

# Cluster Expansion of Thermal States using Tensor Networks

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## Foreword

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## Abstract

Het abstract is maximum één bladzijde en bevat minstens: a) De informatie die werd vermeld op het titelblad (eigen vorm); b) Een summiere beschrijving van het werk (vijftien à twintig regels); c) Eventueel: drie tot vijf goed gekozen trefwoorden die het onderwerp best omschrijven.

## Extended Abstract

De extended abstract heeft een standaardlengte van minimaal twee bladzijden, met een maximum van zes bladzijden.

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# Chapter 1

## Introduction

There is nothing new to be discovered in physics now. All that remains is more and more precise measurement.

---

Lord Kelvin, 1900

### 1.1 Introduction

In 2015, there were about 5.6 million known physics papers in literature. At the current rate, this number doubles every 18.7 years [1]. Despite this enormous body of literature, there are a lot of things which are not completely understood. Some examples include a self-consistent theory of quantum gravity, the need for dark energy and matter in cosmology, the arrow of time, the matter-antimatter asymmetry. There even is no interpretation of quantum mechanics where everyone agrees upon.

But certainly not all open problems have to do with 'new' physics. In many areas of physics, computing the implications of relatively simple laws becomes exceedingly difficult for many particles. Of historical importance is the three-body problem, describing the trajectory of 3 gravitational bodies such as the earth, moon and sun. The general case is not solved, despite developments over the last 300 years.

In reality, the real challenge is to model the macroscopic properties of quantum many-body system with around  $10^{23}$  particles. Needless to say, this is not an easy task at all. Finding good and computable approximations is of primary importance in the fields of quantum chemistry, condensed matter physics, and materials science.

In computational chemistry, the many-body problem is tackled with methods which fall in one of the following categories: (post-) Hartree-Fock methods, density functional theory (DFT) and force-field methods. While they have many

vind bron en voorbeelden

applications, these methods are not fully able to capture all the properties of the so called strongly correlated matter.

Examples of phase of strongly correlated matter which are not yet understood include high-T superconductors, topological ordered phases, quantum spin liquids [2]. There exist different methods to investigate these exciting materials. A very limited number of models is quantum integrable, meaning they can be solved in a non perturbative way. Also, some properties of models near criticality can be determined exactly with conformal field theory (CFT). But for some systems, we can only simulate the behaviour with numerical techniques. To make progress, new fast and accurate numerical methods are needed, because exact diagonalisation becomes unfeasible for large systems.

Some examples of such numerical techniques, which will not be discussed, are: Dynamical Mean Field (DMFT) / Dynamical Cluster Approximation (DCA), Series expansion, Density Matrix Embedding Theory (DMET), Fixed-node Monte Carlo, Diagrammatic Monte Carlo, Variational Monte Carlo, Functional renormalization group (FRG) and Coupled-cluster methods. [3]

In this thesis, a technique is proposed that builds on the broad field of tensor networks.

## 1.2 Tensor networks

This is often referred to as the curse of dimensionality. The size of the Hilbert space of quantum states grows exponentially fast. This prevents an efficient description of all possible quantum states.

area law+picture

sign problem monte carlo,

...

write about tensor networks

## Chapter 2

# Tensor networks



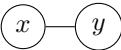
The aim of this chapter is to give a basic introduction to tensor networks from a computational viewpoint. First the graphical notation for tensors, which is ubiquitous in the field, will be introduced and explained. A incomplete classification of the different kinds of tensor networks will be discussed. Some routine tensor network manipulations are explained. Also, a selection of different algorithms, which are from a computational point of view the reason behind the success of tensor networks, are presented very briefly.

### 2.1 Tensor networks

#### 2.1.1 Graphical notation

Before explaining tensor networks, some graphical notation should be introduced. This really is a way to conveniently write vectors, matrices and in general tensors without the need to introduce many labels. A tensor  $T$  is represented by a circle with a number of external legs, according to the number of external indices. Connected legs are summed. Some examples are shown in table 2.1. Every leg which is connected to multiple tensors, is contracted.

Table 2.1: Caption

conventional	Einstein	tensor notation
$\vec{x}$	$x_\alpha$	
$M$	$M_{\alpha\beta}$	
$\vec{x} \cdot \vec{y}$	$x_\alpha y_\alpha$	

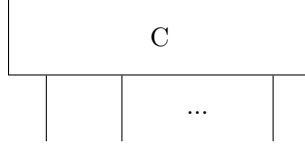


Figure 2.1: Caption

### 2.1.2 Representing a quantum state

Tensor network come in many shapes and forms. Tensor networks are really used to represent a tensor with many legs. A general quantum state with  $N$  sites can be described in a given basis  $|i\rangle$  in the following way:

$$|\Psi\rangle = \sum_{i_1 i_2 \dots i_n} C^{i_1 i_2 \dots i_n} |i_1\rangle \otimes |i_2\rangle \otimes \dots \otimes |i_n\rangle \quad (2.1)$$

Here the tensor  $C$  holds all the information of the quantum state. The graphical representation can be seen in fig. 2.1.

This requires an exponential number  $d^n$  of coefficients  $C$  where  $d$  is the dimensions of basis  $|i\rangle$ . In order to make the problem tractable, the following form is proposed as wave function:

$$C^{i_1 i_2 \dots i_n} = C^{1 i_1}_{\alpha_1} C^{2 i_2}_{\alpha_1 \alpha_2} \dots C^{n i_n}_{\alpha_{n-1}} \quad (2.2)$$

Where summation over shared indices is implied. It is always possible to find such a representation by means of matrix decomposition (see section 2.2). The summation over  $\alpha_i$  are called a virtual bond and their dimension is denoted by  $\chi$ . At this point, this is not yet an improvement because the bond dimension needs to be exponentially large in order to represent the tensor  $C$  exactly.

Explicit translational invariance is given by tensor  $C_{\alpha\beta}^i$  that don't depend on the location. The chain is closed by a matrix  $M$  which contains the boundary conditions. Setting  $\alpha_n = \alpha_0$ . We can now write this as a Trace over matrix products:

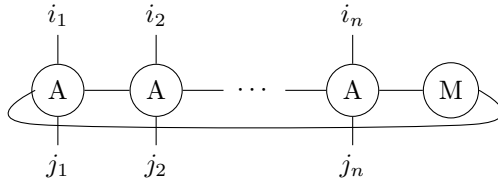
$$|\Psi\rangle = \text{Tr}(C^{i_1} C^{i_2} \dots C^{i_n} M) |i_1\rangle \otimes |i_2\rangle \otimes \dots \otimes |i_n\rangle \quad (2.4)$$

### 2.1.3 Classification

#### 2.1.3.1 MPS

**2.1.3.1.1 MPO** Similarly, a Matrix Product Operator (MPO) is of the following form:

$$\hat{O} = \sum \text{Tr}(A^{i_1 j_1} A^{i_2 j_2} \dots A^{i_n j_n} M) \times |i_1\rangle \langle j_1| \otimes |i_2\rangle \langle j_2| \otimes \dots \otimes |i_n\rangle \langle j_n| \quad (2.6)$$



The matrix M contains the boundary conditions of the operator. Many Hamiltonians can be represented by an MPO. For ins

**2.1.3.1.2 PEPS** Exact contraction is hashtag P-Hard:

No exact canonical form

#### 2.1.3.1.3 PEPO

**2.1.3.1.4 Others** MERA, TTN,

ather tns

## 2.2 Tensor network manipulations

This section serves as an introduction of tensor network manipulations. The overview mainly focusses on MPS/MPO networks, but most of the operations translate to the 2D case.

The MPS's are processed by transforming the tensor into a matrix, performing some matrix calculations and casting it back into its original form. In this way, the standard methods from linear algebra can be used. This section gives some examples how this is done in practice:

## 2.2.1 Basics

### 2.2.1.1 Grouping legs

One of the most basic manipulations is to group some legs of a tensor into one leg:

$$\begin{aligned}
 T^{i_1 i_2 j_1 j_2} &= \begin{array}{c} i_1 \quad i_2 \\ | \quad | \\ \boxed{T} \\ | \quad | \\ j_1 \quad j_1 \end{array} \\
 &\cong (i_1 j_1) \boxed{T} (i_2 j_2) \\
 &= T^{(i_1 j_1)(i_2 j_2)}
 \end{aligned} \tag{2.7}$$

The dimension of the new leg is the product of the dimension of the individual legs. Contracting 2 merged legs with 2 merged legs is exactly the same as contracting them separately. The both The 4 leg tensor and matrix contain exactly the same information. Manipulating this in memory requires both permute and reshape commands. This requires some time, the internal representation of the matrix changes.

### 2.2.1.2 decomposition

The grouping above can be applied to decompose a tensor into 2 tensor with matrix techniques. An example, which will be needed later on, is give here.

$$\begin{aligned}
 \begin{array}{c} i_1 \quad i_2 \\ | \quad | \\ \boxed{O^{uv,vw}} \\ | \quad | \\ j_1 \quad j_1 \end{array} \begin{array}{c} u \\ \text{---} \end{array} \begin{array}{c} w \\ \text{---} \end{array} &= O_{\alpha_u \gamma_w}^{i_1 i_2 j_1 j_2} \\
 &\cong O_{(\alpha_u i_1 j_1)(\gamma_w i_2 j_2)}^{uv} \\
 &= O_{(\alpha_u i_1 j_1) \alpha_v}^{uv} O_{\alpha_v (\alpha_w i_2 j_2)}^{vw} \\
 &\cong \begin{array}{c} i_1 \quad i_2 \\ | \quad | \\ \text{---} \text{O} \text{---} \text{O} \text{---} \\ | \quad | \\ j_1 \quad j_1 \end{array} \begin{array}{c} u \\ \text{---} \end{array} \begin{array}{c} v \\ \text{---} \end{array} \begin{array}{c} w \\ \text{---} \end{array}
 \end{aligned} \tag{2.8}$$

The indices U,V and W represent blocks indices. Step 2 reshapes and groups the indices on to one index on the left and one on the right. The dimension of



this index is the product of the separate dimensions. Step 3 decomposes the matrix into a product of 2 matrices. The last step transforms the indices back to separate legs.

For an exact representation, the bond dimension of virtual level  $v$  is:

$$\dim v = \min(\dim u, \dim v) + 2 \dim i \quad (2.9)$$

Many matrix decompositions exist. Some useful examples here are SVD decomposition, eigenvalue decomposition, QR,  $\dots$ .

### 2.2.1.3 virtual levels

In the previous example, the levels were indicate with a block index or virtual level. The idea is to create separte the contraction into blocks. This is completely analogous to matrix block multiplication. This wil be a more natural form to represent the algorithm. Of course, one can easily switch between block representation and the full one.

### 2.2.1.4 inverse

Suppose we want to find a MPO  $O$  for given tensors  $A$  and  $B$  such that the following holds:

$$\begin{array}{c} i_1 \quad i_2 \quad i_3 \\ \boxed{A} \\ j_1 \quad j_2 \quad j_3 \end{array} \begin{array}{c} u \\ \text{---} \end{array} \begin{array}{c} \bigcirc O \\ i_3 \\ j_3 \end{array} \begin{array}{c} v \\ \text{---} \end{array} = \begin{array}{c} i_1 \quad i_2 \quad i_3 \\ \boxed{B} \\ j_1 \quad j_2 \quad j_3 \end{array} \begin{array}{c} u \\ \text{---} \end{array} \begin{array}{c} v \\ \text{---} \end{array} \quad (2.10)$$

Again, the indices can be taken together in the following way:  $\alpha = (ui_1j_1i_2j_2)$  and  $\beta = (i_3j_3v)$ :

$$A_{\alpha\gamma} O_{\gamma\beta} = B_{\alpha\beta} \quad (2.11)$$

This a a standard matrix equation and can hence be solved with linear algebra packages. Note that it is not necessary to calculate  $A^{-1}$  to obtain the solution. Linear solver are generally much faster. As this is one of the core problems to solve both in 1D and 2D, this will be discussed in detail in section 5.1.

### 2.2.1.5 contraction order

contraction order

### 2.2.1.6 Gauge freedom

gauge

### 2.2.1.7 truncation

svd truncation

## 2.2.2 MPS algorithms

### 2.2.2.1 cononical form

schmidt decomp,

### 2.2.2.2 DMRG

### 2.2.2.3 Expectation values

Suppose that there is an MPO representation of  $e^{-\beta\hat{H}}$   $A$  and that the mpo representation for  $X$   $Y$  is localised over  $n$  sites, then the expectation value is given by:

$$\langle X \rangle = \frac{\text{Diagram with two rows of tensors: the top row has tensors X and A, and the bottom row has tensors A. The tensors are connected horizontally and vertically, forming a closed loop. The diagram is labeled (2.12).}}{\text{Diagram with one row of tensors: the row has tensors A. The tensors are connected horizontally, forming a closed loop. The diagram is labeled (2.12).}}$$

In the thermodynamic limit there are an infinity number of  $A$  to the left and the right. This can be simulated by taking the left and right fixed points of the traced MPO  $A$  corresponding to the largest eigenvector  $\lambda$ .

$$G_l - \text{Diagram with tensor A and a loop} = \lambda G_l \text{ ---} \quad (2.13)$$

$$\text{---} \text{Diagram with tensor A and a loop} G_r = \lambda \text{ ---} G_l \quad (2.14)$$

Equation eq. (2.12) can now be easily calculated:

$$\langle X \rangle = \frac{\lambda^n G_r - G_r}{\text{horizontal line}} \quad (2.15)$$

## 2.3 Tensor network algorithms

### 2.4 MPS algorithms

This section, and the following one, will introduce some different tensor network algorithms. The goal is to provide an intuitive explanation how these algorithms work. For rigorous derivations and mathematical details, other sources can be read.

#### 2.4.1 Canonical form

A translation invariant MPS has the following form:

$$\text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \quad (2.16)$$

It can be easily seen that inserting  $XX^{-1}$  on each bond doesn't change the contracted tensor.

$$A_l = \text{---} \triangle \text{---} = X \text{---} \bigcirc \text{---} X^{-1} \quad (2.17)$$

$$A_r = \text{---} \triangle \text{---} = Y \text{---} \bigcirc \text{---} Y^{-1} \quad (2.18)$$

and

$$\text{---} \diamond \text{---} = XY^{-1} = C \quad (2.19)$$

Then eq. (2.16) can be written as follows:

$$\text{---} \triangleleft \triangleleft \diamond \triangleright \text{---} \quad (2.20)$$

Introducing one more tensor:

$$A_c = \text{---} \square \text{---} = \text{---} \triangleleft \diamond \text{---} = \text{---} \diamond \triangleright \text{---} \quad (2.21)$$

At the moment the matrices  $X$  and  $Y$  are not yet defined. To bring an MPS  $A$  in its canonical form, the following choice is made

$$\begin{array}{c} \text{---} \triangleleft \text{---} \\ \text{---} \triangleleft \text{---} \end{array} = \begin{array}{|c|} \hline \\ \hline \end{array} = I \quad (2.22a)$$

$$\begin{array}{c} \text{---} \triangleleft \text{---} \\ \text{---} \triangleleft \text{---} \end{array} = \begin{array}{|c|} \hline \\ \hline \end{array} = I \quad (2.22b)$$

more on MPS

#### 2.4.1.1 DMRG

### 2.5 2D tensor network contraction

PEPS contraction concerns the following problem:

$$\rho_{i,j} = \begin{array}{c} \dots \quad \dots \quad \dots \quad \dots \quad \dots \\ \dots \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \dots \\ \dots \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \dots \\ \dots \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \dots \\ \dots \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \dots \\ \dots \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \dots \\ \dots \quad \dots \quad \dots \quad \dots \quad \dots \end{array} \quad (2.23)$$

Contract an infinite lattice of identical tensors, with some irregularities on a small patch. For example, the one patch can be used to calculate an expectation value of a certain observable.

