

# Analyzing Quantum Many-Body Systems with ITensor and PastaQ

Matthew Fishman

Center for Computational Quantum Physics (CCQ)

Flatiron Institute, NY

<https://mtfishman.github.io/>

February 26, 2022



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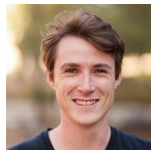
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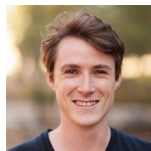
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- ▶ We are hiring postdocs, full-time scientists, part-time and full-time software developers, interns, etc.

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SIMONS FOUNDATION

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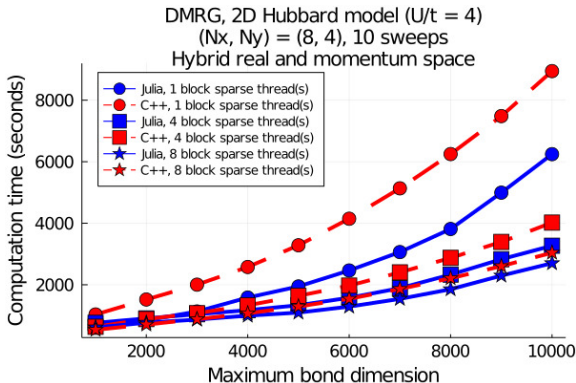
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- ▶ Great numerical libraries, code ended up faster.



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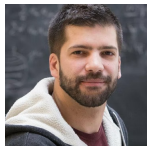
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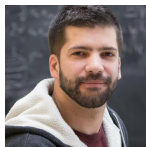
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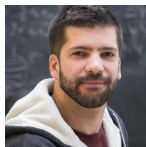
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- ▶ Find out more: [github.com/GTorlai/PastaQ.jl](https://github.com/GTorlai/PastaQ.jl)





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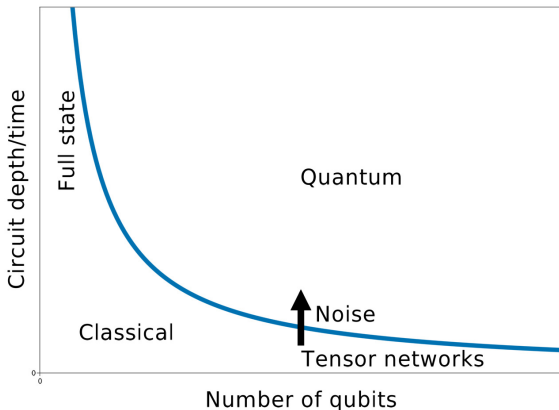
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  - ▶ This is not the focus of ITensor and PastaQ at the moment, specialized libraries like [Yao.jl](#) may be faster.
- ▶ Tensor networks are a common, general language for reasoning about quantum many-body systems (for example, quantum circuits).

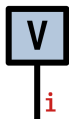
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- ▶ In my opinion, tensor networks are the best general purpose tool we have right now for studying quantum many-body systems (while we wait for a general purpose quantum computer).

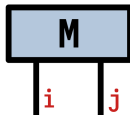


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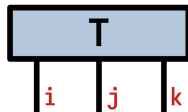
Vector  $V_i$   
Order-1 tensor



Matrix  $M_{ij}$   
Order-2 tensor



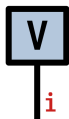
Tensor  $T_{ijk}$   
Order-3 tensor





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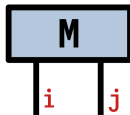
Vector  $V_i$   
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Vector inner  
product



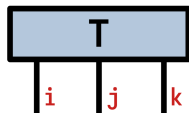
Matrix  $M_{ij}$   
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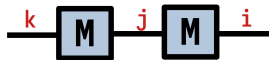
Matrix-vector  
multiplication



Tensor  $T_{ijk}$   
Order-3 tensor

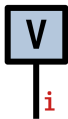


Matrix-matrix  
multiplication

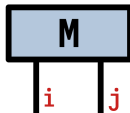


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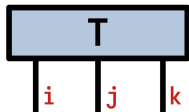
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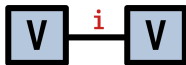
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Order-2 tensor



Tensor  $T_{ijk}$   
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Vector inner  
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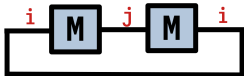
Matrix-vector  
multiplication



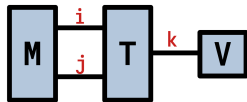
Matrix-matrix  
multiplication



Trace of matrix-matrix  
multiplication



General tensor  
network



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2. Launch Julia:

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3. Type the following commands:

```
1 julia> using Pkg
2
3 julia> Pkg.add("ITensors")
4 [...]
5
6 julia> Pkg.add("PastaQ")
7 [...]
```

## *How do I install ITensor/PastaQ?*

Now you can use ITensors and PastaQ:

```
1 julia> using ITensors
2
3 julia> i = Index(2);
4
5 julia> A = ITensor(i);
6
7 julia> using PastaQ
8
9 julia> gates = [("X", 1), ("CX", (1, 3))];
10
11 julia> = runcircuit(gates);
```

## *Tutorial: One-site state basics*

```
1  using ITensors
2
3  i = Index(2)
4
5
```

Load ITensor

2-dimensional labeled  
Hilbert space  
(dim=2|id=510)

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## Tutorial: One-site state basics

```
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```
1 Zp = ITensor(i)
2 Zp[i=>1] = 1
3
4 Zp = ITensor([1, 0], i)
```

i

$$|Z+\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Construct from a Vector

