# FYS4480 ORAL EXAM, MIDTERM ONE AND TWO HELIUM AND BERYLLIUM USING CIS AND HARTREE-FOCK, PAIRING MODEL

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### **SETUP**

Represent states using creation  $a_p^{\dagger}$  and annihilation  $a_q$  operators (occupation representation/second quantization), obeying

$$\{a_p^{\dagger}, a_q\} = \delta_{pq}, \qquad \{a_p^{\dagger}, a_q^{\dagger}\} = \{a_p, a_q\} = 0$$

p and q are sets of relevant quantum numbers. We need to pick a single particle (SP) computational basis, having n possible single particle states.

3D HO: 
$$p = \{n_r, l, m_l, s, m_s\}, \quad a_p^{\dagger} |0\rangle = |p\rangle \longrightarrow \psi_p(\mathbf{x}) = \psi_{n_r l m}(r, \theta, \phi) = \dots$$

Need a Hamiltonian to solve  $\hat{H} = \hat{H}_0 + \hat{V}$ , with  $\hat{H}_0$  representing single particle energy contributions, and  $\hat{V}$  interactions.

Often start with an N-particle ground state ansatz  $|\Phi_0\rangle=a_1^\dagger,\dots a_N^\dagger\,|0\rangle$ .

And consider excitations of this

1p1h 
$$|\Phi_{i}^{a}\rangle = a_{a}^{\dagger}a_{i}|\Phi_{0}\rangle$$
  
2p2h  $|\Phi_{ij}^{ab}\rangle = a_{a}^{\dagger}a_{b}^{\dagger}a_{j}a_{i}|\Phi_{0}\rangle$   
NpNh  $|\Phi_{ij...}^{ab...}\rangle = a_{a}^{\dagger}a_{b}^{\dagger}\dots a_{j}a_{i}|\Phi_{0}\rangle$ 

# FULL CONFIGURATION INTERACTION (FCI)

Start with *N* particle ground state ansatz  $|\Phi_0\rangle$ .

Not an eigenstate of  $\hat{V}$  and therefor not the true ground state of the system.

By considering every possible N particle state in our system (using  $\{a_1^{\dagger}, \dots, a_N^{\dagger}, \dots, a_n^{\dagger}\}$ ), we can construct our ground state  $|\Psi_0\rangle$  as a linear combination of excited states.

$$|\Psi_0\rangle = C_0 + \sum_{ai} C_i^a |\Phi_i^a\rangle + \sum_{ai} C_{ij}^{ab} |\Phi_{ij}^{ab}\rangle + \dots$$

Normally solved by considering the Hamiltonian in matrix representation, with elements  $H_{XY} = \langle \Phi_X | \hat{H} | \Phi_Y \rangle$ , with  $X, Y \in \{0p0h, 1p1h \dots NpNh\}$ , giving the eigenvalue problem

$$H\mathbf{c} = E\mathbf{c}$$

When solved, the smallest eigenvalue  $E^{(0)}$  will yield the ground state energy, and  $|\Psi_0\rangle$  can be found by considering the eigenvectors  $\mathbf{c}^{(0)} = (C_0, C_i^a \dots C_{ij}^{ab} \dots C_{ij}^{ab} \dots C_{ij}^{ab} \dots$ ). Excited states can also be found by considering the other eigenvalues and vectors  $E^{(i)}, \mathbf{c}^{(i)}$ .

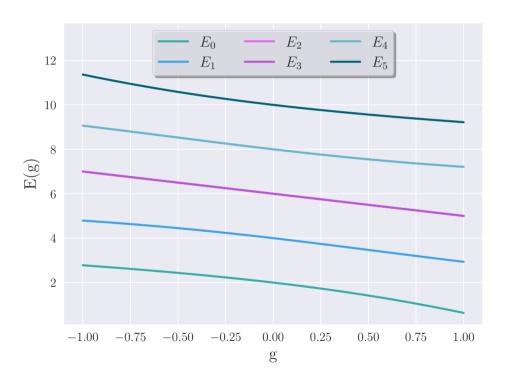
Example, pairing interaction:  $\hat{V}$  only works between spin-paired states at the same energy level. We let p=1,2,3,4 denote energy levels, with SP states also having a spin  $\sigma=\pm$ , a total of n=8 SP states.

$$\hat{H} = \hat{H}_0 + \hat{V}, \qquad \hat{H}_0 = \sum_{p\sigma} (p-1)a^{\dagger}_{p\sigma}a_{p\sigma}, \qquad \hat{V} = -\frac{1}{2}g\sum_{pq}a^{\dagger}_{p+}a^{\dagger}_{p-}a_{q-}a_{q+}$$

Diagonal SP Hamiltonian  $\hat{H}_0$  and two body interaction  $\hat{V}$ . Consider N=4 particles with total spin S=0, writing  $|PQ\rangle=|p+p-q+q-\rangle$  we have a total of six different many body states.