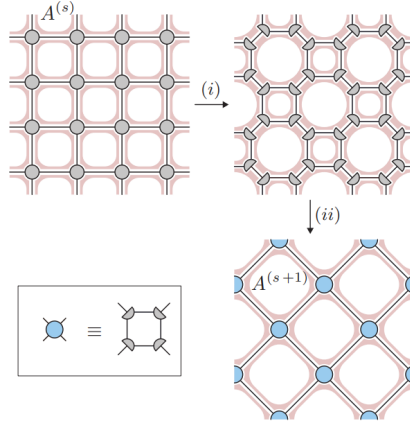


Figure 2.2: Steps in TRG procedure. Figure taken from [5].

### 2.5.1 Overview methods

As this is a difficult task, and often the bottleneck in simulations, many algorithms exist to perform this. The classification is taken from [4]

#### 2.5.1.1 Real-space renormalization-group methods

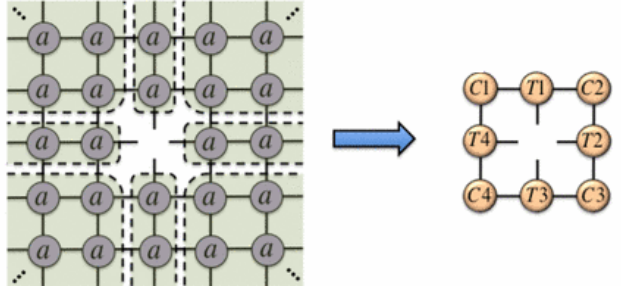
The general idea is to course grain the tensor network. Of the many methods, historically the first one was Tensor renormalization group (TRG) [5], shown in fig. 2.2. In the first step i, the tensor are split with a SVD in 2 different ways, depending on the position on the lattice. ii recombines 4 halves of the decomposition into 1 new tensor. The result is a rotated grid, with side length  $\sqrt{2}$ . The bond dimension is truncated during the SVD step.

Many other variants exist, such as Tensor Network Renormalization (TNR), which wont be discussed here.

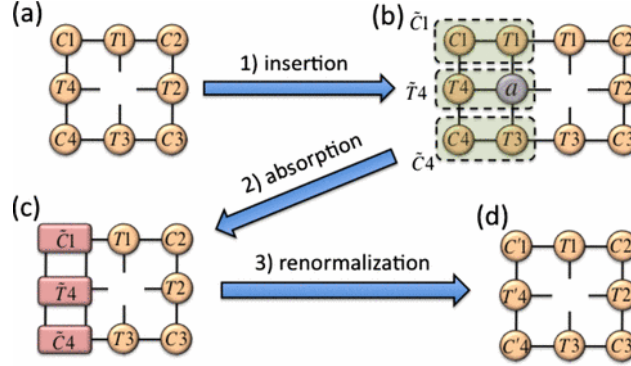
#### 2.5.1.2 Corner transfer matrix methods

Another method goes by the name Corner transfer matrix renormalisation group (CTMRG), as described in [6]. The full network is approximated by 4 corner matrices and 4 C and 4 half row transfer matrices T as shown in fig. 2.3a. The idea is to find matrices such that inserting a row and truncating it again results in the same tensors. This is shown in fig. 2.3b. First, a new row is inserted. From fig. 2.3a it can be seen that this should not change the environment. Some of the tensors are taken together in step 2. As a final step, the tensor are suitably truncated. Once this cycle converges, the environment is known.

Note that many important details are not written down here, and this is merely to give some intuition about tensor network contraction algorithms.



(a) Definition of environment



(b) 3 steps of CTMRG

Figure 2.3: Figures adapted from [6]

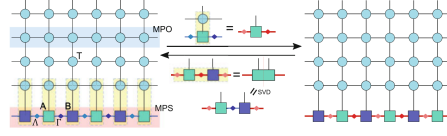


Figure 2.4: Figure taken from [?].

### 2.5.1.3 Boundary methods

The goal of these methods is to find an MPS fixed point for the infinite lattice:

$$\begin{array}{c} \dots \rightarrow \triangle \rightarrow \square \rightarrow \triangleleft \rightarrow \dots \\ \dots \rightarrow \circ \rightarrow \circ \rightarrow \circ \rightarrow \dots \end{array} \approx \begin{array}{c} \dots \rightarrow \triangle \rightarrow \square \rightarrow \triangleleft \rightarrow \dots \\ \dots \rightarrow \circ \rightarrow \circ \rightarrow \circ \rightarrow \dots \end{array} \quad (2.24)$$

**2.5.1.3.1 Time-evolving block decimation** Time-evolving block decimation was introduced as a method to calculate thermal states (this will also be encountered later on in ?? 3.3.3.1.1). It can also be used for contraction.

The idea is shown in fig. 2.4. The boundary MPS is applied, and afterwards truncated.

update

## 2.5.2 Vumps

The purpose of this section is to give some intuition on the variational uniform Matrix Product State (VUMPS) algorithm, which will be used later on in this thesis.

The goal is to find a MPS layer for the MPO such that:

$$\begin{array}{c} \dots \rightarrow \triangle \rightarrow \square \rightarrow \triangleleft \rightarrow \dots \\ \dots \rightarrow \circ \rightarrow \circ \rightarrow \circ \rightarrow \dots \end{array} \approx \begin{array}{c} \dots \rightarrow \triangle \rightarrow \square \rightarrow \triangleleft \rightarrow \dots \\ \dots \rightarrow \circ \rightarrow \circ \rightarrow \circ \rightarrow \dots \end{array} \quad (2.25)$$

This is very similar to section 2.2.2.3. The expression holds approximately, because the MPS on the left hand side has a larger bond dimension than on the right hand side.

### 2.5.2.1 The equations

Suppose there are tensor which fullfil the conditions stated below:

$$\text{Diagram (2.26a)} \quad (2.26a)$$

$$\text{Diagram (2.26b)} \quad (2.26b)$$

These eigentensor equations are solved in pratice in a slightly different manner:

$$\text{Diagram (2.27)} \quad (2.27)$$

Wher the  $A_r$  tensor was connected from below on both sides and eq. (2.22) was used on the right hand side. The blocks in eq. (2.26) form a zipper from the left and right respectively. Each application brings down one of the MPS tensors in the following way.

$$\text{Diagram (2.28)} \quad (2.28)$$



The left and right zipper can now move towards each other, until they meet at the center. In order to become the same MPS as before the application to the MPO (eq. (2.25)), one more condition is needed:

$$\text{Diagram (2.29)} = \lambda \text{Diagram (2.29)}$$

This completely determines the problem, but one more equation is used to solve the problem. Combining on of the equations eq. (2.26), the definition of  $A_c$  eq. (2.29) and one of eq. (2.22) gives  $C$ :

$$\text{Diagram (2.30)} = \lambda \text{Diagram (2.30)}$$

fix gap

Then eq. (2.25) is solved:

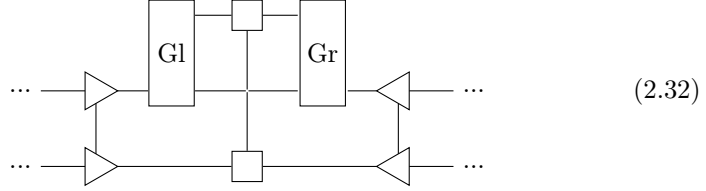
$$\text{Diagram (2.31)} = \text{Diagram (2.31)}$$

check lambdas in literature

Contracting a 2D tensor network is thus reduced to solving the eq. (2.21), eq. (2.26), eq. (2.29) and eq. (2.30) simultaneously. Inspection of the equations show that the following cycle needs to be solved:

- $A_c, C \rightarrow A_l, A_r$  eq. (2.21)
- $A_l, A_r \rightarrow G_l, G_r$  eq. (2.26)
- $G_l, G_r \rightarrow A_c, C$  eq. (2.29) and eq. (2.30)

The calculated environment can now be used to solve the original problem. Due to symmetry, the same MPS can be applied from below. eq. (2.30) now becomes:



### 2.5.2.2 The derivation

While the above derivation is reasonable, it is not very rigorous. The algorithm finds its origins in tangent space methods, such as explained in [7].

Not every state in the many body Hilbert space can be represented by an MPS. For a

By carefully constructing the tangent space and making use of the available gauge freedom, a compact expression can be found for the tangent space projector  $\mathcal{P}_A$ . This projects a state from the many body Hilbert space to the tangent space of an MPS A. For an optimal MPS approximation A, the error made in the approximation should be orthogonal to the tangent space. For eq. (2.25) this means that the application of the projection of the error on the tangent space should be zero, i.e. the MPS is at a variational minimum. [4]

Applying the projector  $\mathcal{P}_A$  to eq. (2.25) exactly gives rise to the equations stated earlier.

## Chapter 3

# Strongly correlated matter

### 3.1 Phases and Criticality

#### 3.1.1 Phases of matter

An important area of research is the study of the different phases of (quantum) matter. A phase is a state of matter in which the macroscopic physical properties of the substance are uniform on a macroscopic length scale. These phase can be measured by thermodynamic function, i.e. by function of a few macroscopic parameters. [8]. More precisely, for a given phase the properties vary as an analytic function of the macroscopic variables.

Interesting physics happens at the boundary between 2 or more distinct phases. The phase transitions were classified by Ehrenfest [9], who looked at the free energy across the phase boundary. If the free energy shows a discontinuity, it is called first order (or discontinuous) phase transition. Similarly, if the derivative shows a discontinuity, it is called second order (or continuous). Higher order phase transitions are possible, and there are even examples of infinite order transitions, such as the BKT transition.

#### 3.1.2 symmetry breaking

Sometimes, but not always, a phase transition is related to spontaneous symmetry breaking. A state  $|\Psi\rangle$  is said to be symmetric under a unitary transformation  $U$  if the state only changes by a phase factor:  $\hat{U}|\Psi\rangle = e^{i\phi}|\Psi\rangle$ . A hamiltonian possesses a symmetry if it commutes with  $U$ :  $[H, U] = 0$  [10]. A remarkable fact is that many ground states are not invariant under a symmetry  $U$  of the hamiltonian.

For phase transitions associated with a broken symmetry, one can define an order parameter. This parameter evaluates to 0 for the symmetric phase, but not for the spontaneous broken phase.

In continuous or second-order phase transitions the order parameter increases continuously from zero as the critical temperature is traversed. The

entropy also changes continuously. On the other hand, the correlation length and related energy scales diverge at the critical temperature. In fact, at the critical temperature of a second-order phase transition, scale invariance systems become scale-invariant, in the sense that physical properties no longer depend on the length (or energy) scale at which they are probed. Many symmetry-breaking phase transitions are second-order, with the onsets of superfluidity, (anti)ferromagnetism and many phases of liquid crystals as famous examples.[10]

### 3.1.3 Universality

Universality looks at the behaviour of the system near a continuous phase transition. These can be described well by so called power laws. For classical phase transitions (driven by temperature) near critical temperature  $T_c$ , observables  $a_i$  depend in the following way on the reduced temperature  $t = \frac{T-T_c}{T_c}$ :  $a_i(t) \sim t^{\alpha_i}$ . One would expect that the set of critical exponents  $\alpha_i$  depends on the precise form of the hamiltonian of the system, but it turns out these exponents can be captured by a limited number of universality classes. This means that the physics near criticality is completely understood once it is understood for one member of the class.

### 3.1.4 Critical exponents for spin systems

The following table defines some of the critical exponents for the Ising system.

Symbol	name
m	magnetisation
$\xi$	correlation length
g	external field
t	reduced temperature
d	dimension

The 2 point correlation function is defined as  $f(x, y) = \langle m(x)m(y) \rangle - \langle m(x) \rangle \langle m(y) \rangle$ . At larger distances this decays exponentially fast (see section 3.1.6)  $f(x, y) = e^{-\frac{|x-y|}{\xi}}$ , where  $\xi$  is the correlation length.

for the ordered phase, the following relations hold:  $m \sim |t|^\beta$ ,  $\xi(t) \approx |t|^{-\nu}$ . At the critical temperature near a quantum phase transition  $m \approx |g - g_c|^{\frac{1}{\delta}}$ .

### 3.1.5 Finite size scaling

Phase transitions only occur for systems with an infinite number of degrees of freedom. This poses a problem, as for instance Monte Carlo simulations only finite grids can be simulated. One computationally expensive way to extract the properties in the thermodynamic limit is by making the grid increasingly bigger until the properties have converged. Fisher's insight was that the properties could also be extrapolated from different finite size calculations by making the following assumption: near a critical point, every thermodynamic property

scales as an universal function of  $L/\xi$ , with  $L$  the size of the system and  $\xi$  the correlation length.

Define  $t = \frac{T-T_c}{T_c}$ . The mathematical formulation is as follows:

$$A(T, L) = L^{\kappa/\nu} f_A(tL^{1/\nu}) \quad (3.1)$$

This holds for  $t$  small (near critical point) and  $L$  sufficient large compared to the lattice spacing. The exponents can be fitted by plotting  $A(T, L)L^{-\kappa/\nu}$  as a function of  $tL^{1/\nu}$  for different sizes  $L$  and temperature  $t$ . For the correct critical exponents and critical temperature, all the points should collapse to one single graph.

From the ansatz eq. (3.1), other ways can be derived to determine certain coefficients.

### 3.1.5.1 Finite size scaling for MPS

The finite size scaling for MPS is somewhat different. In [11], it is argued that  $\delta$  can take the place of  $1/L$ .

Suppose  $\lambda_i$  are the eigenvalues of eq. (2.27) order from largest real part to smallest. Then

$$\epsilon_i = -\log(|\lambda_i|) \quad (3.2)$$

Then

$$\delta = -\sum_i c_i \epsilon_i \quad \sum_i c_i = 0 \quad (3.3)$$

The intuition behind this is as follows: for a MPS approximation, the gaps in the transfer spectrum go always to zero for sufficient large bond dimension. The distance between these gaps is thus a measure for the system size.

### 3.1.5.2 subleading corrections

In [12], it is argued that the form proposed in eq. (3.1) does also not fully take into account the finite size effects. Subleading correction could be introduced as follows:

$$A(T, L) = L^{\kappa/\nu} (1 + cL^{-\omega}) f_A(tL^{1/\nu} - dL^{-\phi/\nu}) \quad (3.4)$$

Indeed, this reduces to eq. (3.1) for sufficient large  $L$ . Because there is always data need around the critical point, the original procedure could be biased and result in wrong parameters. On the other hand, introducing extra parameters can lead to overfitting, again not improving the result.

### 3.1.6 CFT

One of the tools to derive some properties of phase transitions is Conformal Field Theory (CFT). CFT is a quantum field theory, which obeys an additional rule: the physics remains invariant under a conformal transformation. The exact form of these transformation is:

$$g'_{\mu\nu}(x') = \Lambda(x) g_{\mu\nu}(x) \quad (3.5)$$

In 2D, the shape of the correlation functions can be determined exactly, and indeed correspond to the form from previous section.

One of the properties characterising a conformal field theory is the central charge  $c$ , which can only take a discrete number of values. In the case of unitary systems with  $c \leq 1$ , this has turned out to give a complete classification of possible two dimensional critical behavior. [13]

For the 2D Ising model, the central charge is  $c = 1/2$ . A scaling relation used in the results chapter is [14]

$$Le^{6S(T,L)/c} \sim \xi \quad (3.6)$$

### 3.1.7 Quantum phase transitions

A traditional 2nd order phase transition is driven by a change in temperature. Quantum phase transitions on the other hand happen at zero temperature under influence of another parameter  $g$  of the model. At finite temperature, 2 things can happen: either there is a line connecting a classical 2nd order phase transition to the quantum phase transition, or the phase transition disappears at finite temperature [15].

## 3.2 Models

The goal of numerical techniques is to simulate the physics of real world systems. These are, to some extent, captured by different models. Models are a simplified mathematical description that captures some relevant physics of more complicated systems. This section introduces some specific models, their relevance and some properties. These models will be used later to benchmark the developed tensor network expansion.