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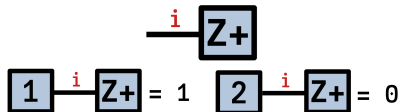
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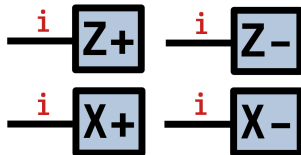
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2 Zm = ITensor([0, 1], i)
3 Xp = ITensor([1, 1]/√2, i)
4 Xm = ITensor([1, -1]/√2, i)
```

$$|Z+\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, |Z-\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
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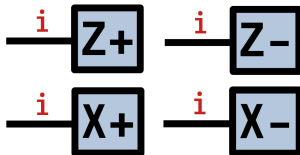
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```

```
1 (Zp + Zm)/√2
2 dag(Zp) * Xp
3 (dag(Zp) * Xp)[]
4 inner(Zp, Xp)
5 norm(Xp)
```

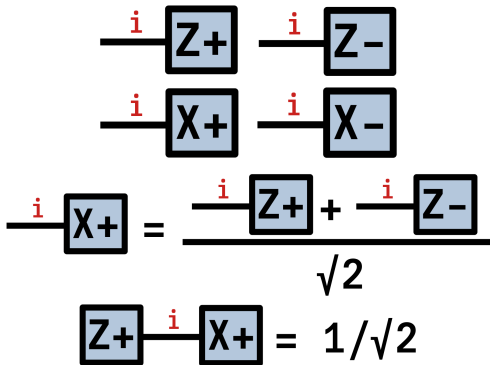


$\approx Xp$   
 $\approx \text{ITensor}(1/\sqrt{2})$   
 $\approx 1/\sqrt{2}$   
 $\approx 1/\sqrt{2}$   
 $\approx 1$

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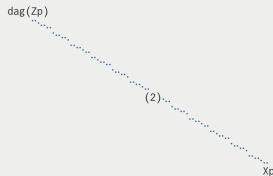
```
1 (Zp + Zm)/√2
2 dag(Zp) * Xp
3 (dag(Zp) * Xp)[]
4 inner(Zp, Xp)
5 norm(Xp)
```



# *Tutorial: One-site state basics*

```
1 using
    ITensorUnicodePlots
2
3 @visualize dag(Zp) * Xp
```

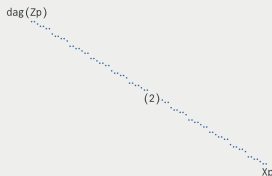
```
julia> @visualize dag(Zp) * Xp
```



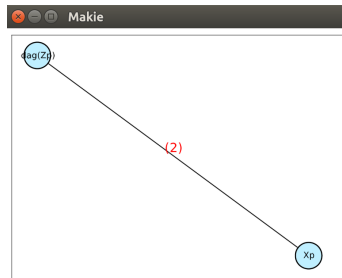
# Tutorial: One-site state basics

```
1 using
   ITensorUnicodePlots
2
3 @visualize dag(Zp) * Xp
```

```
julia> @visualize dag(Zp) * Xp
```



```
1 using ITensorGLMakie
2
3 @visualize dag(Zp) * Xp
```





## *Tutorial: One-site state basics*

```
1 i = Index(2, "S=1/2")
```

```
2
```

```
3
```

```
4
```

```
5
```

```
6
```

```
# "S=1/2" defines an
```

```
# operator basis
```

```
#
```

```
# Additionally:
```

```
# "Qubit", "Qudit",
```

```
# "Electron", ...
```

## Tutorial: One-site state basics

```
1 i = Index(2, "S=1/2")
2
3
4
5
6
```

```
# "S=1/2" defines an
# operator basis
#
# Additionally:
# "Qubit", "Qudit",
# "Electron", ...
```

```
1 Zp = state("Z+", i)
2 Zm = state("Z-", i)
3 Xp = state("X+", i)
4 Xm = state("X-", i)
```

```
# ITensor([1 0], i)
# ITensor([0 1], i)
# (Zp + Zm)/√2
# (Zp - Zm)/√2
```

## *Tutorial: Custom one-site states*

```
1 import ITensors: state
2
3
4 function state(
5     ::StateName "iX-",
6     ::SiteType "S=1/2"
7 )
8     return [im -im]/√2
9 end
```

```
# Overload ITensors.jl
# behavior

# Define a state with the
# name "iX-"
```

## Tutorial: Custom one-site states

```
1 import ITensors: state
2
3
4 function state(
5     ::StateName"iX-",
6     ::SiteType"S=1/2"
7 )
8     return [im -im]/√2
9 end
```

```
# Overload ITensors.jl
# behavior

# Define a state with the
# name "iX-"
```

```
1 iXm = state("iX-", i)
2
3 inner(Zp, iXm)
4 inner(Zm, iXm)
```

```
#  $\approx \text{im} * X_m$ 

#  $\approx \text{im}/\sqrt{2}$ 
#  $\approx -\text{im}/\sqrt{2}$ 
```

## *Tutorial: Priming*

```
1 i = Index(2)
2 j = Index(2)
3
4 i == j
```

```
# (dim=2|id=837)
# (dim=2|id=899)

# false
```

## Tutorial: Priming

```
1 i = Index(2)
2 j = Index(2)
3
4 i == j
```

```
# (dim=2|id=837)
# (dim=2|id=899)

