Electronic Excited States from a Variance-Based Contracted Quantum Eigensolver

August 22, 2023

This is the demonstration jupy ter notebook of "Electronic Excited States from a Variance-Based Contracted Quantum Eigensolver" by Yuchen Wang and David A. Mazziotti.

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[1]: #import qiskit, scipy, numpy
from qiskit import QuantumCircuit,QuantumRegister,ClassicalRegister
from qiskit import Aer,execute
import numpy as np
import itertools
from scipy.linalg import expm
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[2]: #Pauli matrices
     sigma_x = np.array([[0, 1], [1, 0]], dtype=complex)
     sigma_y = np.array([[0, -1j], [1j, 0]], dtype=complex)
     sigma_z = np.array([[1, 0], [0, -1]], dtype=complex)
     iden = np.array([[1, 0], [0, 1]], dtype=complex)
     #initialize quantum circuit
     def init_qc(N):
         q = QuantumRegister(N)
         c = ClassicalRegister(N)
         qc = QuantumCircuit(q,c)
         return q,c,qc
     #apply tensor product of a list of matrices from right to left
     def tensors(list_of_mats):
         num_mats = len(list_of_mats)
         tensor_product = list_of_mats[0]
         i = 0
         while i+1 < num_mats:</pre>
             tensor_product = np.kron(tensor_product,list_of_mats[i+1])
         return tensor_product
     #define the tensor product from Pauli string
     def str2mat(pstr):
         matrix_list = []
         for char in pstr:
             if char == 'I':
                 matrix_list.append(iden)
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elif char == 'X':
            matrix_list.append(sigma_x)
        elif char == 'Y':
            matrix_list.append(sigma_y)
        elif char == 'Z':
            matrix_list.append(sigma_z)
    return tensors(matrix list)
#this returns a dictionary of Pauli String from any unitary operator
def HS(m, n):
    return (np.dot(m.conjugate().transpose(), n)).trace()
def decomposeNqubit(H, N=4):
    pmat = [iden, sigma_x, sigma_y, sigma_z]
    char = ['I', 'X', 'Y', 'Z']
    char_list = [''.join(i) for i in itertools.product(char, repeat = N)]
    pmat_list = [i for i in itertools.product(pmat, repeat = N)]
    decomposed_dict = {}
    for i in range(len(char_list)):
        a_{ij} = 1/(2**N) * HS(tensors(pmat_list[i]), H)
        if np.abs(a_ij) > 0.00000001:
            decomposed_dict[char_list[i]] = a_ij
    return decomposed_dict
```

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[4]: #binary dictionary of Pauli string
     def z to binary(z string, N=4):
         labels = ['0', '1']
         b string = [''.join(i) for i in itertools.product(labels, repeat = N)]
         tensor list = []
         b_dict = {}
         for i in z_string:
             if i == 'Z':
                 tensor_list.append(np.real(sigma_z))
             elif i == 'I':
                 tensor_list.append(np.real(iden))
         matrix = tensors(tensor_list)
         for i in range(len(b_string)):
             b_dict[b_string[i]] = matrix[i, i]
         return b dict
     #get z-basis string from statevector/qasm simulator
     def measure_zbasis(qc, N=4):
         result = execute(qc, Aer.get_backend('statevector_simulator')).result()
         statevector = result.get_statevector()
         #this sums up all binary string
         b_coeff = np.zeros(2**N)
         b_label = ['0', '1']
         b_string = [''.join(i) for i in itertools.product(b_label, repeat = N)]
         for i in range(len(np.asarray(statevector))):
             b_coeff[i] = np.abs(statevector[i])**2
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#this sums up all z basis string
    z_{coeff} = np.zeros([2**N]) #z_{coeff} is what we want to calculate
    z_label = ['Z', 'I']
    z string = [''.join(i) for i in itertools.product(z label, repeat = N)]
    for k in (range(len(z_string))):
        dict_binary = z_to_binary(z_string[k])
        for j in range(len(b_string)):
            if b_string[j] in dict_binary:
                z_coeff[k] += b_coeff[j] * dict_binary[b_string[j]]
    return z_coeff, z_string
#change the basis to get X, Y value
def measure_abasis(qc, string, N=4):
    for pos,i in enumerate(string):
        if i == 'X' :
            qc.h(pos) #convert qubit to X
        elif i == 'Y' :
            qc.z(pos) #convert qubit to Y
            qc.s(pos)
            qc.h(pos)
    a_coeff, a_string = measure_zbasis(qc)
    for pos,i in enumerate(string):
        if i == 'X' or i == 'Y':
            for j in range(len(a_string)):
                if a string[j][pos] == 'Z':
                    a_string[j] = a_string[j][:pos]+i+a_string[j][(pos+1):]
   return a_coeff, a_string
#measure the value w.r.t given dict
def measure_from_dict(qc, m_dict):
    e = 0
    for item in m_dict:
        qc\_copy = qc.copy()
        a_coeff, a_string = measure_abasis(qc_copy, item)
        for i in range(len(a_string)):
            if item == a_string[i]:
                e += m_dict[item] * a_coeff[i]
    return e
#this applys pauli string to circuits
def apply_pauli_string(qc, pauli):
    for q, i in enumerate(pauli):
        if i == "X":
            qc.x(q)
        elif i == "Y":
            qc.rz(-np.pi / 2, q)
            qc.x(q)
            qc.rz(np.pi / 2, q)
        elif i == "Z":
            qc.rz(np.pi, q)
```