### 3.2.1 Ising model

The prototypical example of a model in the field of strongly correlated matter is the Ising model. It was first introduced 1925 by Ernest Ising, as a model to capture ferromagnetism. He proved that for a linear chain, there is no phase transition at finite temperature. He wrongly concluded that this would also be the case in higher dimensions, but it turned out to be one of the deepest and far-reaching problems in 20th century [16].

The Ising model, in essence, assigns an energy contribution to neighbouring spins. These spins sit on a fixed position on a chain (1D) or lattice (2D/3D/...). In classical Ising, the operators in the Hamiltonian all commute with each other. An energy is assigned between neighbouring spins and possible energy for alignment with an external magnetic field in the same direction. In quantum Ising model, a transversal field is added. Often, the particles on the grid are spin 1/2 particles, but of course other particles are possible.

q potts,...

Many generalizations exist for the Ising model.

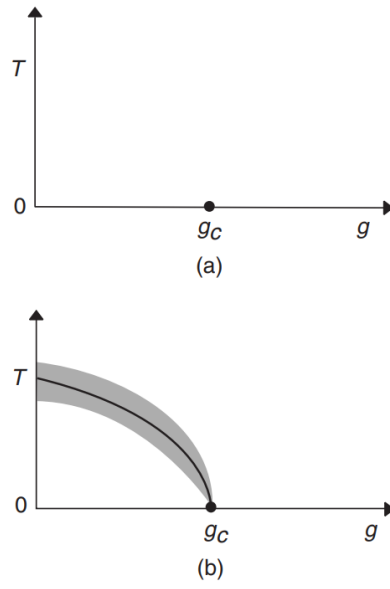


Figure 3.1: Two possible phase diagrams of a system near a quantum phase transition. In both cases there is a quantum critical point at  $g = g_c$  and  $T = 0$ . In (b), there is a line of  $T > 0$  second-order phase transitions terminating at the quantum critical point. The theory of phase transitions in classical systems driven by thermal fluctuations can be applied within the shaded region of (b). Figure and caption taken from [15].

### 3.2.1.1 Classical Ising

The classical ising model is given by the following hamiltonian:

$$H = -J \left( \sum_{\langle ij \rangle} \sigma_i \sigma_j + h \sum_i \sigma_i \right) \quad (3.7)$$

where  $\langle ij \rangle$  runs over all neighbouring lattice sites. The possible values of  $\sigma$  depends on the spin dimension. For spin 1/2 lattices  $\sigma \in -1, +1$ .  $g$  encodes the interaction strength of the external magnetic field.

The sign  $J$  determines the low temperature ground state. A positive  $J$  will tend to align all neighbouring spins at low temperature. This is often called ferromagnetic, because all the aligned spins cause a macroscopic magnetisation. On the other hand, a negative  $J$  causes neighbouring spins to have an opposite sign.

Depending on the sign of the longitudinal field  $h$ , the spins tend to align or anti align with this external field. This lifts the degeneracy of the groundstate.

blabla

**3.2.1.1.1 1D** The classical 1D model was solved analytically by Lens.

**3.2.1.1.2 2D** In 2D, it becomes important to define the lattice. Here, and in the simulations, we will consider a square lattice. This model was famously solved by Lars Onsager in 1944, by using the transfer matrix method. In 2 dimensions, the Ising model has a phase transition at finite temperature. The critical temperature is  $T_c = \frac{2J}{T \ln(\sqrt{2}+1)}$ .

Only the  $h = 0$  case the is solved analytically. For higher dimensions, no analytical solution is known. For these cases, we need to use numerical techniques if we want to understand the behaviour of these models.

On different lattices, interesting thins can happen. For instance, the ground-state of an antiferromagnet on a triangular lattice is not obvious to determine. The spins tend to antialign, but at least 2 of 3 spins on the corner of a triangle have to align. Remarkably, as will be explained in section 3.1, the physics at the phase transition does remain invariant when the lattice is changed.

### 3.2.1.2 Quantum Ising

As we all know, the real world behaves, certainly at small length and time scales, quantum mechanically. Therefore, it is important to understand how the quantum Ising model differs from the classical model. In the quantum Ising model, the operators no longer commute with each other. An example is the transversal Ising model given by the following hamiltonian:

$$\hat{H} = -J \left( \sum_{\langle ij \rangle} \sigma_i^x \sigma_j^x + g \sum_i \sigma_i^z \right) \quad (3.8)$$

In the case that  $g = 0$ , this is the classical Ising model (in the  $h = 0$  case).



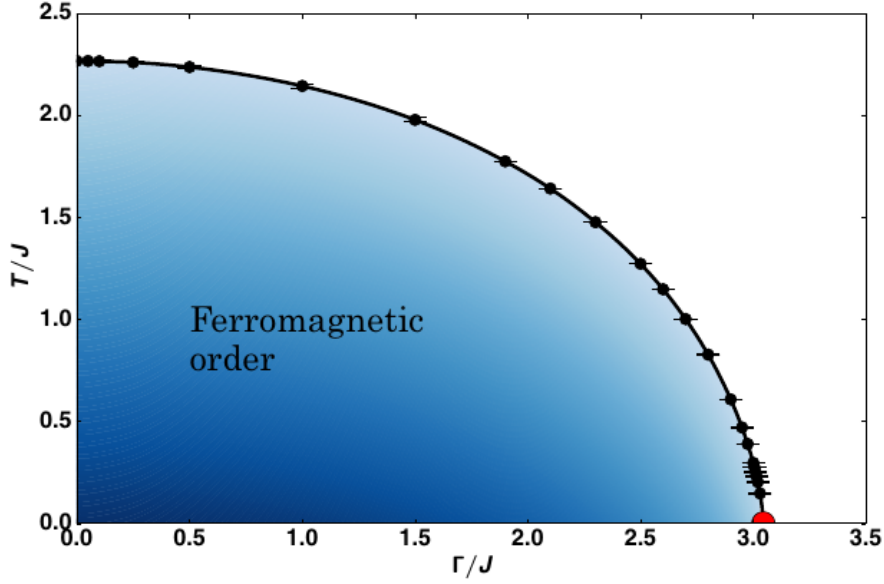


Figure 3.2: Phase diagram for 2D transversal Ising model. Figure taken from [17]

**3.2.1.2.1 1D** Different to the classical case, the 1D model already contains a phase transition.

**3.2.1.2.2 2D**

### 3.2.2 Heisenberg

The heisenberg model is given by:

$$\hat{H} = - \left( \sum_{\langle ij \rangle} J_x \sigma_i^x \sigma_j^x + J_y \sigma_i^y \sigma_j^y + J_z \sigma_i^z \sigma_j^z + h \sum_i \sigma_i^z \right) \quad (3.9)$$

These models have different names depending on the values of  $J_\alpha$  with  $\alpha = x, y, z$ .  $J_x = J_y \neq J_z = \Delta$  is called the XXZ model.

### 3.2.3 Random

It's also possible to construct random hamiltonians.

in basis: hermitian H

### 3.3 Operator exponentials

find citation and examples

While it is often possible to find exact MPO representation to represent a wide class of hamiltonians, it is much harder to do the same for exponentiated operators. These operators play an important role: they act as time evolution operators for quantum systems  $|\Psi(t)\rangle = \exp(-i\hat{H}t)|\Psi(0)\rangle$ . A very similar operator governs the partition function in statistical mechanics: the probability of finding a system at inverse temperature  $\beta = \frac{1}{T}$  in a microstate  $i$  is given by  $p_i \exp\{-\beta\hat{H}_i\}$ . This is often called "imaginary" time, due to the substitution  $\beta = it$ . The ability to calculate these operators is essential for understanding the dynamics of a given quantum model, and making contact with real world observations of these systems at finite temperature.

explain link CFT

#### 3.3.1 Statistical mechanics

The physics of a system in thermodynamical equilibrium can be derived from it's partition function  $Z$ . The classical formula generalises to a density matrix  $\rho$  as follows:

$$\begin{aligned} Z &= \sum e^{-\beta E_n} \\ &= \sum_n \langle n | e^{-\beta \hat{H}} | n \rangle \\ &= \text{Tr}(e^{-\beta \hat{H}}) \end{aligned} \tag{3.10}$$

The first line is the partition function for classical discrete systems. The index  $n$  runs of all possible microstates. It is known that the propability to find the system in a given microstates is given by:

$$p_i = \frac{\sum e^{-\beta E_i}}{Z} \tag{3.11}$$

An useful quantity is the density matrix  $\rho$ .

$$\begin{aligned} \rho &= \sum_j p_j |\Psi_j\rangle \langle \Psi_j| \\ &= \sum_j \frac{e^{-\beta \hat{H}}}{Z} |\Psi_j\rangle \langle \Psi_j| \end{aligned} \tag{3.12}$$

Whith this notation, the partition function  $Z$  and ensemble average of an operator  $\hat{X}$  are given by:

$$\begin{aligned} Z &= \text{Tr}(\rho) \\ \langle X \rangle &= \text{Tr}(\rho \hat{X}) \end{aligned} \tag{3.13}$$

### 3.3.2 Time evolutions

theorema herbekijken

In quantum field theory, calculation of n-point correlation functions is extremely important to understand a given field theory.

#### 3.3.2.1 ground state

One practical way of finding the ground state is cooling an initial state down very small  $T$ .

### 3.3.3 Tensor network methods

In the following section I will give a very short review of the current tensor network methods to simulate real or imaginary time evolution. This overview is mainly based on the review paper [18].

#### 3.3.3.1 Approximations to $\hat{U}(\delta)$

The goal is to approximately make a MPO for a small timestep  $\delta$  which gives a new MPS at time  $t + \delta$ .

**3.3.3.1.1 TEBD** Time-evolving block decimation (TEBD) uses the Trotter-Suzuki decomposition. Suppose the chain is split in even and odd sites.

$$\hat{H} = \hat{H}_{\text{even}} + \hat{H}_{\text{odd}} \quad (3.14)$$

$$\begin{aligned} \hat{U} &= e^{-i\delta\hat{H}_{\text{even}}} e^{-i\delta\hat{H}_{\text{odd}}} e^{-i\delta[\hat{H}_{\text{even}}, \hat{H}_{\text{odd}}]} \\ &\approx e^{-i\delta\hat{H}_{\text{even}}} e^{-i\delta\hat{H}_{\text{odd}}} \end{aligned} \quad (3.15)$$

This is now easy to solve: first apply  $e^{-i\delta\hat{H}_{\text{even}}}$  for every even bond and afterwards  $e^{-i\delta\hat{H}_{\text{odd}}}$ . The error is  $O(\delta^2)$  and the number of steps to reach temperature  $\beta$  is  $\beta/\delta$ . The error can be made small. This can be generalized to higher order schemes.

**3.3.3.1.2 MPO  $W^{I,II}$**  These methods directly use the MPO representation of a certain hamiltonian. This is a more recent method (2015) to construct an MPO first detailed in [19]. The idea is to generalise

MPO w1 and 2

$$1 + \delta \sum_x H_x \rightarrow \prod_x (1 + \delta H_x) \quad (3.16)$$

The error is formally still  $O(\delta^2)$ , but includes many more terms. The advantages lay in the fact that the form above has an efficient representation as an MPO. MPO  $W^I$  and  $W^{II}$  are capable of dealing with long-ranged interaction terms which makes it suitable to simulate 2D systems ??

### 3.3.3.2 global Krylov method

Krylov methods are widely used in linear algebra to calculate eigenvectors. An example is the Lanczos algorithm. These methods are applied to MPS's, but do not fully make use of its structure. For this method, only a MPO representation is needed.

expand on this

### 3.3.3.3 MPS-local methods

**3.3.3.3.1 local Krylov** The krylov methods from previous pragraph can be adapted to work on a reduced basis.

**3.3.3.3.2 TDVP** Time-dependent variational principle (TDVP) can be seen as a further development of the local krylov method.

Its also been formulated in as a tangent space algorithm, similar to the VUMPS derivation (section 2.5.2.2).

$$i \frac{\partial |\Psi(A)\rangle}{\partial t} = \mathcal{P}_{\mathcal{A}(\sqcup)} H |\Psi(A)\rangle \quad (3.17)$$

Where the richt hand side is projected on the tanget space, because the left hand side is a also a tangent vector. (See [7]).

## 3.4 Quantum to classical mapping

In a certain sense, a quantum model in d dimensions can be mapped to a classical model in d+1 dimension.

Due to the Lie product formula

$$e^{A+B} = \lim_{M \rightarrow \infty} (e^{A/M} e^{B/M})^M \quad (3.18)$$

we can rewrite

$$\begin{aligned} Z &= \sum_n e^{-\beta E_n} \\ &= \sum_n \langle n | e^{-\beta \hat{H}} | n \rangle \\ &= \sum_{n_1 \dots n_M} \langle n_1 | e^{-\beta \hat{H}/M} | n_2 \rangle \langle n_2 | e^{-\beta \hat{H}/M} | n_3 \rangle \dots \langle n_M | e^{-\beta \hat{H}/M} | n_1 \rangle \\ &= \sum_{n_1 \dots n_M} \langle n_1 | e^{-\beta \hat{H}/M} | n_2 \rangle \langle n_2 | e^{-\beta \hat{H}/M} | n_3 \rangle \dots \langle n_M | e^{-\beta \hat{H}/M} | n_1 \rangle \end{aligned} \quad (3.19)$$

Where completeness relations  $I = \sum_{n_i} |n_i\rangle \langle n_i|$  inserted M times in line 3. This is similar to a path integral.

The quantum imaginary time has become a spatial dimension. The quantum model of  $d$  dimensions now has the structure of a classical model in  $d+1$  dimensions.

In this way, the 2D transversal field Ising model can be mapped to the 3D classical Ising model. [20]



## Chapter 4

# Construction Cluster expansion

### 4.1 Introduction

#### 4.1.1 Notation

In the following, the external legs and virtual level 0 will be omitted:

$$\begin{array}{c} i \\ | \\ 0 \text{---} \bigcirc \text{---} 0 \\ | \\ j \end{array} = \bigcirc \quad (4.1)$$

$$\begin{array}{c} i_1 \quad i_2 \\ | \quad | \\ 0 \text{---} \bigcirc \text{---} 1 \text{---} \bigcirc \text{---} 0 \\ | \quad | \\ j_1 \quad j_2 \end{array} = \bigcirc \text{---} 1 \text{---} \bigcirc \quad (4.2)$$

$$\begin{array}{c} i_1 \quad i_2 \quad i_3 \\ | \quad | \quad | \\ 0 \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} 0 \\ | \quad | \quad | \\ j_1 \quad j_2 \quad j_3 \end{array} = \bigcirc \text{---} \bigcirc \text{---} \bigcirc \quad (4.3)$$

This hamiltonian consists of 1 and 2 site operators. Of course more general

hamiltonians can also be captured.

$$\hat{H} = \left( \sum_{\langle ij \rangle} H_2^i H_2^j + \sum_i H_1^i \right) \quad (4.4)$$

The same notation will be used to denote the hamiltonin evaluated on the given geometry:

$$\begin{aligned} H(\bigcirc - \bigcirc - \bigcirc) = & H_1 \otimes 1 \otimes 1 \\ & + 1 \otimes H_1 \otimes 1 \\ & + 1 \otimes 1 \otimes H_1 \\ & + H_2 \otimes H_2 \otimes 1 \\ & + 1 \otimes H_2 \otimes H_2 \end{aligned} \quad (4.5)$$

#### 4.1.2 Idea

This chapter shows the main construction of dissertation. A cluster expansion is used to approximate  $e^{\hat{H}}$  for every possible geometry. The goal is to make a MPO/PEPO which captures the tensor exponential in the thermodynamic limit.

symmetry, speed

This cluster expansions introduced in [21]. The main idea is to make an extensive expansion by adding blocks which solve the model exactly on a local patch. Crucially, the expansion is not in the inverse temperature  $\beta$  but in the size of the patches. The local patches are separated by a virtual level 0 bond.