# false
```

```
1 i
2
3 prime(i)
4 i'
5 i == i'
6 noprime(i')
```

```
# (dim=2|id=837)

# (dim=2|id=837)'
# (dim=2|id=837)'
# false
# (dim=2|id=837)
```

## *Tutorial: One-site operators*

```
1 Z = ITensor(i', i)
2 Z[i'=>1, i=>1] = 1
3 Z[i'=>2, i=>2] = -1
```

```
# TODO: Diagram
# Set elements
```

## Tutorial: One-site operators

```
1 Z = ITensor(i', i)
2 Z[i'=>1, i=>1] = 1
3 Z[i'=>2, i=>2] = -1
```

```
# TODO: Diagram
# Set elements
```

```
1 z = [
2     1 0
3     0 -1
4 ]
5
6 Z = ITensor(z, i', dag(i))
7
8 Z = op("Z", i)
```

```
# Matrix representation
# of Z
```

```
# Convert to ITensor
```

```
# Use predefined definition
```



## *Tutorial: One-site operators*

```
1  Z = op("Z", i)
2  X = op("X", i)
3
4  Zp = state("Z+", i)
5  Zm = state("Z-", i)
```

```
# Z
# X

# |Z+⟩
# |Z-⟩
```

## Tutorial: One-site operators

```
1 Z = op("Z", i)
2 X = op("X", i)
3
4 Zp = state("Z+", i)
5 Zm = state("Z-", i)
```

```
# Z
# X

# |Z+⟩
# |Z-⟩
```

```
1 XZp = X * Zp
2 XZp == Zm
3 XZp == Zm'
4 noprime(XZp) == Zm
```

```
# X|Z+⟩ = |Z-⟩
# false
# true
# true
```

## *Tutorial: One-site operators*

[TODO: Add visualization of inner product]

```
1 XZp = X * Zp
2
3 inner(Zm, XZp)
4
5 inner(Zm', XZp)
6 inner(Zp', XZp)
```

#  $X|Z+\rangle = |Z-\rangle$

# error: not a scalar value

#  $\approx 1$

#  $\approx 0$

## Tutorial: One-site operators

[TODO: Add visualization of inner product]

```
1 XZp = X * Zp
2
3 inner(Zm, XZp)
4
5 inner(Zm', XZp)
6 inner(Zp', XZp)
```

#  $X|Z+\rangle = |Z-\rangle$

# error: not a scalar value

#  $\approx 1$

#  $\approx 0$

```
1 (dag(Zm)' * X * Zp)[]
2 inner(Zm', X, Zp)
3
4 XZp = apply(X, Zp)
5 inner(Zm, XZp)
```

#  $\approx 1$

#  $\approx 1$

# = `noprime(X * Zp)`

#  $\approx 1$

## *Tutorial: Custom one-site operators*

```
1  import ITensors: op
2
3  function op(
4      ::OpName"iX",
5      ::SiteType"S=1/2"
6  )
7      return [
8          0 im
9          im 0
10     ]
11  end
```

```
# Overload ITensors.jl
# behavior
```

## Tutorial: Custom one-site operators

```
1 import ITensors: op
2
3 function op(
4     ::OpName "iX",
5     ::SiteType "S=1/2"
6 )
7     return [
8         0 im
9         im 0
10    ]
11 end
```

```
# Overload ITensors.jl
# behavior
```

```
1 op("iX", i)
```

```
# im * X
```

## *Tutorial: Two-site states*

```
1 i1 = Index(2, "S=1/2")
2 i2 = Index(2, "S=1/2")
3
4 i1 == i2
5
6 ZpZm = ITensor(i1, i2)
7 ZpZm[i1=>1, i2=>2] = 1
```

```
# (dim=2|id=505|"S=1/2")
# (dim=2|id=576|"S=1/2")

# false

#  $|Z+\rangle_1 |Z-\rangle_2 = |Z+Z-\rangle$ 
```

## Tutorial: Two-site states

```
1 i1 = Index(2, "S=1/2")
2 i2 = Index(2, "S=1/2")
3
4 i1 == i2
5
6 ZpZm = ITensor(i1, i2)
7 ZpZm[i1=>1, i2=>2] = 1
```

```
# (dim=2|id=505|"S=1/2")
# (dim=2|id=576|"S=1/2")

# false

#  $|Z+\rangle_1|Z-\rangle_2 = |Z+Z-\rangle$ 
```

```
1 Zp1 = state("Z+", i1)
2 Zp2 = state("Z+", i2)
3
4 Zm1 = state("Z-", i1)
5 Zm2 = state("Z-", i2)
6
7 ZpZm = Zp1 * Zm2
8 ZmZp = Zm1 * Zp2
```

```
#  $|Z+\rangle_1$ 
#  $|Z+\rangle_2$ 

#  $|Z-\rangle_1$ 
#  $|Z-\rangle_2$ 

#  $|Z+Z-\rangle = |Z+\rangle_1|Z-\rangle_2$ 
#  $|Z-Z+\rangle = |Z-\rangle_1|Z+\rangle_2$ 
```



## *Tutorial: Two-site states*

[TODO: Add visualization of  $\text{inner}(\text{Cat}, \text{Cat})$ , SVD]