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def pauli_string(qc, loc, sigma="x", inv=False):
    if sigma in ["Z", "z"]:
        pass
    elif sigma in ["X", "x"]:
        qc.h(q[loc])
    elif sigma in ["I", "i"]:
        pass
    elif sigma in ["Y", "y"]: #
        if inv: \# Y \rightarrow Z
            qc.h(q[loc])
            qc.s(q[loc])
        else: #Z \rightarrow Y
            qc.sdg(q[loc])
            qc.h(q[loc])
#this gadget performs exp(1j*val*pauli) over wavefunction
def pauli_gadget(qc, val, pauli):
    s, c = pauli, val
    scaling = -2.0
    if len(s) == 1:
        if s == "I":
            pass
        elif s == "X":
            qc.rx(val * scaling, q[0])
        elif s == "Y":
            qc.ry(val * scaling, q[0])
        elif s == "Z":
            qc.rz(val * scaling, q[0])
    else:
        pauliTerms = 0
        ind = []
        terms = []
        for n, i in enumerate(s):
            if not i in ["I"]:
                pauliTerms += 1
                ind.append(n)
                terms.append(i)
        if pauliTerms == 0:
            pass
        else:
            # basis
            for n, p in zip(ind, terms):
                pauli_string(qc, n, p)
            # exp cnot
            for n in range(0, pauliTerms - 1):
                qc.cx(q[ind[n]], q[ind[n + 1]])
            # parameter
            qc.rz(val * scaling, q[ind[-1]])
```

```
# exp cnot
            for n in reversed(range(pauliTerms - 1)):
                qc.cx(q[ind[n]], q[ind[n + 1]])
            # inv. basis
            for n, p in zip(ind, terms):
                pauli_string(qc, n, p, inv=True)
#We import an example of the Jordan-Wigner mapped dictionary of H2 Hamiltonian
→ and verify the HF solution
p_dict_H2 = {"IIII": -0.53393635, "ZIII": 0.06727930, "IZII": 0.00665130, __
→"IIZI": 0.06727930,
"IIIZ": 0.00665130, "ZZII": 0.06501570, "ZIZI": 0.12736570, "XXXX": 0.06478462,
"YYXX": 0.06478462, "XXYY": 0.06478462, "YYYYY": 0.06478462, "ZIIZ": 0.12980031,
"IZZI": 0.12980031, "IZIZ": 0.13366603, "IIZZ": 0.06501570}
for p_str in ["IYZX","ZYXI","XZYI"]:
   q,c,qc = init_qc(4)
   apply_pauli_string(qc, p_str)
   print (measure_from_dict(qc, p_dict_H2))
```

- -0.78379264
- -0.6653988599999999
- -0.5412806400000001