0p0h 
$$|12\rangle$$
  
2p2h  $|13\rangle$ ,  $|14\rangle$ ,  $|23\rangle$ ,  $|24\rangle$   
4p4h  $|34\rangle$ 

By setting up the matrix  $\langle KL | \hat{H} | RS \rangle$ , we get a small eigenvalue problem of a 6 × 6 matrix.



In all but very simple problems, this approach is unfeasible. In general, we have to consider

$$\binom{n}{N} = \frac{n!}{N!(n-N)!}$$

many body states. Taking our pairing model example, lifting the S=0 restriction yields 70 different states. This is still possible, but increasing both n and N results in disaster

$N\downarrow/n\to$	8	32	64	128
4	70	$10^{4}$	$10^{5}$	$10^{7}$
8		$10^{7}$	$10^{9}$	$10^{12}$
16		$10^{8}$	$10^{14}$	$10^{19}$
32			$10^{18}$	$10^{30}$

Table. NB: Order of magnitude values

#### **FCI**

### Pros:

- ▶ Provides exact solutions within a truncated basis set
- Understandable and relatively easy to set up
- ► Excited states thrown into the bargain

### Cons:

- Computational complexity, bad scaling
- ▶ Only possible for tiny systems, with few states and particles.
- ► Practically only a benchmarking tool

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section title.

# **CUSTOM TITLE**

**CUSTOM SUBSECTION WITH FOOTNOTE** 

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<sup>&</sup>lt;sup>1</sup>This is a footnote. See also Author (2022).

#### **TYPOGRAPHICS**

# These examples follow the Metropolis Theme

- ► Regular
- ► Alert
- ► Italic
- ► Bold

#### LISTS

#### **Items**

- ► Cats
  - British Shorthair
- ► Dogs
- ► Birds

### **Enumerations**

- 1. First
  - 1.1 First subpoint
- 2. Second
- 3. Last

# **Descriptions**

Apples Yes

Oranges No

Grappes No

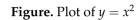
### TABLE

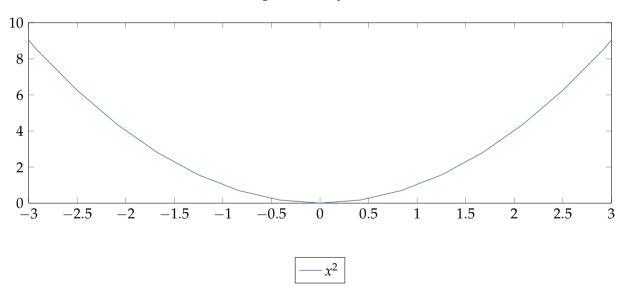
**Table.** Largest cities in the world (source: Wikipedia)

City	Population
Mexico City	20,116,842
Shanghai	19,210,000
Peking	15,796,450
Istanbul	14,160,467

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Mexico City	20,116,842
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# **FIGURES**





### **BLOCKS**

# **Default**

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# **Alert**

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# Example

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# MATHS

### **EQUATIONS**

► A numbered equation:

$$y_t = \beta x_t + \varepsilon_t \tag{1}$$

► Another equation:

$$\mathbf{Y} = \boldsymbol{\beta}\mathbf{X} + \boldsymbol{\varepsilon}_t$$



**THEOREM** 

▶ Theorems are numbered consecutively.

# **Theorem 1 (Example Theorem)**

Given a discrete random variable X, which takes values in the alphabet  $\mathcal{X}$  and is distributed according to  $p: \mathcal{X} \to [0,1]$ :

$$H(X) := -\sum_{x \in \mathcal{X}} p(x) \log p(x) = \mathbb{E}[-\log p(X)]$$
 (2)

### **M**ATHS

#### **DEFINITIONS**

▶ Definition numbers are prefixed by the section number in the respective part.

# **Definition 1.1 (Example Definition)**

Given a discrete random variable X, which takes values in the alphabet  $\mathcal{X}$  and is distributed according to  $p: \mathcal{X} \to [0,1]$ :

$$H(X) := -\sum_{x \in \mathcal{X}} p(x) \log p(x) = \mathbb{E}[-\log p(X)]$$
(3)

# MATHS EXAMPLES

Examples are numbered as definitions.

### **Example 1.1 (Example Theorem)**

Given a discrete random variable X, which takes values in the alphabet  $\mathcal{X}$  and is distributed according to  $p: \mathcal{X} \to [0,1]$ :

$$H(X) := -\sum_{x \in \mathcal{X}} p(x) \log p(x) = \mathbb{E}[-\log p(X)]$$
(4)

# Part I

# **DEMO PRESENTATION PART 2**

# REFERENCES I

Author, Example (2022). "Reference Title". In: *Journal of Examples* 0.0, pp. 1–10.