Quantum Transport in Nanoporous Graphene Bachelor defense

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Outline

Introduction

Project aim and nanoporous graphene

Tight Binding

 π -orbitals, π -electrons and the TB approximation

The Hamiltonian

Onsites, hops and the full TB Hamiltonian

Green's functions and recursion

Green's matrix, recursion and LDOS

Transmission

Device Green's functions, left/right geometry, rate matrices and spectral functions.

Transmission in 2D

Exploring GNR bridges

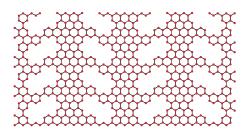
Para-O₄-NPG, Para-(OH)₄-NPG, Meta-O₂-NPG, Meta-(OH)₂-NPG

Project aim

"Development of tight-binding routines in Python in order to understand electron transport in novel nanoporous graphene devices (NPGs)"

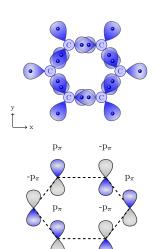
Nanoporous graphene

- Planar graphene sheets
- Periodically removed atoms
- Ribbons and bridges
- Ballistic electron movement
- Potential for controlling currents on nanoscale



π -orbitals and π -electrons

- In plane electrons are bound
- ightharpoonup 1 p_z -electron per site
- "Tightly bound" hops between sites



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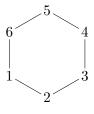
TB approximation

- Electrons tightly bound to sites
- Hops with potential
- Average electron energy on site
- The Hamiltonian is a hop matrix

$$V_{pp\pi} = \langle \phi_{\pi}(m) | \hat{\mathbf{H}} | \phi_{\pi}(n) \rangle$$
$$\epsilon_{0} = \langle \phi_{\pi}(i) | \hat{\mathbf{H}} | \phi_{\pi}(i) \rangle$$

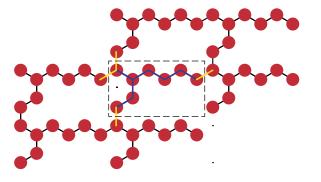
Hamiltonian for benzene

$$\mathbf{H} = V_{pp\pi} \begin{pmatrix} 1 & 2 & 3 & 4 & 5 & 6 \\ 1 & 0 & 1 & 0 & 0 & 0 & 1 \\ 2 & 0 & 1 & 0 & 0 & 0 & 0 \\ 3 & 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 1 \end{pmatrix}$$



Creating the first Hamiltonian

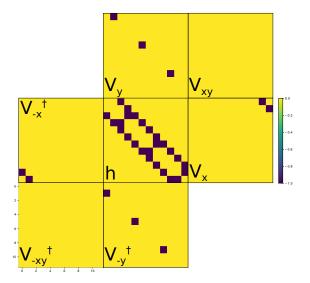
- Atom coordinates
- Interatomic distances
- Applying potential



```
h = np.zeros((xyz.shape[0], xyz.shape[0])) # Empty matrix
33
        for i in range(xyz.shape[0]): # Take an atomic coordinate
34
            for j in range(xyz.shape[0]): # Take another atomic coordinate
35
                h[i, j] = LA.norm(np.subtract(xyz[i], xyz[j])) # Measure distances
36
        h = np.where(h < 1.6, Vppi, 0) # Replace distances under 1.6 angstrom with Vppi
37
        h = np.subtract(h, Vppi * np.identity(xyz.shape[0])) # Remove the diagonal
38
```

Hopping matrices

- Shift by lattice vector
- Resulting matrices: $\mathbf{h_0}, \mathbf{V}, \mathbf{V}^\dagger$