To make this somewhat more precise, the first steps of the expansion are shown here. The smallest patch, i.e. 1 site, encodes the exponential of that hamiltonian.

$$\bigcirc = \exp(-\beta H(\bigcirc)) \quad (4.6)$$

If there were no 2 site interaction, this already captures the full diagonalisation. Of course, such a model wouldn't be useful. The next step is to introduce 2 site interactions, where the one site interactions previously introduced interaction are subtracted from the diagonalised hamiltonian.

$$\begin{aligned} \bigcirc \overset{1}{-} \bigcirc &= \exp -\beta H(\bigcirc - \bigcirc) \\ &\quad - \bigcirc \overset{0}{-} \bigcirc \end{aligned} \quad (4.7)$$

At this stage, all seperated networks with maximally 2 connected sites in a row are diagonalised exactly. Notice that here, 2 new blocks are introduced:



$$\begin{array}{c} i \\ | \\ 0 \text{---} \bigcirc \text{---} 1 \\ | \\ j \end{array} \quad \text{and of course also} \quad \begin{array}{c} i \\ | \\ 1 \text{---} \bigcirc \text{---} 0 \\ | \\ j \end{array} .$$

As can be seen, the dimension

of sublevel 1 needs to be  $d^2$ , with  $d$  the dimension of physical level. Although different possible constructions already differ in the next step, one more step is added to make the construction and notation clear.

$$\begin{aligned}
 \bigcirc \text{---}^1 \bigcirc \text{---}^1 \bigcirc &= \exp -\beta H \left( \begin{array}{c} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \\ \text{---} \bigcirc \text{---}^0 \bigcirc \text{---}^0 \bigcirc \\ \text{---} \bigcirc \text{---}^1 \bigcirc \text{---}^0 \bigcirc \\ \text{---} \bigcirc \text{---}^0 \bigcirc \text{---}^1 \bigcirc \end{array} \right) \\
 &= \exp -\beta H \left( \begin{array}{c} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \\ \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \end{array} \right)
 \end{aligned} \tag{4.8}$$

It is clear that the right-hand side of eq. (4.8) can also be omitted, as it is just evaluating the exponentiated hamiltonian on the same geometry as the left hand side and subtracting all possible contractions of the blocks which were added previously.

### 4.1.3 Preview

In the following sections some possible configurations in 1D and 2D will be discussed. At this point, the focus is on the construction and its bond dimension. The results will be discussed later in chapter 6. T

## 4.2 Construction MPO

### 4.2.1 Type A

This type was originally proposed in [21]. The first few blocks in the expansion are the following:

$$\begin{array}{c}
\bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^1 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^2 \text{---} \bigcirc \text{---}^1 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^2 \text{---} \bigcirc \text{---}^2 \text{---} \bigcirc \text{---}^1 \text{---} \bigcirc
\end{array} \tag{4.9}$$

The following types of blocks appear in the cluster expansion

$$\begin{array}{c} | \\ \text{---}^n \bigcirc \text{---}^m \text{---} \\ | \end{array}, \quad \begin{array}{c} | \\ \text{---}^m \bigcirc \text{---}^n \text{---} \\ | \end{array} \quad \text{and} \quad \begin{array}{c} | \\ \text{---}^n \bigcirc \text{---}^n \text{---} \\ | \end{array} \quad \text{with } n \in \mathbb{N}_0 \text{ and } m =$$

$n - 1$ .

The  $O^{nn}$  block is in defined for a chain with an odd number of sites. The contraction of  $O^{nm}$  and  $O^{mn}$  is defined by for a chain with even order. The decomposition is defined up to a gauge transformation.

**4.2.1.0.1 Dimension** In this scheme, virtual level  $n$  has dimension  $d^n$ . Of course, this dimension can be lowered if some error is allowed for the longest chain.

**4.2.1.0.2 discussion** Type A can form long chains, which where not explicitly optimised for. The question arise whether this will results in accurate results for cyclic systems or not.

## 4.2.2 Type B

Type B only contains blocks of the following form;  $O^{mn}$  and  $O^{n0}$ . The first few blocks are:

$$\begin{array}{c}
\bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^1 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^2 \text{---} \bigcirc \text{---}^3 \text{---} \bigcirc \\
\bigcirc \text{---}^1 \text{---} \bigcirc \text{---}^2 \text{---} \bigcirc \text{---}^3 \text{---} \bigcirc \text{---}^4 \text{---} \bigcirc
\end{array} \tag{4.10}$$

$$\begin{array}{c}
i_n \quad i_{n+1} \\
| \quad | \\
\text{m} \text{---} \bigcirc \text{---} \text{n} \text{---} \bigcirc \text{---} 0 \\
| \quad | \\
j_n \quad j_{n+1}
\end{array} = U^n \Sigma V^\dagger \quad (4.11)$$

The following split is made:  $O^{mn} \cong U^n$  and  $O^{n0} \cong \Sigma V^\dagger$ . In this way the left inverse exists and doesn't need any calculation:  $O^{mn} = U^\dagger$ .

**4.2.2.0.1 dimension** From the construction the bond dimension grows from the left to the right. For the last step, there are only  $d^2$  non zero singular values. Each steps adds  $d^2$  to the dimension. For the last step, only  $d^2$  non zero singular values need to be kept. With the following notation:

$$\begin{array}{c}
i \\
| \\
\text{m} \text{---} \bigcirc \text{---} \text{n} \\
| \\
j
\end{array} = A_{(\alpha ij)\beta}^m \quad (4.12)$$

$$\begin{array}{c}
i \\
| \\
\text{n} \text{---} \bigcirc \text{---} 0 \\
| \\
j
\end{array} = B_{(\alpha ij)\beta}^n$$

The bond dimension of lower virtual levels can be reduced if we can solve the following equations simultaneously:

Then the MPO doesn't change if there are matrices  $A'^n$ ,  $A'^{n+1}$  and  $B'^n$  such that

$$\begin{aligned}
S &= A^m A^n = A'^m A'^n \\
T &= A^m B^n = A'^m B'^n
\end{aligned} \quad (4.13)$$

Such matrices with optimal bond dimension can be found with generalised SVD. Generalised SVD decomposes 2 matrices as follows:

$$\begin{aligned}
S^\dagger &= (U \Sigma_1) Q^\dagger \\
T^\dagger &= (V \Sigma_2) Q^\dagger
\end{aligned} \quad (4.14)$$

The new bond dimension is the  $\dim n' = d^2 \cdot \min(\dim n - 1, \dim(n + 1))$ . This is higher than the dimension of type A.

**4.2.2.0.2 Discussion** The bond dimension is larger than type A, but the long chains from A are absent. The left inverse is always well defined and doesn't need any computation, because hermitian matrix  $U$  can be inverted easily. One major drawback is that for long chains, the virtual bonds are very large before they can be shrunk with the gsvd procedure.

primed virtual levels

### 4.2.3 Type C

This type implements the same strict type as Type B, but in a different way. No calculation is involved, except the calculation of the the exponentiated hamiltonian to certain order. The following kind of MPO strings are allowed:

$$\begin{array}{c}
 \bigcirc \\
 \bigcirc \text{---} 1 \text{---} \bigcirc \\
 \bigcirc \text{---} 1' \text{---} \bigcirc \text{---} 1' \text{---} \bigcirc \\
 \bigcirc \text{---} 1'' \text{---} \bigcirc \text{---} 2'' \text{---} \bigcirc \text{---} 3'' \text{---} \bigcirc \\
 \bigcirc \text{---} 1''' \text{---} \bigcirc \text{---} 2''' \text{---} \bigcirc \text{---} 3''' \text{---} \bigcirc \text{---} 4''' \text{---} \bigcirc
 \end{array} \tag{4.15}$$

and so forth. All but one MPO elements are chosen to be the identity matrix. The middle one is the exponentiated hamiltonian with reshaped legs.

**4.2.3.0.1 discussion** As can be expected from the construction, the bond dimension grows very fast. This type is just as precise as Type B.

### 4.2.4 Type D

This type uses a different setup which tries to capture the best of both Type A and B. Type could handle long range correlation better because of the introduction of  $O^{nn}$ , but the inverse was not necessarily well defined. Type B had well conditioned inverses, but performed in most of the cases worse. The block appearing in type D are as follows:

$$\begin{array}{c}
 \text{---} \bigcirc \text{---} \bigcirc \text{---} \bigcirc \text{---} \\
 \text{m} \quad \text{n} \quad \text{n} \quad \text{m}
 \end{array}
 \quad \text{and} \quad
 \begin{array}{c}
 \text{---} \bigcirc \text{---} \\
 \text{n} \quad \text{n}
 \end{array}$$

Similar to type A,

$$\begin{array}{c} \text{m} \\ \text{---} \bigcirc \text{---} \text{n} \end{array} \begin{array}{c} \text{---} \bigcirc \text{---} \text{n} \\ \text{---} \end{array} \begin{array}{c} \text{---} \bigcirc \text{---} \text{m} \\ \text{---} \end{array} = \text{n} \begin{array}{c} \text{---} \boxed{L_n^{-1} M_{2n+2} R_n^{-1}} \text{---} \text{n} \\ \text{---} \end{array} \quad (4.16)$$

$$= U \Sigma V^\dagger$$

Matrix  $D_n$  is the singular value diagonal matrix divided by a normalisation factor  $\phi$ . Both  $U$  and  $V$  are multiplied by  $\sqrt{\phi}$ .

**4.2.4.0.1 discussion** It's not completely clear what the values of  $\phi$  should be. If  $\phi$  is too large, large chains are not suppressed. If  $\phi$  is too small, the  $O^{nn}$  blocks will become large and hence the chain will diverge again. A reasonable value is the sum of the singular values.

other things could be tried here, WIP

**4.2.4.0.2 matrisation** The cost of this type lies in the fact that it has no compact way of casting it to a matrix. The following works, but has quite a large dimension:

$O_{00}$	$O_{01}$	$O_{12}$	$-2O_{01}$	$-2O_{12}$	$O_{01}$	$O_{12}$	$O_{01} D_1^{1/2}$
$O_{10}$	$O_{21}$	$O_{21}$	$O_{10}$	$O_{21}$	$O_{10}$	$O_{21}$	$D_1^{-1/2} O_{12} D_2^{1/2}$
$D_1^{1/2} O_{10}$	$D_2^{1/2} O_{21} D_1^{-1/2}$	$D_1^{-1/2} O_{12} D_2^{1/2}$	$O_{11}$	$O_{22}$	$O_{11}$	$O_{22}$	$O_{11}$

fix this

## 4.2.5 Type E

Again, this is a strict variant which needs exactly twice the bond dimension of type A. The idea is to split every chain in a left and a right part. For the left part, the numbers increase while right part they decrease. This construction carries over well to higher dimensions. The first few blocks are:

$$\begin{array}{c}
 \bigcirc \\
 \bigcirc \text{---} 1 \text{---} \bigcirc \\
 \bigcirc \text{---} 1 \text{---} \bigcirc \text{---} 1' \text{---} \bigcirc \\
 \bigcirc \text{---} 1 \text{---} \bigcirc \text{---} 2 \text{---} \bigcirc \text{---} 1' \text{---} \bigcirc \\
 \bigcirc \text{---} 1 \text{---} \bigcirc \text{---} 2 \text{---} \bigcirc \text{---} 2' \text{---} \bigcirc \text{---} 1' \text{---} \bigcirc
 \end{array} \quad (4.17)$$

The construction is very similar to type A

### 4.2.6 Type F

The idea behind this type is very similar to type D. The blocks look as follows:

$$\begin{array}{c}
 \bigcirc \\
 \bigcirc \xrightarrow{1'} \bigcirc + \bigcirc \xrightarrow{1'} \bigcirc \\
 \bigcirc \xrightarrow{1} \bigcirc \xrightarrow{1} \bigcirc \\
 \bigcirc \xrightarrow{1} \bigcirc \xrightarrow{2} \bigcirc \xrightarrow{1} \bigcirc + \bigcirc \xrightarrow{1} \bigcirc \xrightarrow{2'} \bigcirc \xrightarrow{1} \bigcirc \\
 \bigcirc \xrightarrow{1} \bigcirc \xrightarrow{2} \bigcirc \xrightarrow{2} \bigcirc \xrightarrow{1} \bigcirc
 \end{array} \tag{4.18}$$

The blocks  $O^{n-1,n}$  and  $O^{n,n-1}$  unitary matrices from the svd decomposition, scaled by the largest singular value. The blocks  $O^{n-1,n'}$  and  $O^{n',n-1}$  are then used to actually solve the problem for the given chain. In the next step,  $O^{n,n}$  is added as usual. The idea here is once again to keep this block small, in order to not cause any divergences.

### 4.2.7 Conclusion

Many 1D constructions are discussed here. They can be roughly divided in 2 groups. B, C and E are strict variants, meaning only the explicitly constructed blocks will appear in the final expansion. As a consequence, they have exactly the same predictive power. While B has some advantageous properties such as its final bond dimension and well-defined inverses, type E will be used in the results chapter due to its simplicity and scalability. It also generalises well to 2D, in contrast to type B.

The second category are the unstrict types. Type A has the lowest possible bond dimension to exactly represent a chain of a given length. One hurdle to be overcome are the badly conditioned inverses, when implemented naively. Types D and F try to remedy this. D scales very badly with the maximum number of sites, and has a construction which doesn't fit in with the simple diagrams. The construction was only implemented in 1D code due to this. This type won't be reported, and has a similar performance to type F.

In short, type A, E and F will be reported in section 6.1.

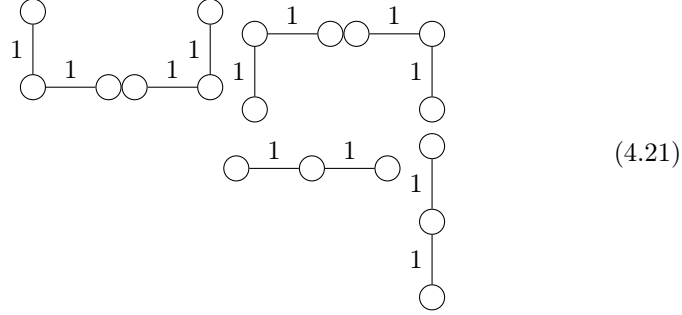
## 4.3 Construction PEPO

While there were some interesting choices in the 1D construction, the number of possibilities in 2D is virtually limitless. The focus will mainly be to generalise type A to 2D. As can be expected, the construction starts of quite similar:

$$\bigcirc \tag{4.19}$$



From 3 blocks onwards, the number of extra blocks start to increase:



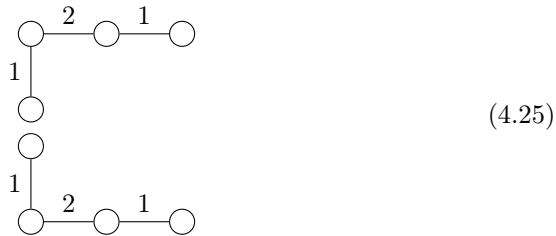
Of course, besides linear blocks also T and + shaped need to be constructed to fully capture the model.