```
1 Cat = ITensor(i1, i2)
2 Cat[i1=>1, i2=>2] = 1/√2
3 Cat[i1=>2, i2=>1] = 1/√2
4
5 Cat = (Zp1 * Zm2 +
6        Zm1 * Zp2)/√2
```

#  $(|Z+\rangle|Z-\rangle + |Z-\rangle|Z+\rangle)/\sqrt{2}$

# From single-site states

## Tutorial: Two-site states

[TODO: Add visualization of inner(Cat, Cat), SVD]

```
1 Cat = ITensor(i1, i2)
2 Cat[i1=>1, i2=>2] = 1/√2
3 Cat[i1=>2, i2=>1] = 1/√2
4
5 Cat = (Zp1 * Zm2 +
6        Zm1 * Zp2)/√2
```

#  $(|Z+\rangle|Z-\rangle + |Z-\rangle|Z+\rangle)/\sqrt{2}$

# From single-site states

```
1 inner(Cat, Cat)
2 inner(ZpZm, Cat)
3
4 U, S, V = svd(ZmZp, i1)
5 s = diag(S)
6
7 U, S, V = svd(Cat, i1)
8 s = diag(S)
```

#  $\approx 1$

#  $\approx 1/\sqrt{2}$

#  $\approx [1, 0]$

#  $\approx [1/\sqrt{2}, 1/\sqrt{2}]$

## Tutorial: Two-site operators

[TODO: Add visualization of H]

```
1 H = ITensor(i1', i2', i1, i2)
2 H[i1'=>2, i2'=>1,
3   i1=>2, i2=>1] = -1
4 # ...
```

```
# Make a Hamiltonian:
# Transverse field Ising
# n=2 sites
#  $H = -\sum_j^{n-1} Z_j Z_{j+1} + h \sum_j^n X_j$ 
```

## Tutorial: Two-site operators

[TODO: Add visualization of H]

```
1 H = ITensor(i1', i2', i1, i2)
2 H[i1'=>2, i2'=>1,
3   i1=>2, i2=>1] = -1
4 # ...
```

# Make a Hamiltonian:

# Transverse field Ising

# n=2 sites

#  $H = -\sum_j^{n-1} Z_j Z_{j+1} + h \sum_j^n X_j$

```
1 Id1 = op("Id", i1)
2 Z1 = op("Z", i1)
3 X1 = op("X", i1)
4 # ...
5
6 ZZ = Z1 * Z2
7 XI = X1 * Id2
8 IX = Id1 * X2
9
10 h = 0.5
11 H = -ZZ + h * (XI + IX)
```

# Alternative:

# Build from single-site

# operators.

# Less error-prone.

## *Tutorial: Two-site operators*

[TODO: Add visualization of  $\langle H \rangle$ ]

```
1 ZpZp = Zp1 * Zp2
2
3
4
5 (dag(ZpZp)' * H * ZpZp)[]
6 inner(ZpZp', H, ZpZp)
7 inner(ZpZp, apply(H, ZpZp))
```

```
# Expectation value:
#  $\langle H \rangle = \langle Z+Z+ | H | Z+Z+ \rangle$ 
```

```
#  $\approx -1$ 
```

## Tutorial: Two-site operators

[TODO: Add visualization of  $\langle H \rangle$ ]

```
1 ZpZp = Zp1 * Zp2
2
3
4
5 (dag(ZpZp)' * H * ZpZp)[]
6 inner(ZpZp', H, ZpZp)
7 inner(ZpZp, apply(H, ZpZp))
```

# Expectation value:  
#  $\langle H \rangle = \langle Z+Z+ | H | Z+Z+ \rangle$

#  $\approx -1$

```
1 D, U = eigen(H)
2 diag(D)
```

#  $\approx [-\sqrt{2}, -1, 1, \sqrt{2}]$

## Tutorial: Custom two-site operators

```
1 import ITensors: op
2
3 function op(
4     ::OpName"CRy",
5     ::SiteType"S=1/2";
6      $\theta$ 
7 )
8     c = cos( $\theta/2$ )
9     s = sin( $\theta/2$ )
10    return [
11        1 0 0 0
12        0 1 0 0
13        0 0 c -s
14        0 0 s  c
15    ]
16 end
```

```
# Controlled-Ry (CRy)
# rotation gate

# CRy( $\theta$ )
```

## *Tutorial: Custom two-site operators*

[TODO: Add visualization of CH|ZpZm>]

```
1 CH = op("CRy", i1, i2;  
2        $\theta=\pi/2$ )
```

```
# Controlled-Hadamard gate  
# CH = CRy( $\theta=\pi/2$ )
```



## Tutorial: Custom two-site operators

[TODO: Add visualization of  $\text{CH}|\text{ZpZm}\rangle$ ]

```
1 CH = op("CRy", i1, i2;  
2          $\theta=\pi/2$ )
```

```
# Controlled-Hadamard gate  
# CH = CRy( $\theta=\pi/2$ )
```

```
1 ZpZm = Zp1 * Zm2  
2  
3 CH_Xm = apply(CH, ZpZm)  
4  
5 CH_Xm  $\approx$  Zp1 * Xm2
```

```
#  $|Z+Z-\rangle = |Z+\rangle_1 |Z-\rangle_2$   
#  $\text{CH}|Z+Z-\rangle = |Z+X-\rangle$   
# true
```

## *Tutorial: Two-site state optimization*

[TODO: Add visualization of minimizing  $\langle \psi | H | \psi \rangle$ ]

```
1 function E( $\psi$ )
2    $\psi H \psi = \text{inner}(\psi', H, \psi)$ 
3    $\psi \psi = \text{inner}(\psi, \psi)$ 
4   return  $\psi H \psi / \psi \psi$ 
5 end
```