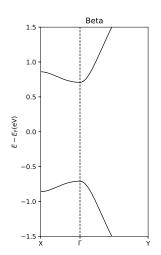
Full Hamiltonian and first band plots

$$\begin{aligned} \mathbf{H}(k_x, k_y) \boldsymbol{\phi}_k &= \boldsymbol{\epsilon}_n(k_x, k_y) \boldsymbol{\phi}_k \\ \mathbf{H}(k_x, k_y) &= \mathbf{h}_0 + (\mathbf{V}_x e^{-ik_x} + \mathbf{V}_x^{\dagger} e^{ik_x} + \mathbf{V}_y e^{-ik_y} + \mathbf{V}_y^{\dagger} e^{ik_y} \\ &+ \mathbf{V}_{xy} e^{-ik_x} e^{-ik_y} + \mathbf{V}_{xy}^{\dagger} e^{ik_x} e^{ik_y}) \end{aligned}$$

```
Ham = Ham + (V1 * np.exp(-1.0j * x)) # Onsite hops and hops in x
73
                     + np.transpose(V1) * np.exp(1.0j * x) # Hops in -x
74
                     + V2 * np.exp(-1.0j * y) # Hops in y
75
                     + np.transpose(V2) * np.exp(1.0j * y) # Hops in -y
76
                     + V3 * np.exp(-1.0j * x) * np.exp(-1.0j * y) # Hops in both x and y
77
                     + np.transpose(V3) * np.exp(1.0j * x) * np.exp(1.0j * y)) # Hops in both -x and -y
78
        e = LA.eigh(Ham)[0] # Eigen energies from the Hamiltonian
79
        v = LA.eigh(Ham)[1] # Eigen vectors from the Hamiltonian
80
```

Full Hamiltonian and first band plots

X:
$$\mathbf{H}_{X} = \mathbf{h}_{0} + (\mathbf{V}_{x}e^{ik_{x}} + \mathbf{V}_{x}^{\dagger}e^{-ik_{x}} + \mathbf{V}_{y} + \mathbf{V}_{y}^{\dagger} + \mathbf{V}_{xy}e^{ik_{x}} + \mathbf{V}_{xy}^{\dagger}e^{-ik_{x}})$$
Y:
$$\mathbf{H}_{Y} = \mathbf{h}_{0} + (\mathbf{V}_{x} + \mathbf{V}_{x}^{\dagger} + \mathbf{V}_{y}e^{-ik_{y}} + \mathbf{V}_{y}^{\dagger}e^{ik_{y}} + \mathbf{V}_{xy}e^{-ik_{y}} + \mathbf{V}_{xy}^{\dagger}e^{ik_{y}})$$



Green's matrix

- Solution to the Scödinger Equation
- Propagator

$$[(E+i\eta)\mathbf{1} - \mathbf{H}]\mathbf{G}(E) = \mathbf{1}$$

$$\downarrow$$

$$\mathbf{G}(E) = \mathbf{1}([(E+i\eta)\mathbf{1} - \mathbf{H}])^{-1}$$

Recursion

Semi-infinite chain

$$\begin{pmatrix} z\mathbf{1} - \mathbf{H}_c & -\mathbf{V}^{\dagger} \\ -\mathbf{V} & (z - \varepsilon')\mathbf{1} \end{pmatrix} \begin{pmatrix} \mathbf{X} & \mathbf{G}_{0c} \\ \mathbf{G}_{c0} & \mathbf{G}_{00} \end{pmatrix} = \begin{pmatrix} \mathbf{1} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} \end{pmatrix}$$

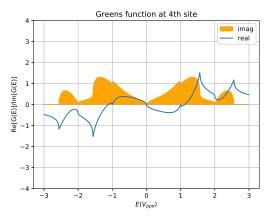
$$\mathbf{G}_{00}(z) = \left[(z - \varepsilon') - \mathbf{V}(z\mathbf{1} - \mathbf{H}_c)\mathbf{V}^{\dagger} \right]^{-1}$$

$$= (z - \varepsilon' - \Sigma(z))^{-1}$$
where
$$z = E + i\eta$$

$$a_0 = \mathbf{V}^{\dagger}, \quad b_0 = \mathbf{V}$$
 $e_{s0} = \mathbf{h}_s, \quad e_0 = \mathbf{h}$
in loop:
 $a_1 = a_0 \times g_0 \times a_0$
 $b_1 = b_0 \times g_0 \times b_0$
 $e_1 = e_0 + a_0 \times g_0 \times b_0 + b_0 \times g_0 \times a_0$
 $e_{1s} = e_{0s} + a_0 \times g_0 \times b_0$
 $g_1 = (z - e_1)^{-1}$
 $\mathbf{\Sigma}_R = e_s - h$
 $\mathbf{\Sigma}_L = e - h - \mathbf{\Sigma}_R$
 $\mathbf{G00} = (z - e_s)^{-1}$