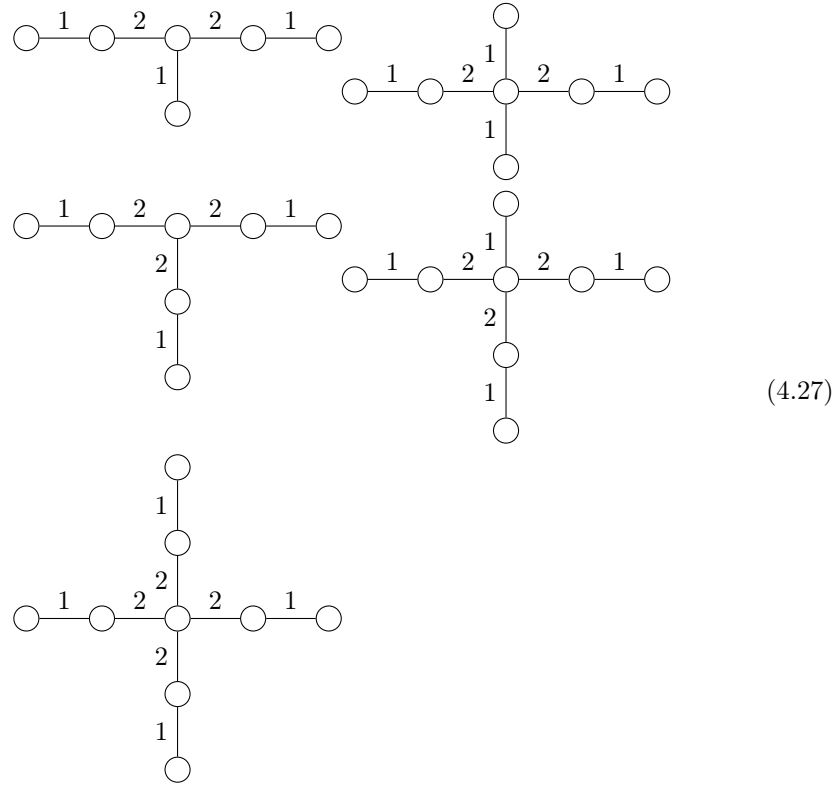
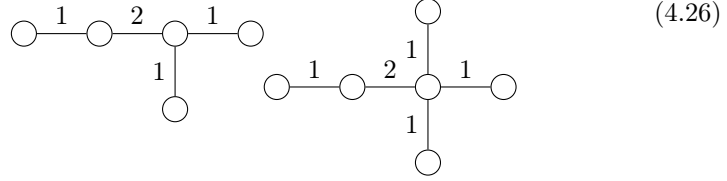
Here, and in the following, care has to be taken in which order the blocks are made. In general, every block which fits in the given block needs to be constructed earlier. The construction continues by introducing a 2 block as before:



Again, 2 blocks are introduced. For other variations of the linear chain, only one block needs to be solved



Due to the way they are constructed, the error for every linear chain of a given length will be the same as in the 1D case. Once again, all possible T and + blocks are created. Some of these blocks are shown here:



For each block shown, there are still multiple permutations of the legs possible. It is clear that a completely automated solver is needed to construct all these different blocks. From here on the construction generalises easily to higher block numbers, and to higher dimensions.

The difference between eq. (4.26) and the blocks in eq. (4.27) is in the largest chain. For eq. (4.26), only chains of order 4 are present, while eq. (4.27) has chains of length 5. It seems that when a virtual level is present, it is most advantageous to create both chains and all blocks, but this will not be the case when a virtual level is truncated.



### 4.3.1 loops

While the blocks above certainly encode a large number of finite size patches, there are still quite some patches need to be encoded. The simplest case is a 1 square loop.



It is clear that this problem cannot be solved with the techniques from the previous section. The square loop needs a new virtual level  $\alpha$ . In general, all the loop levels will be named with greek letters for convenience. The simplest choice for the loop is:



At this points all blocks of order 4 are solved.

#### 4.3.1.1 Single extensions

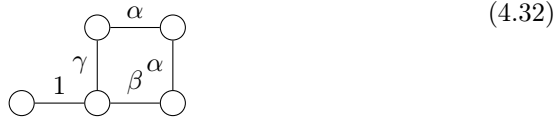
The loops need to be connected to the linear blocks. One way to do this is as follows:



With the given blocks, the following combination is also possible

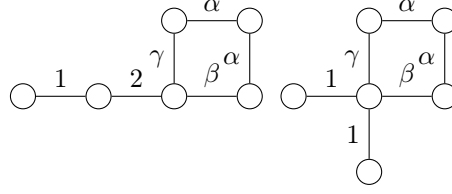


This results in very large errors, and it would require many more blocks to be added in order to counteract this. Luckily, there are many options in 2D. One example is:



No other combinations are possible, except a loop with on one corner an extension. Of course, there is no need to stop here, the following blocks can now be

constructed easily.



#### 4.3.1.2 Double extensions

It seems as if one of the corner pieces can be used as follows:

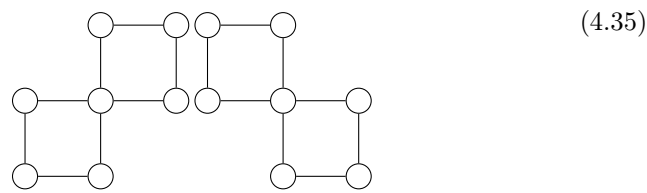


But in order to make a meaningful change in the residual error, the bond dimension of both  $\alpha$  and  $\beta$  need to be enlarged significantly. It is more advantageous to introduce yet another level  $\delta$ , which forms the link between the 2 parts. As both corner tensors can be optimised at once, the total bond dimension is lower than for the previous suggestion, but still larger than the dimensions of the other loop levels.

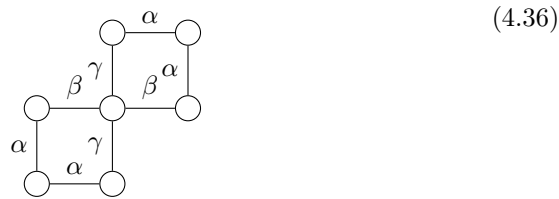


#### 4.3.1.3 Larger loops

One question which comes to mind is where the focus should be for the construction. Making blocks for all possible single loop extensions comes at an increasing cost in total bond dimension. From a physics point of view, the model will be better approximated when smallest non solved patch is solved by introducing new blocks. On the other hand, the already included blocks may cause an error which was not present before the blocks were introduced. One example is a linear chain which closes upon itself, in for instance a  $2 \times 3$  rectangle. Another example are  $+$  blocks which connect upon themselves in the following shape:



One way to solve this, without breaking the rotation pattern of  $\beta$  and  $\gamma$ , is



But this block introduces an infinite tiling, corrupting the expansion.

4.3.1.4 failed ideas

Here it starts to get really tricky.



## Chapter 5

# Framework implementation

An expert is a person who has  
made all the mistakes that can be  
made in a very narrow field

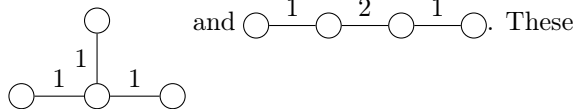
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Niels Bohr

### 5.1 Solvers

The construction can be written down quite compactly as done in the previous section, but actually implementing the code in full generality is somewhat more complex. In practice, the 1D and 2D implementation were performed separately, where a part of the code but mainly the ideas and lessons learned from 1D were taken to the 2D code. Of course, the 1D construction is just a subset of the 2D implementation. In fact, the 2D implementation even outperforms the 1D implementation for reasons with will be explained later. The main focus of this chapter will go to the 2D implementation. At the end some particular optimisations in 1D will be highlighted.

Solving the blocks introduced in the previous sections need 2 different approaches. The graphs of the maps can be split in 2 groups. The first one, considers problems where there are no loops and at most one node with more than 2 legs Examples include:



will be reduced to a standard matrix problem, and solve with matrix (pseudo-) inversion.

The other group, of course, constitutes the nonlinear problems. This includes every problem where a block (or rotated version) occurs more than once, problems which include loops, ...

### 5.1.1 Linear solver

The linear solver is a general purpose block solver which reduces the problem to a set of linear matrix equations. Linear block consist of a tree structure, where the new block is the root of the tree, and all the branches need to be inverted. Let  $I^m = (i_1^1 i_2^1 \dots i_{n_1}^1)$ , then the problem can in general, after some tedious tensor leg bookkeeping, be rewritten in the following form:

$$\begin{aligned} & A_{I_1 I_2 \dots I_n \alpha^1 \alpha^2 \dots \alpha^m} X_{\alpha^1 \alpha^2 \dots \alpha^m j} \\ & = B_{I_1 I_2 \dots I_n j} \end{aligned} \quad (5.1)$$

Here  $i_N^M$  has the following meaning: M numbers the different legs or branches of the tree, N number of sites of the leg and i numbers the bra and ket states and has dimension  $d^2$ . Hence the bond dimension of  $I_n = d^{2n_m}$ . The most obvious way to solve this system is by using a linear solver. The problem is that the bond dimension increases very fast: matrix A has dimension  $d^{2 \sum_m n_m} \times d^{2 \sum_m n_m}$ . Although using a linear solver instead of full inversion is considerably faster, this becomes infeasible for very quickly. A second method consist of solving the following sequence of linear problems one leg at a time:

$$\begin{aligned} A_{I^1 \alpha^1}^1 X_{\alpha^1 I^2 \dots I^m j} &= B_{I_1 I_2 \dots I_n j} \\ A_{I^2 \alpha^2}^2 X_{\alpha^1 \alpha^2 I^3 \dots I^m j} &= B_{\alpha^1 I_2 \dots I_n j} \\ &\vdots \\ A_{I^m \alpha^m}^m X_{\alpha^1 \alpha^2 \dots \alpha^m j} &= B_{\alpha^1 \alpha^2 \dots \alpha^{m-1} I_m j} \end{aligned} \quad (5.2)$$

While this method is very quick and scales well, in practice it results in unstable result. This is a result of the potentially ill conditioned inverses inherent to the construction. A pseudo-inverse of the full matrix can be easily obtained and resolves this issue. Solving in a sequential way, the errors of the pseudo-inverses accumulate. Luckily the problem can be resolved by first performing an SVD decomposition of  $A^m = U^m S^m V^{m\dagger}$  matrices, with S diagonal and U and V unitary. All the  $U^m$  matrices can be inverted by applying the hermitian transpose to B. The Tensor  $S^1 \otimes S^2 \dots \otimes S^m$  is very sparse and can be inverted at once. The last step consist of inverting all unitary V.

link to right section

### 5.1.2 Nonlinear solver

In some cases, the above solver does not return the best possible solution to a given problem. The reason is that it is not able to incorporate symmetries or solve problems where the new blocks appear more than once. A new solver is needed which does not rely on methods from linear algebra, but on more general non-linear least squares solvers.

In essence, the non-linear least squares solver needs as input a vector with the error values  $\vec{f}(\vec{x})$ , and if possible also the jacobian, i.e.  $J_{I,J} = \frac{\partial f_I}{\partial x_J}$ . An

improved point  $x$  is chosen by the algorithm, until some convergence criterium is reached. The implementation uses matlab `fsolve` routine, which uses Levenberg-Marquardt algorithm under the hood.

**5.1.2.0.1 automatic differentiation** With some care, the jacobian can be calculated for a general tensor network in an automated way. Suppose we want to differentiate the contracted tensor  $T^{i_1 \dots i_n}$  with respect to one of the PEPO blocks  $x_n = O_{\alpha\beta\gamma\delta}^{i_n}$ . Denote  $I = (i_1 \dots i_n)$  and  $J = (i_m \alpha \beta \gamma \delta)$ , and this block only occurs once. Then  $J_{IJ} = \frac{\partial T^{i_1 \dots i_n}}{\partial O_{\alpha\beta\gamma\delta}^{i_m}} = T_{i_m \alpha \beta \gamma \delta}^{i_1 \dots i_n} \delta_{i_m}^{i_n}$  amounts to contracting the network with the tensor  $x_m$  removed, and treating the non contracted indices as external ones. If a tensor appears in multiple places, the sum of these contributions has to be taken into account.

source

**5.1.2.0.2 Symmetry** The non-linear solver can handle rotated and permuted blocks. For instance, a simple loop (square) can be solved by rotating one tensor  $T_{\alpha\alpha 00}^I$  4 times, once for every corner. Another example is the following decomposition:  $X_\alpha^I X_\alpha^J = T^{IJ}$ .

### 5.1.3 Sequential linear solver

While from the previous section it seems all non-linear problems need to be solved with the non-linear solver, this is in fact not the case. For instance the following corner block can perfectly be solved with the linear solver.



The loop is treated exactly the same as regular leg. The error is of the same magnitude as the non-linear solver, but performs much faster, and is therefore the solver of choice. For problems with multiple tensors, which may be a rotated version of each other such as in a 2x3 rectangle, the linear solver can be useful.



The idea is to solve each of the tensors with a linear solver. As this is not truly a linear system, the error will not be zero after one pass. But solving the tensors repeatedly lowers the error at each step, giving an iterative procedure. This procedure can be sped up by reusing some parts of the calculations involved in the linear solver. For example, the exponentiated hamiltonian and contraction of all virtual levels that do not involve a given block only need to be performed once.

## 5.2 Optimisation

### 5.2.1 Bookkeeping

One important aspect of programming these general solvers is to devise a scheme that keeps track of all the involved tensors and transforms to problem to the form discribed above. In the code, the geometric info is represented by a map. This keeps track of the neighbours for each site, numbers the internal and external legs and a list to perform the contractions.

The framework provides some tools to transform these maps into other maps, for instance by removing 1 site.

### 5.2.2 Fast contraction

One particular task is to determine all the possible combinations of virtual levels for a given geometry. Simply looping over all possible combinations scales as  $n^m$ , with the number of virtual levels and m the number of internal legs. This quickly becomes a bottleneck. This problem can be restated as a PEPS contraction in the following way: for each site make a tensor  $T_{\alpha\beta\gamma\delta}^i$  where i encodes all the non-empty combinations of legs  $(\alpha\beta\gamma\delta)$ . On each site, the right boundary conditions need to be applied to get the right geometry. After setting the boundary conditions, the sparse PEPS network can be contracted and the resulting tensor gives, after decoding, all the possible contractions. Due to its sparsity, this performs quite fast. As an added bonus, removing a tensor from T gives all contractions without this tensor. As both results are sorted list, the subset of contractions containing a given tensor can also be found fast.

### 5.2.3 Normalisation

For many of the end results, the PEPO cells can be divided by a normalisation factor. Normalising the calculations is important, because  $\exp(\hat{H})$  scales exponentially in the number of sites. Luckily, the exponential can be calculated directly in normalised form. Suppose H is the matrisation of the hamiltonian evaluated for a certain geometry. This is a hermitian matrix and can be diagonalised  $H = QDQ^\dagger$  with Q unitary. Then

$$\exp(H_{d^N} - N\alpha I) = Q \exp(D - N \log(\alpha) I) Q^\dagger \quad (5.5)$$

$$= Q \begin{bmatrix} \exp(D_{11} - N \log(\alpha)) & & \\ & \ddots & \\ & & \exp(D_{d^N d^N} - N \log(\alpha)) \end{bmatrix} Q^\dagger \quad (5.6)$$

$$= \frac{\exp(H_{d^N})}{\alpha^N} \quad (5.7)$$

. With  $I$  the unit matrix. Next to a global normalisation factor, every block calculation calculates a specific normalisation factor such that the largest eigen-



value of  $\exp(H)$  is of the order 1.

### 5.2.4 Internal representation

Two main internal representations are used to construct the given MPO. Either, the MPO is stored as a cell of matrices, or as one big matrix where the blocks are added to during the construction. The output type can be chosen. For some types, sparse matrices are used during the construction. Given that Matlab doesn't support multidimensional matrices by default, this library is used. [expand](#)

### 5.2.5 Even faster inverses

While the inversion procedure above states how to make use of pseudoinverses, it was not yet clear in the 1D case it was needed. The 1D implemetation uses a trick to get all the inverses for free from the SVD decomposition. Take the MPO which corresponds to a unitary matrix:

$$\begin{array}{c} i \\ | \\ \delta \text{---} \bigcirc O_n \text{---} \beta \\ | \\ j \end{array} \cong U_{\alpha(ij\beta)}^n \quad (5.8)$$

Then the inverse MPO can be calculated by taking its Hermitian conjugate and reshaping.

$$\begin{array}{c} i \\ | \\ \beta \text{---} \bigcirc O_n^{-1} \text{---} \gamma \\ | \\ j \end{array} \cong U_{(ij\beta)\gamma}^{n\dagger} \quad (5.9)$$

The left inverse looks like

$$2 = \begin{array}{c} \alpha \text{---} \bigcirc O_n^{-1} \text{---} \bigcirc O_m^{-1} \text{---} \dots \text{---} \bigcirc O_1^{-1} \text{---} 0 \end{array} \quad (5.10)$$

where the physical indices need to be contracted with the corresponding indices of the right hand side. [update](#)

### 5.2.6 Buffering Results

Some calcalations, such as calculating the matrix exponential, take some time. In 1D code, the same calculations were performed over and over again, and hence

a buffer mechanism was written to store these results. In the 2D framework, this not necessary as the solvers only calculate the matrix exponential once and return the blocks together with the made error.

profiling

### 5.2.7 Profiling

## 5.3 Calculating phase diagrams

This section details how the phase diagrams are calculated, stored and the critical parameters fitted. The results are discussed in section 6.2.2

### 5.3.1 Points sampling

This concerns the problem which temperatures to select to calculate the the phase diagram. As the transition between 2 phases is sharp, an uniform sampling in  $T$  is not the best option. A very fine grid is needed to capture the transition well, requiring high computation times. The sampling starts by calculating the magnetisation for  $N$  uniformly sample points between 2 temperatures. These calculations are performed in parallel on a multicore server. When they are all finished (or have run for a maximum amount of iterations), all the arch lengths are calculated, and repeatedly a new  $T$  point is inserted in the largest interval until  $N$  new points are selected. The arch length in the  $m$ - $T$  plan can be changed to require more points in the  $m$  direction than  $T$  direction.