```
# Function to minimize:
# Expectation value of the
# energy.
#
#  $E(\psi) = \langle \psi | H | \psi \rangle / \langle \psi | \psi \rangle$ 
```

## Tutorial: Two-site state optimization

[TODO: Add visualization of minimizing  $\langle \mathbf{v} | \mathbf{H} | \mathbf{v} \rangle$ ]

```
1 function E( $\psi$ )
2    $\psi H \psi = \text{inner}(\psi', H, \psi)$ 
3    $\psi \psi = \text{inner}(\psi, \psi)$ 
4   return  $\psi H \psi / \psi \psi$ 
5 end
```

```
# Function to minimize:
# Expectation value of the
# energy.
#
#  $E(\psi) = \langle \psi | H | \psi \rangle / \langle \psi | \psi \rangle$ 
```

```
1 function minimize(f,  $\partial f$ , x;
2   nsteps,  $\gamma$ )
3   for n in 1:nsteps
4      $x = x - \gamma * \partial f(x)$ 
5   end
6   return x
7 end
```

```
# Simple gradient descent.
# Must provide function  $f(x)$ 
# to minimize and  $\partial f(x)$ ,
# the gradient of  $f$  at  $x$ .

#  $\gamma$  is the gradient
# descent step size.
```

## Tutorial: Two-site state optimization

[TODO: Add visualization of minimizing  $\langle \mathbf{v} | \mathbf{H} | \mathbf{v} \rangle$ ]

```
1   $\psi_0 = (\text{Zp1} * \text{Zm2} +$   
2       $\text{Zm1} * \text{Zp2})/\sqrt{2}$   
3  
4   $E(\psi_0)$   
5  
6  using Zygote: gradient  
7   $\partial E(\psi) = \text{gradient}(E, \psi)[1]$   
8  
9   $\text{norm}(\partial E(\psi_0))$ 
```

```
#  $|\psi_0\rangle = (|Z+Z+\rangle +$   
#       $|Z-Z-\rangle)/\sqrt{2}$   
  
#  $\approx -1$   
  
# Using Zygote for automatic  
# differentiation of the energy.  
  
#  $\approx 2$ 
```

## Tutorial: Two-site state optimization

[TODO: Add visualization of minimizing  $\langle \mathbf{v} | \mathbf{H} | \mathbf{v} \rangle$ ]

```
1   $\psi_0 = (\mathbf{Z}_{p1} * \mathbf{Z}_{m2} +$   
2       $\mathbf{Z}_{m1} * \mathbf{Z}_{p2}) / \sqrt{2}$   
3  
4   $\mathbf{E}(\psi_0)$   
5  
6  using Zygote: gradient  
7   $\partial \mathbf{E}(\psi) = \text{gradient}(\mathbf{E}, \psi)[1]$   
8  
9   $\text{norm}(\partial \mathbf{E}(\psi_0))$ 
```

```
#  $|\psi_0\rangle = (|Z+Z+\rangle +$   
#       $|Z-Z-\rangle) / \sqrt{2}$ 
```

```
#  $\approx -1$ 
```

```
# Using Zygote for automatic  
# differentiation of the energy.
```

```
#  $\approx 2$ 
```

```
1   $\psi = \text{minimize}(\mathbf{E}, \partial \mathbf{E}, \psi_0;$   
2       $\text{nsteps}=10, \gamma=0.1)$   
3  
4   $\mathbf{E}(\psi_0), \text{norm}(\partial \mathbf{E}(\psi_0))$   
5   $\mathbf{E}(\psi), \text{norm}(\partial \mathbf{E}(\psi))$   
6
```

```
# Minimize over  $\psi$ :  
#  $\mathbf{E}(\psi) = \langle \psi | \mathbf{H} | \psi \rangle / \langle \psi | \psi \rangle$ 
```

```
#  $(-1, 2)$ 
```

```
#  $(-1.4142131, 0.0010865277)$ 
```

```
#  $\approx (-\sqrt{2}, 0)$ 
```

## Tutorial: Two-site circuit optimization

[TODO: Add visualization of U]

```
1 # Circuit as a vector of gates:
2 U( $\theta$ , i1, i2) = [
3     op("Ry", i1;  $\theta$ = $\theta$ [1]),
4     op("Ry", i2;  $\theta$ = $\theta$ [2]),
5     op("CX", i1, i2),
6     op("Ry", i1;  $\theta$ = $\theta$ [3]),
7     op("Ry", i2;  $\theta$ = $\theta$ [4]),
8 ]
9
10
11
```

```
# PastaQ notation:
u( $\theta$ , j1, j2) = [
    ("Ry", j1, (;  $\theta$ = $\theta$ [1])),
    ("Ry", j2, (;  $\theta$ = $\theta$ [2])),
    ("CNOT", j1, j2),
    ("Ry", j1, (;  $\theta$ = $\theta$ [3])),
    ("Ry", j2, (;  $\theta$ = $\theta$ [4])),
]
U( $\theta$ , i1, i2) =
    buildcircuit(u( $\theta$ , 1, 2), [i1, i2])
```

## *Tutorial: Two-site circuit optimization*

[TODO: Add visualization of minimizing  $\langle 0 | U H U | 0 \rangle$ ]