```
[5]: #Jordan-Wigner mapping
     def JW_anni(pos, N=4):
         tensor_list = []
         for i in range(N):
              if i<pos :</pre>
                   tensor_list.append(sigma_z)
              if i==pos :
                   tensor_list.append((sigma_x+1j*sigma_y)/2)
              if i>pos :
                   tensor_list.append(iden)
              anni matrix = tensors(tensor list)
         return anni_matrix
     def JW_crea(pos, N=4):
         tensor_list = []
         for i in range(N):
              if i<pos :</pre>
                   tensor_list.append(sigma_z)
              if i==pos :
                   tensor_list.append((sigma_x-1j*sigma_y)/2)
              if i>pos :
                   tensor_list.append(iden)
              crea_matrix = tensors(tensor_list)
         return crea_matrix
     #the 2-rdm tensor is stored as a {\displaystyle \frac{a}{a}} = \frac{1}{a} - \frac{1}{a} = \frac{1}{a} = \frac{1}{a}
     #the operator is appllied in the order of k, l, j, i
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def rdmtomo(i,j,k,l):
    tomo = JW_anni(k)
    tomo = np.matmul(JW_anni(1), tomo)
    tomo = np.matmul(JW_crea(j), tomo)
    tomo = np.matmul(JW_crea(i), tomo)
    return tomo
#now we define the rdm decompose and sum
def tomo_decompose(N=4):
    tomo string real = []
    tomo_string_imag = []
    for i in range(N):
        for j in range(N):
            for k in range(N):
                for 1 in range(N):
                    tomo_dict = decomposeNqubit(rdmtomo(i,j,k,1), 4)
                    for item in tomo_dict:
                        if np.isreal(tomo_dict[item]):
                            if item not in tomo_string_real:
                                tomo_string_real.append(item)
                        else:
                            if item not in tomo_string_imag:
                                tomo_string_imag.append(item)
    return tomo_string_real, tomo_string_imag
def tomo sum(mat, N=4):
    new_mat = np.zeros([2**N, 2**N], dtype = complex)
    for i in range(N):
        for j in range(N):
            for k in range(N):
                for 1 in range(N):
                    new_mat += mat[4*i+j,4*k+l] * rdmtomo(i,j,k,l)
    return new_mat
#this block measures the fermionic rdm from given circuit
def rdm2(qc, N=4):
    tomo_string_real, tomo_string_imag = tomo_decompose()
    tomo_string = tomo_string_real + tomo_string_imag
    tomo_values = []
    for string_to_measure in tomo_string:
        qc\_copy = qc.copy()
        value list measured, string list measured = measure abasis(qc copy,
 →string_to_measure)
        #loop over the list
        for i in range(len(string_list_measured)):
            if string_list_measured[i] == string_to_measure:
                tomo_values.append(value_list_measured[i])
    rdm = np.zeros([2**N, 2**N], dtype = complex)
    for i in range(N):
        for j in range(N):
```

```
[6]: #now we define the variance matrices
     def var2_mat(p_dict, para, N=4):
         emat = np.zeros([2**N, 2**N])
         for item in p_dict:
             emat += p_dict[item] * str2mat(item).real
         vmat = emat - para*np.identity(16)
         v2mat = np.matmul(vmat, vmat)
         return v2mat
     def propagate(qc, mat_dict):
         qc\_copy = qc.copy()
         for item in mat_dict:
             pauli_gadget(qc_copy, mat_dict[item].real, item)
         rdm = rdm2(qc_copy)
         return rdm
     def antihermitian(M):
         D = 0.5 * (M + np.conj(M.T))
         C = 0.5 * (M - np.conj(M.T))
         return C
```

```
[9]: q,c,qc = init_qc(4)
     apply_pauli_string(qc, "ZYXI")
     j = 0
     para e = -0.5
     \#para_e = -0.7
     while j < 10:
         print ("at iter ", j)
         para_e_iter = measure_from_dict(qc, p_dict_H2)
         print ("energy is ", para_e_iter)
         v2mat = var2_mat(p_dict_H2, para_e)
         v2_dict = decomposeNqubit(v2mat)
         var = measure_from_dict(qc, v2_dict)
         print ("variance is", var)
         rdmf = propagate(qc, v2_dict)
         v2_dict_inv = {}
         for i in v2_dict:
             v2_dict_inv[i] = -v2_dict[i]
         rdmb = propagate(qc, v2_dict_inv)
```

```
F2_op = antihermitian(tomo_sum((rdmf - rdmb)/2j))
    qc.unitary(expm(F2_{op}),[3,2,1,0])
    v2mat_iter = var2_mat(p_dict_H2, para_e_iter)
    v2_dict_iter = decomposeNqubit(v2mat_iter)
    var_iter = measure_from_dict(qc, v2_dict_iter)
    if var_iter < var:</pre>
        para_e = para_e_iter
        var = var_iter
    j+=1
at iter 0
energy is -0.6653988599999999
variance is (0.09450953470601+0j)
at iter 1
energy is -0.5019743373780265
variance is (0.06715275181671043+0j)
at iter 2
energy is -0.5019743373780265
variance is (0.040445177222490504+0j)
at iter 3
energy is -0.42337420747142573
variance is (0.014754799806467864+0j)
at iter 4
energy is -0.4082830074536329
variance is (0.0012719345040423519+0j)
at iter 5
energy is -0.40626383655782394
variance is (5.868493439864508e-06+0j)
at iter 6
energy is -0.4062632884474563
variance is (1.5073731472886598e-06+0j)
at iter 7
energy is -0.40626346376863487
variance is (1.5982368628680854e-06+0j)
at iter 8
energy is -0.40626345294860766
variance is (1.5926290196366377e-06+0j)
at iter 9
energy is -0.40626345363440985
variance is (1.5929844509321445e-06+0j)
```

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