### 5.3.2 Storing the information

Each run has a template with all the common model info. For each point 2 files are stored. One file contains the info and results, such as temperature, magnetisation, correlation length, etc. The other file is much larger and contains the PEPO tensor, the calculated vumps environments,... The files of the first kind are used in other calculations, such as fitting procedure. Reassembling the files into one structure happens in a central function. Another function is able to reprocess already calculated points. The sampling can be continued from where it was last stopped.

mps bond scaling, ...

### 5.3.3 Fitting

## 5.4 How to use

All the code needed to generate all the results from this dissertation is available on my github page [https://github.com/DavidDevoogdt/Thesis\\_Tensor\\_Networks](https://github.com/DavidDevoogdt/Thesis_Tensor_Networks). The starting points to explore the code are in the readme file.

### 5.4.1 Source code structure 1D

The source code for this project can be found on github. The implementation of these types can be found under `src/generateMPO.m`. In this class the different types of MPO can be constructed. It bundles some helper functions such as contracting a chain or cycle of MPO's or construction of an exponentiated hamiltonian for the given input hamiltonian. Other examples are making  $L_n^{-1}$  by sequential invers MPO contractions,...

`src/test.m` contains the code to create the plots to compare different types and orders. The other files in the folder are self-explanatory.

### 5.4.2 Source code structure 2D

source code 2D

## 5.5 Limitations and outlook

In short, all components are in place in the framework to generate easily and efficiently all the blocks. Since day one, this is a constant work in progress

### 5.5.1 Size limitation

The main bottleneck is, as expected, calculating the matrix exponential for large systems. The other components are efficient enough to not cause any troubles.

### 5.5.2 Lattices

The models studied were all on a square lattice. The universal physics does not depend on the details of the model, such as the lattice. It would be beneficial to be able to simulate to these lattices, but here also higher dimensions could be included. In essence all the information of the lattice is contained in the maps generated for each calculation, such as all the connected sites, how to contract them, etcetera. Although undoubtedly many details will need to be changed to use it in practice, the solvers can stay almost the same.

### 5.5.3 Symmetries

At the moment rotation and permutation symmetries can be included in the construction of the blocks. Internal symmetries are not yet included at the moment. Including them could push the computational boundaries further.

### 5.5.4 Code quality

The first and most important goal of writing numerical code is of course that it compiles and that the results are as correct. But this is only the first step. The 2D framework neatly orders the different task is functions to avoid as much code duplication as possible. For instance, there are functions generating and

manipulating maps, which depends on the geometry of the problem. Other functions only use the bonds and contractions listed in that map object, and hence another lattice can be used without changing the other components too much.

Another example is the solvers, which get the blocks through a common

finish this

function

# Chapter 6

## Results

With four parameters I can fit an  
elephant, and with five I can  
make him wiggle his trunk

---

John von Neumann

This section explains the accuracy and performance of the given cluster expansions. The first section compares the different constructions in 1D based on the error relative to the exact solution. The best algorithm will form the basis for the 2D results. First, similarly to 1D, the results will be checked based on the error relative to the exact solution. Then, the expansion is used to calculate the phase diagram of the 2D transversal field Ising model.

### 6.1 Results 1D

In this section the results of the MPO construction detailed in

#### 6.1.1 Exact tensor matrix exponential

The performance of the MPO construction can be compared with the exact diagonalisation of the hamiltonian for a given number of sites. To obtain a faithful results, the number of sites should be as high as possible. In practice, diagonalisation of large matrices becomes slow and memory consuming. The size grows exponentially in the number of sites:  $d^n \times d^n$ . A double takes 8 bytes of memory. A Rough estimated of the amount of RAM  $R$  needed to store this complex array is:

$$R = d^{2n} \times 16bytes \tag{6.1}$$

Which means a 14 site chain already takes up more than 4 GB of RAM. The complexity to calculate a matrix exponential scales as  $O(n^3)$  [22]. In practice

State of the art algorithm for exact diagonalisation, which include all symmetries and are optimised for parallelisation, can calculate up to 50 sites. [23]

make version for cyclic

trace norm, Schatten p  
norm, ...

(6.2)

Calculating the cyclic norm comes at the extra cost of contracting a cyclic tensor network. In this chapter, the cyclic norm will be given for  $M=8$  sites.

#### 6.1.2.1 Full pseudo-inversion

**6.1.2.1.1 Truncation** The original 1D code does not yet have this full inversion procedure. Instead an optimality criterium was devised to truncate the series. Needless to say, even with this criterium, the results were a lot worse at low beta

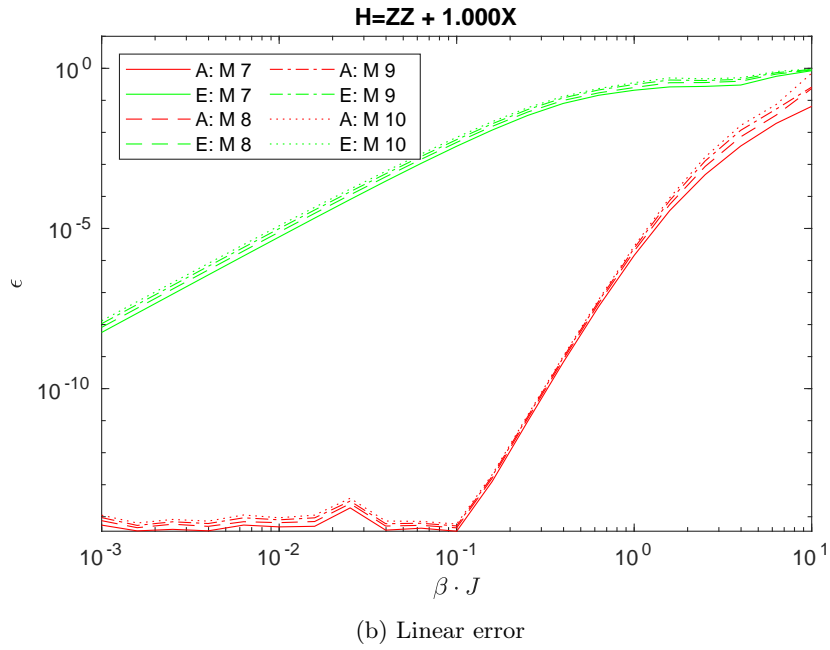
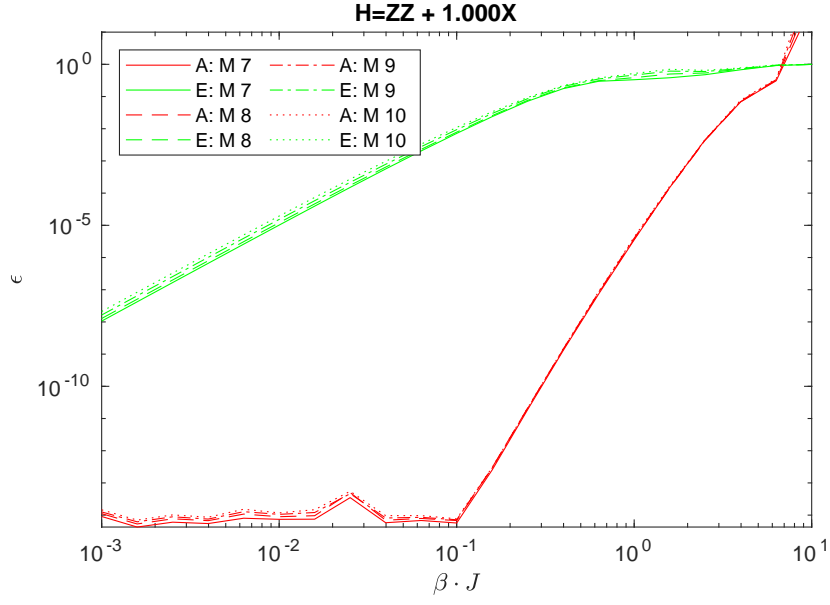


Figure 6.1: Different error measures for 1D transversal Ising model

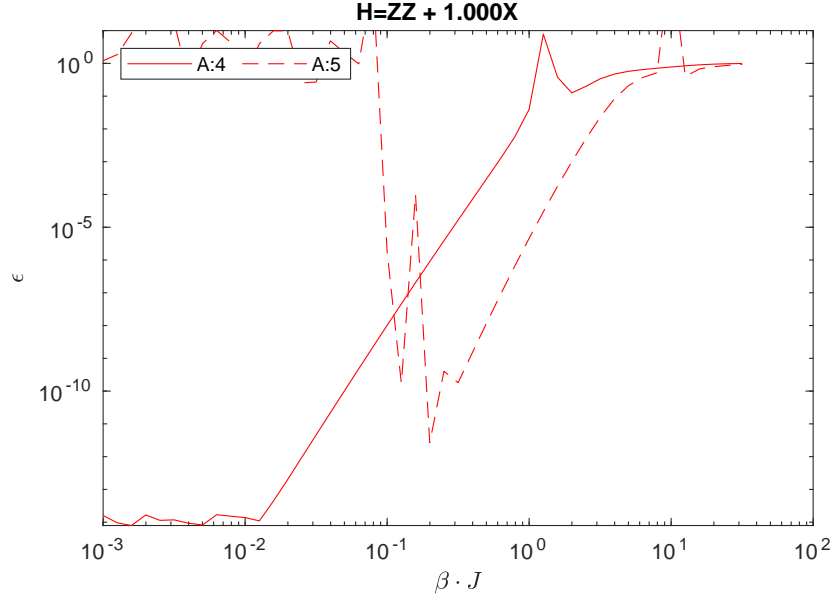


Figure 6.2: Error for transversal Ising, computed with full inversion

### 6.1.3 Models

With these basic definitions and findings out of the way, the different serie expansions can be tested against some different physical models.

#### 6.1.3.1 Ising

The results for the different types are all bundled in fig. 6.3. The vertical axis show the logarithm of the relative error  $\epsilon$ , horizontal axis the normalised inverse  $\beta = \frac{J}{T}$ . The most surprising finding is the fact that the strict type E performs worse than the others. For low  $\beta$ , it is more than 5 orders of magnitude worse. Taken together with its large bond dimension, this expansion and all other strict variants are seen to be of no use.

Now lets focus on the 2 other variants. Type A clearly outperforms type F for all  $\beta < 2$  by quite some margin. Only for large  $\beta$ , type F has the upper hand. While not completely clear in the picture, type A has quite a large error for some orders at  $\beta \in (2, 10)$ , but higher orders seems to solve the problem. The construction of type F requires twice the bond dimension, and hence type A performs clearly better overall

#### 6.1.3.2 Heisenberg

Now we focus on the spin 1/2 Heisenberg model on a chain. The results are again displayd in fig. 6.4. The exact type seems to perform a lot better here,



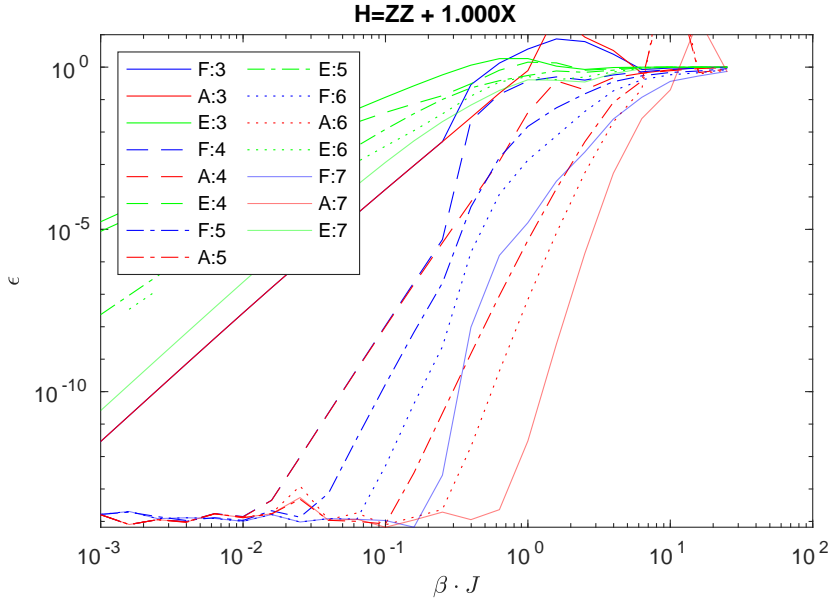


Figure 6.3: Comparison type A, E and F for Transversal Ising.

but still has consistently the largest error of the 3 methods for a given order. For low  $\beta$ , type A again performs better than type F. At  $\beta \approx 1$  and larger, F starts to improve upon the result of A, and doesn't have an divergent error.

### 6.1.3.3 Random

To give a representative overview for random hamiltonians, several simulations were run. The single site and nearest neighbour hamiltonians are generated by making hermitian matrices with random real and complex numbers between -1 and 1. In order to compare the different graphs, the engergy scale is set such that the norm of the hamiltonian evaluated on 2 sites is 1.

run random hamiltonian

### 6.1.4 Conclusion

- strict variant are useless
- full smmetry is not needed
- pseudoinverse neccesary
- Type A is the best
- never add new block if previous one was not fully solved (truncation)
- truncation adds possibility for extrapolation

write this

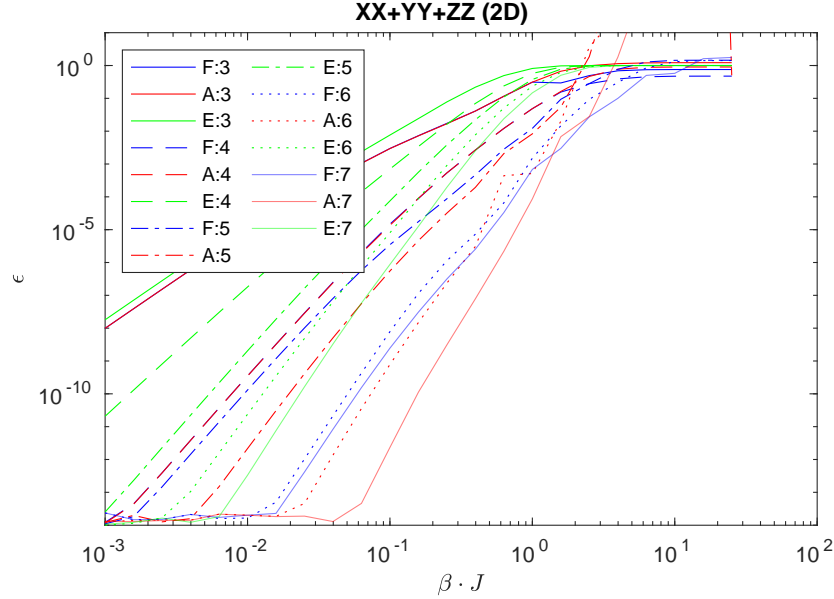


Figure 6.4: Comparison type A, E and F for Heisenberg model.