```
1  $\psi_0 = Z_{p1} * Z_{p2}$ 
2
3
4 function E( $\theta$ )
5      $\psi_\theta = \text{apply}(U(\theta), \psi_0)$ 
6     return inner( $\psi_\theta$ , H,  $\psi_\theta$ )
7 end
```

```
# References state:
#  $|0\rangle = |Z+Z+\rangle$ 

# Find  $\theta$  that minimizes:
#  $E(\theta) = \langle 0 | U(\theta)^\dagger H U(\theta) | 0 \rangle$ 
#            $= \langle \theta | H | \theta \rangle$ 
```

## Tutorial: Two-site circuit optimization

[TODO: Add visualization of minimizing  $\langle 0 | U H U | 0 \rangle$ ]

```
1  $\psi_0 = Z_{p1} * Z_{p2}$ 
2
3
4 function E( $\theta$ )
5      $\psi_\theta = \text{apply}(U(\theta), \psi_0)$ 
6     return inner( $\psi_\theta$ , H,  $\psi_\theta$ )
7 end
```

```
# References state:
#  $|0\rangle = |Z+Z+\rangle$ 

# Find  $\theta$  that minimizes:
#  $E(\theta) = \langle 0 | U(\theta)^\dagger H U(\theta) | 0 \rangle$ 
#            $= \langle \theta | H | \theta \rangle$ 
```

```
1  $\theta_0 = [0, 0, 0, 0]$ 
2  $\theta = \text{minimize}(E, \partial E, \theta_0;$ 
3     nsteps=40,  $\gamma=0.5$ )
4
5  $E(\theta_0), \text{norm}(\partial E(\theta_0))$ 
6  $E(\theta), \text{norm}(\partial E(\theta))$ 
7
```

```
#  $(-1, \sqrt{3}/2)$ 
#  $(-1.4142077, 0.0017584116)$ 
#  $\approx (-\sqrt{2}, 0)$ 
```



## Tutorial: Two-site fidelity optimization

[TODO: Add visualization of minimizing  $\langle \psi | U | \psi_0 \rangle$ ]

```
1   $\psi_0 = Z_{p1} * Z_{p2}$ 
2   $\psi = (Z_p Z_p + Z_m Z_m) / \sqrt{2}$ 
3
4
5
6  function F( $\theta$ )
7     $\psi_\theta = \text{apply}(U(\theta, i1, i2), \psi_0)$ 
8    return -abs(inner( $\psi, \psi_\theta$ ))^2
9  end
```

```
# Reference state:
#  $|0\rangle = |Z+Z+\rangle$ 

# Target state:
#  $|\psi\rangle = (|Z+Z+\rangle + |Z-Z-\rangle) / \sqrt{2}$ 

# Find  $\theta$  that minimizes:
#  $F(\theta) = -|\langle \psi | U(\theta) | 0 \rangle|^2$ 
```

## Tutorial: Two-site fidelity optimization

[TODO: Add visualization of minimizing  $\langle \psi | U | \psi_0 \rangle$ ]

```
1   $\psi_0 = Z_p1 * Z_p2$ 
2   $\psi = (Z_p Z_p + Z_m Z_m) / \sqrt{2}$ 
3
4
5
6  function F( $\theta$ )
7       $\psi_\theta = \text{apply}(U(\theta, i1, i2), \psi_0)$ 
8      return -abs(inner( $\psi$ ,  $\psi_\theta$ ))2
9  end
```

```
# Reference state:
#  $|0\rangle = |Z+Z+\rangle$ 

# Target state:
#  $|\psi\rangle = (|Z+Z+\rangle +$ 
#            $|Z-Z-\rangle)/\sqrt{2}$ 

# Find  $\theta$  that minimizes:
#  $F(\theta) = -|\langle \psi | U(\theta) | 0 \rangle|^2$ 
```

```
1   $\theta_0 = [0, 0, 0, 0]$ 
2   $\theta = \text{minimize}(F, \partial F, \theta_0;$ 
3      nsteps=50,  $\gamma=0.1)$ 
4
5   $F(\theta_0), \text{norm}(\partial F(\theta_0))$ 
6   $F(\theta), \text{norm}(\partial F(\theta))$ 
```

```
#  $\approx (-0.5, 0.5)$ 
#  $\approx (-0.9938992, 0.07786879)$ 
```

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of  $\langle Z_p | \text{Cat} \rangle$ ]

```
1  n = 30
2  i = [Index(2, "S=1/2")
3      for j in 1:n]
4
5  Zp = MPS(i, "Z+")
6  Zm = MPS(i, "Z-")
7
8  Cat = (Zp + Zm)/√2
9
```

#  $n$ -site state

#  $|Z+Z+\dots Z+\rangle$

#  $|Z-Z-\dots Z-\rangle$

#  $(|Z+Z+\dots Z+\rangle +$

#  $|Z-Z-\dots Z-\rangle)/\sqrt{2}$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of  $\langle Z_p | \text{Cat} \rangle$ ]

```
1  n = 30
2  i = [Index(2, "S=1/2")
3      for j in 1:n]
4
5  Zp = MPS(i, "Z+")
6  Zm = MPS(i, "Z-")
7
8  Cat = (Zp + Zm)/√2
9
```

#  $n$ -site state

#  $|Z+Z+\dots Z+\rangle$

#  $|Z-Z-\dots Z-\rangle$

#  $(|Z+Z+\dots Z+\rangle +$

#  $|Z-Z-\dots Z-\rangle)/\sqrt{2}$

