy, using the commands `classicalConvolution`, `freeConvolution`, `freeMultiplicativeConvolution`,...

We investigate the behaviour of Jacobi Parameters under free convolution with a Wigner distribution.

8c `<example3.input 8c>≡` `8d>`
`w:=wignerDistribution t`
 $[0, t, 0, 2 t^2, 0, 5 t^3, 0, 14 t^4, 0, 42 t^5, \dots]$ (30)

Type: Distribution(Polynomial(Integer))

8d `<example3.input 8c>+≡` `<8c 8e>`
`x:=distributionByJacobiParameters([a[k] for k in 0..], [b[k] for k in 0..])`
 $[a_0, b_0 + a_0^2, (a_1 + 2 a_0) b_0 + a_0^3, \dots]$ (31)

Type: Distribution(Polynomial(Integer))

8e `<example3.input 8c>+≡` `<8d 8f>`
`xw:=freeConvolution(x,w)`
 $[a_0, t + b_0 + a_0^2, 3 a_0 t + (a_1 + 2 a_0) b_0 + a_0^3, \dots]$ (32)

Type: Distribution(Polynomial(Integer))

8f `<example3.input 8c>+≡` `<8e ??>`
`jacobiParameters xw`

$$an = \left[a_0, \frac{a_0 t + a_1 b_0}{t + b_0}, \frac{a_0 t^4 + (a_1 + 3 a_0) b_0 t^3 + ((a_2 + 2 a_1 - 2 a_0) b_0 b_1 + 5 a_0 b_0^2 + (a_1^3 - 2 a_0 a_1^2 + a_0^2 a_1))}{t^4 + 4 b_0 t^3 + (b_0 b_1 + 5 b_0^2 + (a_1^2 - 2 a_0 a_1 + a_0^2)) t} \right]$$

$$bn = \left[t + b_0, \frac{t^3 + 3 b_0 t^2 + (b_0 b_1 + 2 b_0^2 + (a_1^2 - 2 a_0 a_1 + a_0^2) b_0) t + b_0^2 b_1}{t^2 + 2 b_0 t + b_0^2}, \frac{t^7 + 7 b_0 t^6 + (3 b_0 b_1 + 18 b_0^2 + \dots)}{\dots} \right]$$
 (33)

Type: Record(an: Stream(Fraction(Polynomial(Integer))),bn: Stream(Fraction(Polynomial(Integer))))

2.3 Example: Orthogonal polynomials

We verify that up to a rescaling the orthogonal polynomials of the Wigner distribution are indeed the Chebyshev polynomials, which can be obtained with the command `chebyshevU`:

```
9a <example4.input 9a>≡ 9b>
    ch2:=orthogonalPolynomials(w)
    [?, ?2 - 1, ?3 - 2 ?, ?4 - 3 ?2 + 1, ?5 - 4 ?3 + 3 ?, ...] (34)
```

Type: Stream(SparseUnivariatePolynomial(Fraction(Integer)))

To give the unknown variable a name, convert it to the polynomial domain $\mathbf{Z}[t]$:

```
9b <example4.input 9a>+≡ <9a 9c>
    ch2::Stream UP(t,Integer)
    [t, t2 - 1, t3 - 2 t, t4 - 3 t2 + 1, t5 - 4 t3 + 3 t, ...] (35)
```

Type: Stream(UnivariatePolynomial(t,Integer))

```
9c <example4.input 9a>+≡ <9b 9d>
    chu:=[chebyshevU(k,t/2) for k in 1..]
    [t, t2 - 1, t3 - 2 t, t4 - 3 t2 + 1, t5 - 4 t3 + 3 t, ...] (36)
```

Type: Stream(Polynomial(Fraction(Integer)))

Now we compare the first ten polynomials:

```
9d <example4.input 9a>+≡ <9c>
    ch2a:=ch2::Stream UP(t,Integer)
    chua:=chu::Stream UP(t,Integer)
    (entries complete first(ch2a, 10) = entries complete first(chua, 10))::Boolean
    true (37)
```

Type: Boolean