## 6.2 Results 2D

The results in 1D are very promising, but the real open challenges are situated in 2 (or higher) dimensions. Therefore, it is an interesting and relevant question whether the results generalise to 2D. These results are twofold. On the one hand, a similar measure as in 1D is calculated to measure how good a method works on a specific 2D grid. Here, we will need to settle for what can be computed. As a second measure, and from physical point of view the most interesting one, some phase transitions of the 2D transversal Ising model are determined, and compared against values in literature.

### 6.2.1 Direct results

#### 6.2.1.1 norm

Once again, a suitable norm has to be derived. The situation is more difficult than in 1D, because contracting a PEPO tensor network has a much larger computational complexity than in 1D. Ideally, the norm would be calculated on a  $n \times n$  cyclical grid (2D grid on a torus). Here  $n$  has to be at least 1 larger than the largest explicitly constructed chain.

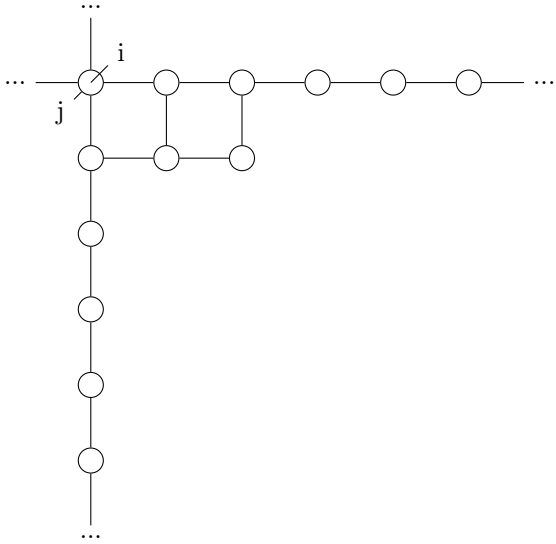
In practice, this is not achievable in reasonable amount of time. The limitations are twofold: the maximum number of sites to calculate the matrix exponential is still around 14, and the contraction of the tensor network is limited

number

to

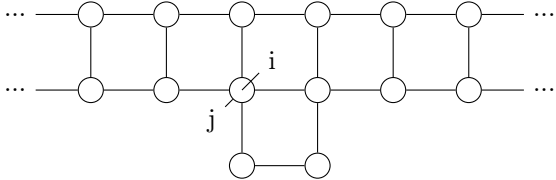
The limitations on the PEPO contraction can be somewhat relaxed by not computing the full network, but computing the reduced density matrix. Here, most of the sites have their physical indices traced out, except for one site.

The final error measures are defined by:



$$\rho_{i,j}^1 = \quad (6.3)$$

and



$$\rho_{i,j}^2 = \quad (6.4)$$

For the first one, all the blocks in the expansion are present in, and it is cyclical in both x and y. The second reduced density matrix is only cyclic in one direction, but includes more loop type contributions. Both relative errors will be used, defined by :

$$\epsilon^\alpha = \frac{\|\rho_{exact,i,j}^\alpha - \rho_{i,j}^\alpha\|_2}{\|\rho_{exact,i,j}^\alpha\|_2} \quad \alpha \in [1, 2] \quad (6.5)$$

For the given error measures, series expansion up till order 5 can be tested. The second norm takes the most time to compute, due to the larger complexity of contracting the tensor network.

### 6.2.1.2 Ising

We first determine the optimal value  $\sigma_0$  for pseudoinversion and test the influence of the addition of plaquette term to the expansions. The series is of order 5, and the transversal field is of magnitude 2.5. The results can be seen in ??.

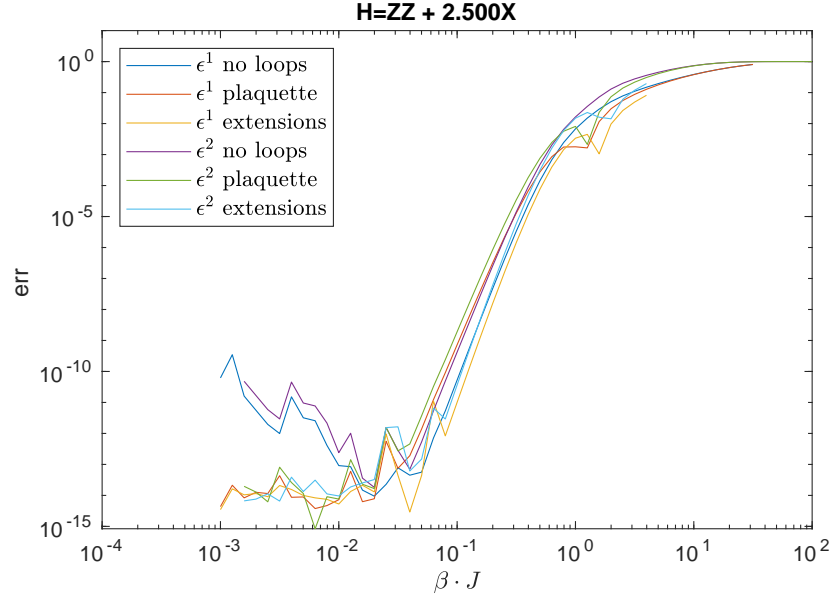


Figure 6.5

The truncation parameter  $\sigma_0$  was optimal at a value of about  $\sigma_0 = 10^{-12}$  in 1D. Smaller values resulted in divergent series expansions. In the graph we see that even better behaved inverses of  $\sigma_0 = 10^{-10}$  improve upon the error made.

The plaquette term (??) is mainly important at low  $\beta$  to keep the error low, and also has some considerable influence at higher temperatures.

fix caption

### 6.2.1.3 Heisenberg

### 6.2.1.4 Conclusion

## 6.2.2 Phase diagram 2D transversal Ising model

Now we have an idea how accurate the cluster expansions is, we can use it to calculate the thermodynamic properties of the transverse field Ising model. The sampling of points was explained in section 5.3

As a reference, fig. 3.2 is shown once more

### 6.2.2.1 classical Ising phase transition

The classical Ising model on a square lattice has an exact solution. Onsager calculated the critical temperature to be  $T_c = \frac{2J}{k \ln(1+\sqrt{2})} \approx 2.69185$ . A test for the construction is to simulate this and compare the fitted (see section 5.3.3) results.

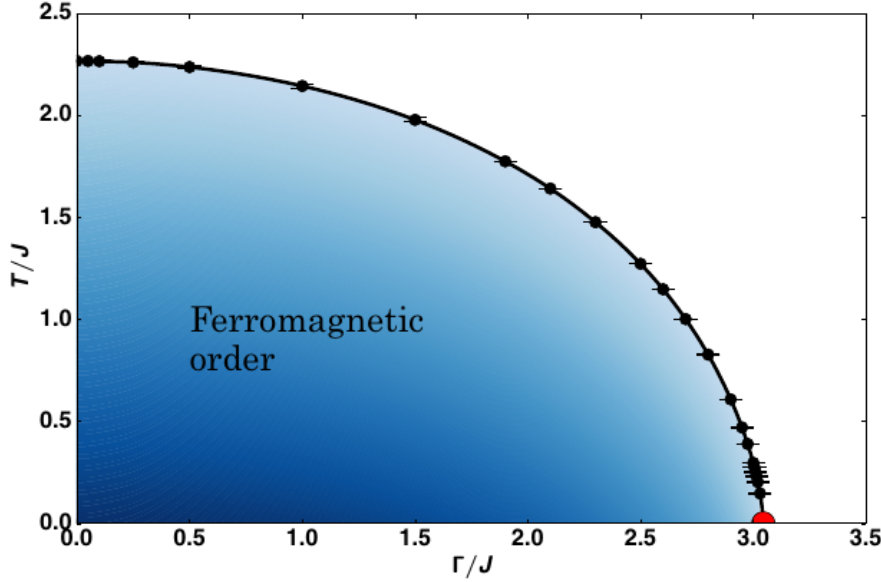


Figure 6.6: Phase diagram for 2D transversal Ising model. Figure taken from [17]

At the relevant temperature, an order 5 series expansion without loops is sufficient. This keeps the bond dimension small and hence the environment can be computed faster with VUMPS.

The results are shown in fig. 6.7. Each of the figures has 4 different subplots. Left upper corner shows  $m$  vs  $T$ . For low  $T$ , there is clearly a nonzero macroscopic magnetisation, while high  $T$  has a zero expectation value as expected. The other three plots show the data collapse of the finite size scaling of the entanglement entropy  $S$ , the correlation length  $\xi$  and the magnetisation  $m$ .  $\delta$  is the gap, a measure for the system size as explained in section 3.1.5.

Near criticality they collapse well, but away from the critical point there is quite some systematic variation. The reason is shown in fig. 3.1. Only in some limited range around the critical region, the universal scaling holds.

A more zoomed in version is shown in fig. 6.8. Clearly, the data collapses a lot better as expected. In fact, the data collapses so well we could determine the critical exponents and the temperature with the given data.

new figure with entropy formula fixed  $6S/c$  instead of  $cS/6$

provide numbers for fit

#### 6.2.2.2 $g=2.5$ phase transition

The previous example simulated the classical 2D Ising model. The question is whether this result will carry on into the quantum regime. The results for the full phase diagram fig. 6.9 and the points in the direct neighbourhood of the phase transition fig. 6.10 are similar to the 1D case.

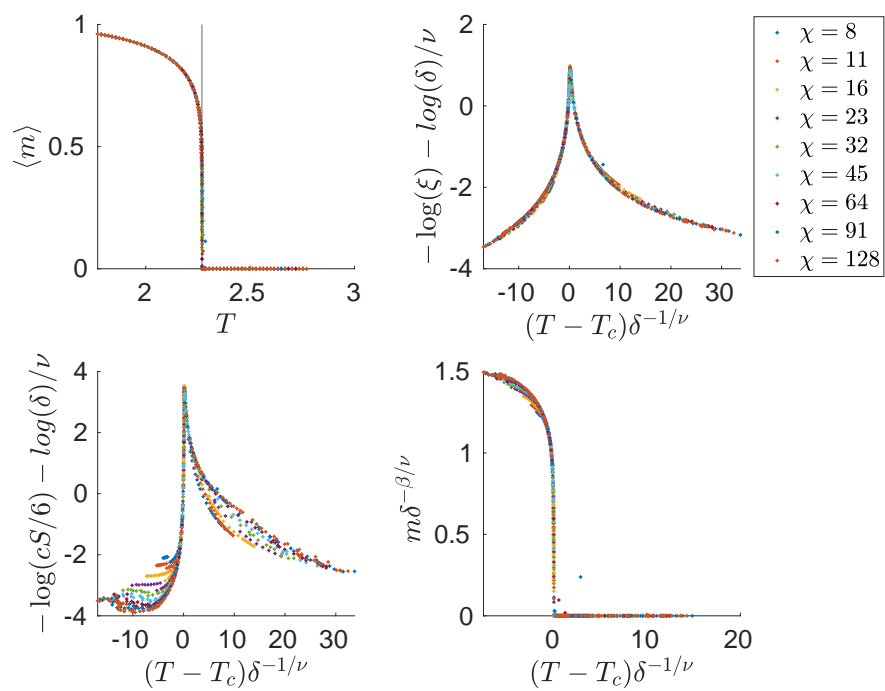


Figure 6.7

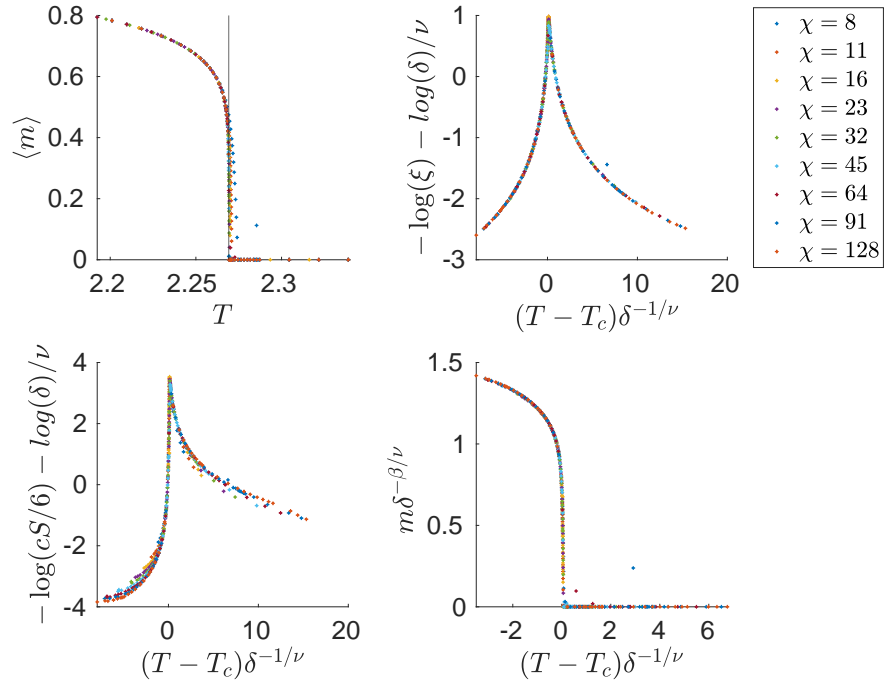


Figure 6.8

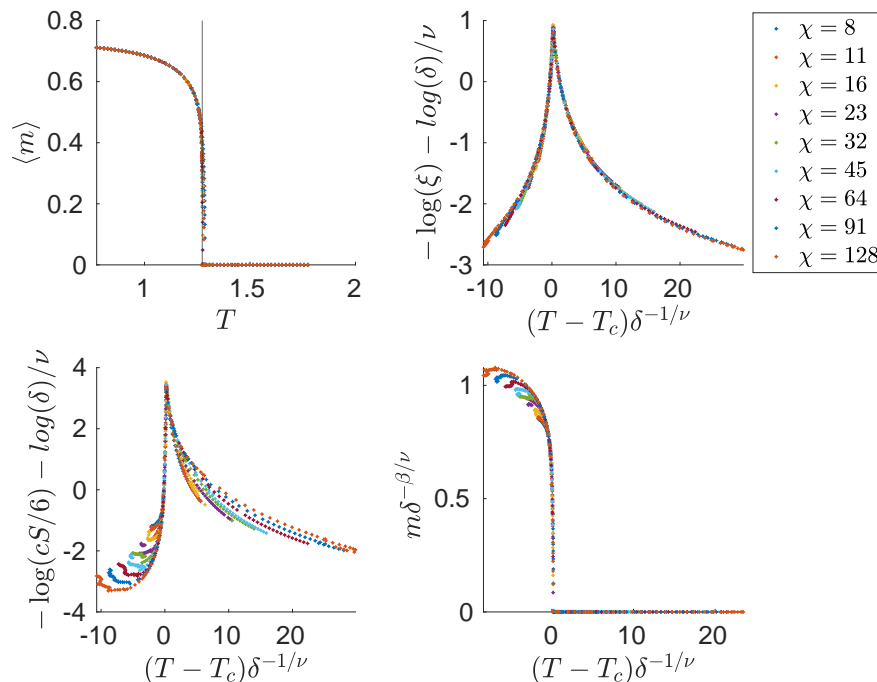


Figure 6.9

Of course, quantitative details are different: the phase transition happens at  $T \approx 1.274$  and the maximum magnetisation is lower than 1, in accordance to fig. 3.2. The variation of  $m$  around the critical temperature for different bond dimensions is much higher. A much larger  $\chi$  is needed for the fixed point MPS. This is not necessarily a problem, as this gives better data for the finite size scaling.

There is no analytical expression for critical temperature known. With quantum monte carlo techniques, a value of  $T_c = 1.2737(6)$  is obtained, while state of the art tensor network techniques provide a value of  $T_c = 1.2737(2)$  [24]

With the series expansion from this paper and Vumps,  $T_c = 1.2736(6)$ , indicating that this faithfully captures the physics. To put this into context: the directly calculated error  $\epsilon^2$  at  $T = 1.2$  was around 0.006.