```
1  maxlinkdim(Zp)
2  maxlinkdim(Cat)
3  inner(Zp, Zp)
4  inner(Zm, Zp)
5  norm(Cat)
6
```

# 1 (product state)

# 2 (entangled state)

#  $\langle Z+ | Z+ \rangle \approx 1$

#  $\langle Z- | Z+ \rangle \approx 0$

#  $(\langle Z+ | + \langle Z- |)(|Z+\rangle + |Z+\rangle)/2$

#  $\approx 1$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of minimizing  $\langle \text{psi} | H | \text{psi} \rangle$ ]

```
1  j = n ÷ 2
2  Xj = op("X", i[j])
3
4  XjZp = apply(Xj, Zp)
5
6
7  state = [k == j ? "Z-" : "Z+"
8           for k in 1:n]
9  XjZp = MPS(i, state)
```

```
# j = n ÷ 2
# Xj

# Xj|Z+Z+...Z+⟩ =
#   |Z+Z+...Z-...Z+⟩

# |Z+Z+...Z-...Z+⟩
```

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of minimizing  $\langle \psi | H | \psi \rangle$ ]

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```

```
# j = n ÷ 2
# Xj
# Xj|Z+Z+...Z+⟩ =
#   |Z+Z+...Z-...Z+⟩
# |Z+Z+...Z-...Z+⟩
```

```
1  maxlinkdim(XjZp)
2  inner(Zp, XjZp)
3  inner(XjZp, apply(Xj, Zp))
```

```
# 1 (product state)
# ≈ 0
# ≈ 1
```

## *Tutorial: n-site states with MPS*

[TODO: Add visualization of H]

```
1 function ising(n; h)
2   H = OpSum()
3   for j in 1:(n - 1)
4     H -= "Z", j, "Z", j + 1
5   end
6   for j in 1:n
7     H += h, "X", j
8   end
9   return H
10 end
```

# n sites

$$\# H = -\sum_j^{n-1} Z_j Z_{j+1} + h \sum_j^n X_j$$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

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9   return H
10 end
```

# n sites

$$\# H = -\sum_j^{n-1} Z_j Z_{j+1} + h \sum_j^n X_j$$

```
1 h = 0.5
2 H = MPO(ising(n; h=h), i)
3 maxlinkdim(H)
4 Zp = MPS(i, "Z+")
5 inner(Zp', H, Zp)
```

# = 3 (local Hamiltonian)

#  $\langle Z_+ Z_+ \dots Z_+ | H | Z_+ Z_+ \dots Z_+ \rangle$



## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1   $\psi_0 = \text{MPS}(i, \text{"Z+"})$ 
2
3   $\psi = \text{minimize}(\text{E}, \partial\text{E}, \psi_0;$ 
4       $\text{nsteps}=50, \gamma=0.1,$ 
5       $\text{maxdim}=10, \text{cutoff}=1\text{e-}5)$ 
6
7   $\text{E}_{dmrg}, \psi_{dmrg} = \text{dmrg}(\text{H}, \psi_0;$ 
8       $\text{nsweeps}=10,$ 
9       $\text{maxdim}=10, \text{cutoff}=1\text{e-}5)$ 
```

#  $|0\rangle = |Z+Z+\dots Z+\rangle$

# Minimize over  $\psi$ :  
#  $\langle\psi|\text{H}|\psi\rangle/\langle\psi|\psi\rangle$

# DMRG solves:  
#  $\text{H}|\psi\rangle \approx |\psi\rangle$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

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# DMRG solves:

#  $\mathbf{H}|\psi\rangle \approx |\psi\rangle$

```
1   $\text{maxlinkdim}(\psi_0)$ 
2   $\text{maxlinkdim}(\psi)$ 
3   $\text{maxlinkdim}(\psi_{\text{dmrg}})$ 
4   $\mathbf{E}(\psi_0), \text{norm}(\partial(\psi_0))$ 
5   $\mathbf{E}(\psi), \text{norm}(\partial\mathbf{E}(\psi))$ 
6   $\mathbf{E}(\psi_{\text{dmrg}}), \text{norm}(\partial\mathbf{E}(\psi_{\text{dmrg}}))$ 
```

# = 1 (product state)

# = 3 (entangled state)

# = 2 (entangled state)

#  $\approx (-29, 5.4772256)$

#  $\approx (-31.0317917, 0.06673780)$

#  $\approx (-31.0356110, 0.02413237)$

## *Tutorial: $n$ -site states with MPS*

[TODO: Add visualization of H]

```
1 Ry_layer( $\theta$ , i) = [op("Ry", i[j];  $\theta=\theta[j]$ ) for j in 1:n]
2 CX_layer(i) = [op("CX", i[j], i[j+1]) for j in 1:2:(n-1)]
```

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1 Ry_layer( $\theta$ , i) = [op("Ry", i[j];  $\theta=\theta[j]$ ) for j in 1:n]
2 CX_layer(i) = [op("CX", i[j], i[j+1]) for j in 1:2:(n-1)]
```

```
1 function U( $\theta$ , i; nlayers)
2   n = length(i)
3    $U_\theta$  = Ry_layer( $\theta[1:n]$ , i)
4   for l in 1:(nlayers - 1)
5      $\theta_l$  =  $\theta[(1:n) .+ l * n]$ 
6      $U_\theta$  = [ $U_\theta$ ; CX_layer(i)]
7      $U_\theta$  = [ $U_\theta$ ; Ry_layer( $\theta_l$ , i)]
8   end
9   return  $U_\theta$ 
10 end
```

```
#  $U(\theta) = \text{Ry}_1(\theta_1) \dots \text{Ry}_n(\theta_n)$ 
#            $\text{CX}_{1,2} \dots \text{CX}_{n-1,n}$ 
#            $\text{Ry}_1(\theta_{n+1}) \dots \text{Ry}_n(\theta_{2n})$ 
#           ...
```

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1  $\psi_0 = \text{MPS}(i, \text{"Z+"})$ 
2  $\text{nlayers} = 6$ 
3 function  $E(\theta)$ 
4    $\psi_\theta = \text{apply}(U(\theta, i; \text{nlayers}), \psi_0)$ 
5   return  $\text{inner}(\psi_\theta', H, \psi_\theta)$ 
6 end
7
8  $\theta_0 = \text{zeros}(\text{nlayers} * n)$ 
9  $\theta = \text{minimize}(E, \partial E, \theta_0;$ 
10    $\text{nsteps}=20, \gamma=0.1)$ 
```

#  $|0\rangle = |Z+Z+\dots Z+\rangle$

# Minimize over  $\theta$ :

#  $|\theta\rangle = U(\theta)|0\rangle$

#  $E(\theta) = \langle \theta | H | \theta \rangle$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1  $\psi_0 = \text{MPS}(i, \text{"Z+"})$ 
2 nlayers = 6
3 function E( $\theta$ )
4      $\psi_\theta = \text{apply}(U(\theta, i; \text{nlayers}), \psi_0)$ 
5     return inner( $\psi_\theta'$ , H,  $\psi_\theta$ )
6 end
7
8  $\theta_0 = \text{zeros}(\text{nlayers} * n)$ 
9  $\theta = \text{minimize}(\text{E}, \partial\text{E}, \theta_0;$ 
10     nsteps=20,  $\gamma=0.1$ )
```

#  $|0\rangle = |Z+Z+\dots Z+\rangle$

# Minimize over  $\theta$ :

#  $|\theta\rangle = U(\theta)|0\rangle$

#  $E(\theta) = \langle\theta|H|\theta\rangle$

```
1 maxlinkdim( $\psi_0$ )
2 maxlinkdim( $\psi_\theta$ )
3 E( $\theta_0$ ), norm( $\partial\text{E}(\theta_0)$ )
4 E( $\theta$ ), norm( $\partial\text{E}(\theta)$ )
```

# 1 (product state)

# 2 (entangled state)

# (-29, 5.773335)

# (-31.017062, 0.000759)

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1  $\psi_0 = \text{MPS}(i, \text{"Z+"})$ 
2  $\text{nlayers} = 6$ 
3 function  $F(\theta)$ 
4    $\psi_\theta = \text{apply}(U(\theta, i; \text{nlayers}), \psi_0)$ 
5   return  $-\text{abs}(\text{inner}(\psi', \psi_\theta))^2$ 
6 end
7
8  $\theta_0 = \text{zeros}(\text{nlayers} * n)$ 
9  $\theta = \text{minimize}(F, \partial F, \theta_0;$ 
10    $\text{nsteps}=20, \gamma=0.1)$ 
```

#  $|0\rangle = |Z+Z+\dots Z+\rangle$

# Minimize over  $\theta$ :

#  $F(\theta) = -|\langle \psi | U(\theta) | 0 \rangle|^2$   
#  $\quad = -|\langle \psi | \theta \rangle|^2$

## Tutorial: $n$ -site states with MPS

[TODO: Add visualization of H]

```
1  $\psi_0 = \text{MPS}(i, \text{"Z+"})$ 
2 nlayers = 6
3 function F( $\theta$ )
4      $\psi_\theta = \text{apply}(U(\theta, i; \text{nlayers}), \psi_0)$ 
5     return -abs(inner( $\psi'$ ,  $\psi_\theta$ ))^2
6 end
7
8  $\theta_0 = \text{zeros}(\text{nlayers} * n)$ 
9  $\theta = \text{minimize}(\text{F}, \partial\text{F}, \theta_0;$ 
10     nsteps=20,  $\gamma=0.1$ )
```

#  $|0\rangle = |Z+Z+\dots Z+\rangle$

# Minimize over  $\theta$ :

#  $F(\theta) = -|\langle\psi|U(\theta)|0\rangle|^2$

#  $\quad\quad = -|\langle\psi|\theta\rangle|^2$

```
1 maxlinkdim( $\psi_0$ )
2 maxlinkdim( $\psi_\theta$ )
3 F( $\theta_0$ ), norm( $\partial\text{F}(\theta_0)$ )
4 F( $\theta$ ), norm( $\partial\text{F}(\theta)$ )
```

# 1 (product state)

# 2 (entangled state)

# (-0.556066, 0.895717)

# (-0.995230, 0.048939)



## *Future directions*

- ▶ More AD, make ITensor fully differentiable (have some work to do, like tensor decompositions and general network contractions, more MPS/MPO functions. You will find bugs!).

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- ▶ Many ongoing projects and directions: quantum chemistry (for example UCC), real space parallel DMRG, TDVP, and TEBD, MPO compression tools, general approximate contraction techniques for unstructured networks, contracting and optimizing general tensor networks with AD, infinite MPS and tensor network tools like VUMPS and TDVP, trying out different network topologies for noisy circuit tomography, simulation and optimization.

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