### 6.2.2.3 $T=0.7$ quantum phase transition

Up until now, the phase diagram in fig. 6.6 was explored at constant  $g$  in function of  $T$ , but of course there is also a transition at constant  $T$  and varying  $g$ .  $T = 0.7$  is chosen. The results are shown in fig. 6.11.

Clearly, the results do not collapse as well as for the other models. The



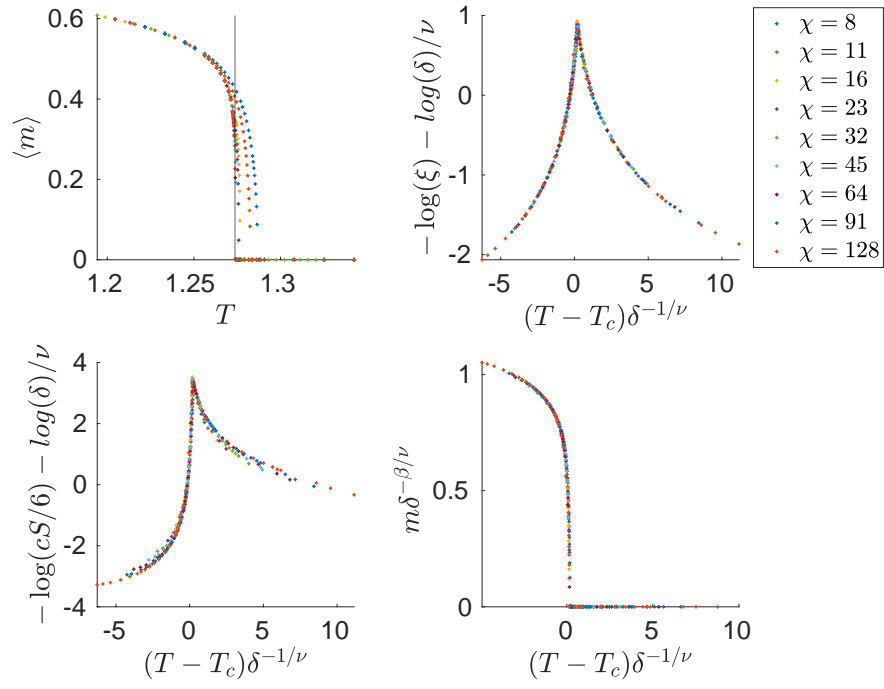


Figure 6.10

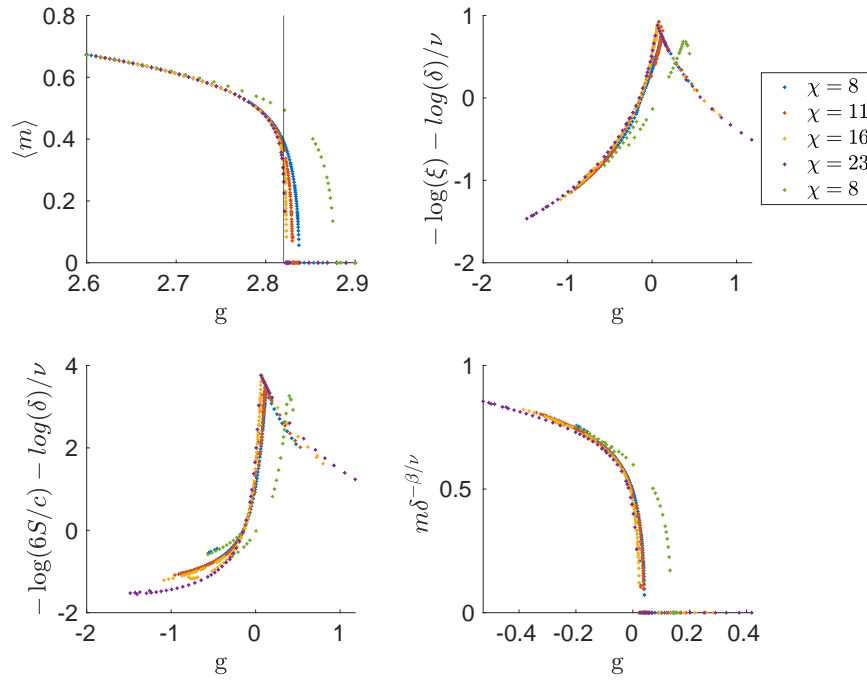


Figure 6.11: Results for  $T = 0.7$  phase transition. The green points are a truncated order 6 construction, the others order 5.

reason is that the order of the expansion is not high enough. The green curve is order 6, where virtual level is truncated to dimension 20. The others are order 5. The green and blue curve both have MPS bond dimension  $\chi = 8$ , but predict quite different magnetisation for large  $g$ . This shows that for  $g > 0.7$ , order 5 is not sufficient anymore.

#### 6.2.2.4 Tricritical point

Now that we know the critical transversal field can be determined, all the tools are present to extrapolate the Tricritical point, indicated by a red dot in fig. 6.6. To achieve this, the following scaling relation can be used:

$$T_c = |g_c - g_{c,q}|^{z\nu_{3D}} \quad (6.6)$$

Near the critical point, the critical temperature  $T_c$  and the critical transversal field  $g_c$  are related to each other by the value of the quantum critical point  $g_{c,q}$  and critical exponent  $z = 1$  and  $\nu_{3D} \approx 0.62998$  [17].

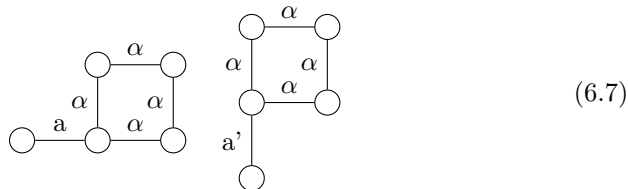
This could perfectly be done, given enough computation time to calculate for many temperatures the critical transversal field with a finite size scaling, similar to section 6.2.2.3. Possibly, also a scaling in truncation dimension for the cluster expansion is needed.

### 6.2.3 Going beyond

With all the built machinery to construct cluster expansions, is logical to calculate the phase diagram with increased precision.

#### 6.2.3.1 loops and extensions

Adding loops decreases the error. but there is also a surprising result: all possible loop extensions result in a higher error. The fluctuation increases drastically. With increasing bond dimension  $\chi$ , this is somewhat better but still not good enough. The question is whether this is a result of a failing cluster expansion, or inability VUMPS to calculate the correct environment. During my thesis, my focus has largely been on the first case. After all, the framework was completely build from scratch and errors happen, and there is no guarantee that the series even converges. But introduction of very strict variants, such as the generalization to type E, where left/upper extension have level  $a$  and the right/lower extensions are of type  $a'$ :

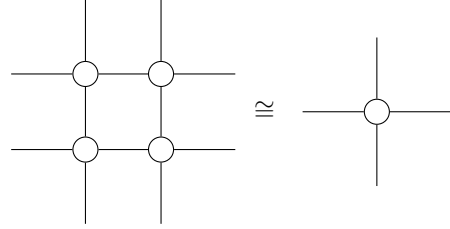


Introduce these variations.

The VUMPS procedure come with 2 implicit assumptions: the original version was derived for hermitian hamiltonians. The MPO resulting from the traced PEPO from the cluster expansion is not an hermitian MPO.

A second assumption made during this thesis is that the wave function can be represented by a 1by1 unit cell. Despite efforts made, the multisite version [4] doesn't seem to produce sensible results, even for version where 1by1 unit cell does give the right results.

It's clear that more research is needed here. One way to locate the problem would be to use another algorithm to contract the network, such as corner transfer matrix renormalization group (CTMRG) as used in citeCzarnik2019, or even using the single site VUMPS algorithm combined with blocking:



$$\approx \quad (6.8)$$

complex matrices instead of real ones??

### 6.2.3.2 Higher order

Going to higher order (i.e. longer linear chains) and adding the single loop contribution works well. The bond dimension of largest virtual level (3 in this case) can be truncated in the construction. The linear and non-linear solver find the least squares solution to the problems.

Care has to be taken in order to not violate one of the conclusions in 1D: never construct a longer chain than the previously fully solved one. For instance if the bond is truncated to 10, the linear chains can be constructed up till order 4. This means eq. (4.26) can be added but not eq. (4.27), because the longest chains are order 5. Level 5 can still be constructed and contracted easily, but virtual level 3 has a bond dimension of 64. Compared to the previous bond dimensions (1,4 and 16) this is quite large and needs to be truncated.

### 6.2.3.3 Better extrapolation

The spirit behind finite size scaling is to calculate more with the data available. In section 3.1.5, 2 additional variations were suggested. One is to account for the subleading finite size corrections, the other to change the way of calculating  $\delta$ .

**6.2.3.3.1 subleading corrections** Introducing subleading corrections requires 4 parameters for each fitted universal function:  $\omega, \phi, c$  and  $d$  from eq. (3.4). Compared to the one parameter  $T_c$  or  $g_c$ , this is a lot and can be used to make

many graphs collapse into a single graph. The analysis in [25] carefully estimates the error made with the fitting procedure. A similar amount of rigour would be required to fully trust the results.

Nevertheless, fitting with subleading corrections results in critical temperatures close to the ones fitted without, and the subleading exponents are close to 1, as expected from the subleading series expansion.

**6.2.3.3.2 Choice  $\delta$**  Different choices of  $c_i$  in eq. (3.3) are possible to construct  $\delta$ . Variational optimisation was done as suggested in [4]. Both the x and y axis should be normalised, because depending on  $\delta$  the scale changes. If scaled based on the outer points, there is a flaw in the fitting procedure. The variational minimum goes to a point where at least for one point  $\delta$  is extremely close to 0, and hence everything on the x axis except that point is pinched together, resulting in very small relative error. Therefore, it is better to normalise according to the points a e.g. 25th percentile and 75th percentile of the range on the axis.

This improves the collapse, but is not clear how much the prediction of  $T_c$  or  $g_c$  is improved.

## 6.3 Conclusion

The results in previous section show that this novel method can compete against some recent tensor network results in literature.



## Chapter 7

# Conclusion and lookout

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# Bibliography

- [1] R. Sinatra, P. Deville, M. Szell, D. Wang, A. L. Barabási, A century of physics (oct 2015). [arXiv:1612.00079](#), [doi:10.1038/nphys3494](#).
- [2] R. Orús, A practical introduction to tensor networks: Matrix product states and projected entangled pair states (oct 2014). [arXiv:1306.2164](#), [doi:10.1016/j.aop.2014.06.013](#).
- [3] P. Corboz, Lecture 1: tensor network states, Tech. rep.
- [4] A. Nietner, B. Vanhecke, F. Verstraete, J. Eisert, L. Vanderstraeten, Efficient variational contraction of two-dimensional tensor networks with a non-trivial unit cell, *Quantum* 4 (2020). [arXiv:2003.01142](#), [doi:10.22331/Q-2020-09-21-328](#).
- [5] M. Hauru, C. Delcamp, S. Mizera, Renormalization of tensor networks using graph independent local truncations, Tech. rep. [arXiv:1709.07460v2](#).
- [6] R. Orús, G. Vidal, Simulation of two-dimensional quantum systems on an infinite lattice revisited: Corner transfer matrix for tensor contraction [doi:10.1103/PhysRevB.80.094403](#).
- [7] L. Vanderstraeten, J. Haegeman, F. Verstraete, Tangent-space methods for uniform matrix product states, *SciPost Phys. Lect. Notes* 7 (2019). [doi:10.21468/SciPostPhysLectNotes.7](#).
- [8] H. Nishimori, G. Ortiz, *Elements of Phase Transitions and Critical Phenomena*, Vol. 9780199577, Oxford University Press, 2011. [doi:10.1093/acprof:oso/9780199577224.001.0001](#).
- [9] G. Jaeger, The ehrenfest classification of phase transitions: Introduction and evolution, *Archive for History of Exact Sciences* 53 (1) (1998) 51–81. [doi:10.1007/s004070050021](#).
- [10] A. J. Beekman, L. Rademaker, J. van Wezel, An Introduction to Spontaneous Symmetry Breaking (sep 2019). [arXiv:1909.01820](#), [doi:10.21468/scipostphyslectnotes.11](#).

- [11] B. Vanhecke, J. Haegeman, K. Van Acoleyen, L. Vanderstraeten, F. Verstraete, Scaling Hypothesis for Matrix Product States, *Physical Review Letters* 123 (25) (2019). [arXiv:1907.08603](#), [doi:10.1103/PhysRevLett.123.250604](#).
- [12] K. S. D. Beach, L. Wang, A. W. Sandvik, Data collapse in the critical region using finite-size scaling with subleading corrections (may 2005). [arXiv:0505194](#).  
URL <http://arxiv.org/abs/cond-mat/0505194>
- [13] P. Ginsparg, E. Brézin, J. Zinn-Justin, Applied Conformal Field Theory, Tech. rep. (1988).
- [14] P. Calabrese, J. Cardy, Entanglement entropy and quantum field theory, *Journal of Statistical Mechanics: Theory and Experiment* (6) (2004). [arXiv:0405152](#), [doi:10.1088/1742-5468/2004/06/P06002](#).
- [15] S. Sachdev, Quantum phase transitions, *Physics World* 12 (4) (1999) 33–38. [doi:10.1088/2058-7058/12/4/23](#).  
URL <https://iopscience.iop.org/article/10.1088/2058-7058/12/4/23>  
<https://iopscience.iop.org/article/10.1088/2058-7058/12/4/23/meta>
- [16] A. Taroni, Statistical physics: 90 years of the Ising model (dec 2015). [doi:10.1038/nphys3595](#).  
URL [www.nature.com/naturephysics](http://www.nature.com/naturephysics)
- [17] S. Hesselmann, S. Wessel, Thermal Ising transitions in the vicinity of two-dimensional quantum critical points, *PHYSICAL REVIEW B* 93 (2016) 155157. [doi:10.1103/PhysRevB.93.155157](#).
- [18] S. Paeckel, T. Köhler, A. Swoboda, S. R. Manmana, U. Schollwöck, C. Hubig, Time-evolution methods for matrix-product states, *Annals of Physics* 411 (2019). [arXiv:1901.05824](#), [doi:10.1016/j.aop.2019.167998](#).
- [19] M. P. Zaletel, R. S. K. Mong, C. Karrasch, J. E. Moore, F. Pollmann, Time-evolving a matrix product state with long-ranged interactions, *PHYSICAL REVIEW B* 91 (2015) 165112. [doi:10.1103/PhysRevB.91.165112](#).
- [20] T. H. Hsieh, From d-dimensional Quantum to d + 1-dimensional Classical Systems, *Student review* (2) (2015) 1–4.
- [21] B. Vanhecke, L. Vanderstraeten, F. Verstraete, Symmetric cluster expansions with tensor networks, *Physical Review A* 103 (2) (2021). [arXiv:1912.10512](#), [doi:10.1103/PhysRevA.103.L020402](#).
- [22] C. Moler, C. Van Loan, Nineteen dubious ways to compute the exponential of a matrix, twenty-five years later (2003). [doi:10.1137/S00361445024180](#).

- [23] A. Wietek, A. M. Läuchli, Sublattice coding algorithm and distributed memory parallelization for large-scale exact diagonalizations of quantum many-body systems, *Physical Review E* 98 (3) (2018). [arXiv:1804.05028](#), [doi:10.1103/PhysRevE.98.033309](#).
- [24] P. Czarnik, P. Corboz, Finite correlation length scaling with infinite projected entangled pair states at finite temperature, *Physical Review B* 99 (2019) 245107. [doi:10.1103/PhysRevB.99.245107](#).
- [25] L. Wang, K. S. Beach, A. W. Sandvik, High-precision finite-size scaling analysis of the quantum-critical point of  $S=1/2$  Heisenberg antiferromagnetic bilayers, *Physical Review B - Condensed Matter and Materials Physics* 73 (1) (2006). [arXiv:0509747](#), [doi:10.1103/PhysRevB.73.014431](#).