Recursion

```
92
         while np.max(np.abs(a0)) > 1e-6: # Loop with hop matrix as threshold
             ag = a0 0 g0 # Product defined here and used multiple times
93
             a1 = ag 0 a0 # New hop matrix (transposed)
94
             bg = b0 0 g0 # Product defined here and used multiple times
95
             b1 = bg 0 b0 # New hop matrix
96
             e1 = e0 + ag 0 b0 + bg 0 a0 # New onsite for "other cells"
97
             es1 = es0 + ag 0 b0 # New onsite
98
             g1 = LA.inv(z - e1) # New Green's function
99
             a0 = a1 # Overwrite old variable
100
101
             b0 = b1 # Overwrite old variable
             e0 = e1 # Overwrite old variable
102
             es0 = es1 # Overwrite old variable
103
             g0 = g1 # Overwrite old variable
104
             q = q + 1 # Counter (for diagnostic purposes)
105
         e, es = e0, es0 # Define the onsite Hamiltonians
106
         SelfER = es - h # Self-energy from the right
107
         SelfEL = e - h - SelfER # Self-energy from the left
108
         GOO = LA.inv(z - es) # Green's functions
109
```

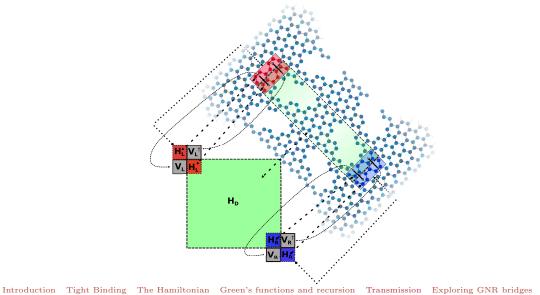


```
G00 = np.zeros((En.shape[0]), dtype=complex) # Empty data matrix for Green's functions
64
    for i in range(En.shape[0]): # Loop iterating over energies
65
        G, SelfER, SelfEL = RecursionRoutine(En[i], h, V, eta) # Invoking the RecursionRoutine
66
        G = np.diag(G) # The Green's functions for each site is in the diagonal of the G matrix
67
        GOO[i] = G[4] # Chosen Green's function (here the 4th site)
68
```

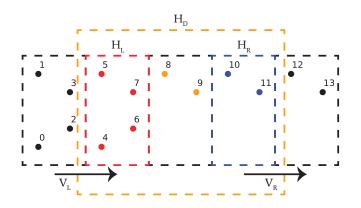
Transmission

Transmission is the probability of an electron being transported through a specific region for a specific range of energies.

Translating from system to matrices



The left and right self-energy



```
210
         for i in En: # Iteration over each energy
             gl, scrap, SEL = RecursionRoutine(i, HL, VL, eta=eta) # From the left
211
212
             gr, SER, scrap = RecursionRoutine(i, HR, VR, eta=eta) # From the right
```

Getting Transmission

$$\mathbf{G}_{D} = \left[\mathbf{1}(E + i\eta) - \mathbf{H}_{D} - \mathbf{\Sigma}_{L}(E) - \mathbf{\Sigma}_{R}(E)\right]^{-1}$$

$$\mathbf{\Gamma}_{L,R} = i\left(\mathbf{Sigma}_{L,R} - \mathbf{\Sigma}_{L,R}^{\dagger}\right)$$

```
GD["GD[:d]".format(q)] = scp.linalg.inv( # Device Green's functions are saved
scp.identity(HD.shape[0]) * (i + eta) - HD - SEL - SER) # in a dictionary.

GammaL["GammaL[:d]".format(q)] = 1j * (SEL - SEL.conj().transpose()) # Rate matrices are
GammaR["GammaR[:d]".format(q)] = 1j * (SER - SER.conj().transpose()) # likewise saved.
```

Getting Transmission

$$T(E) = \text{Tr}\Big[\mathbf{\Gamma}_R \mathbf{G}_D \mathbf{\Gamma}_L \mathbf{G}^{\dagger}\Big](E)$$

```
for i in range(En.shape[0]): # Iteration for every energy point.

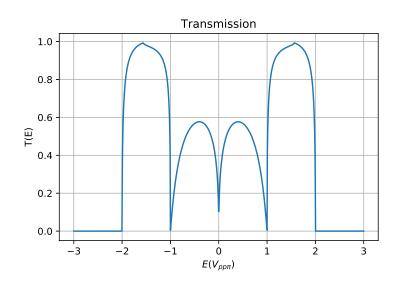
T[i] = np.trace((GammaR["GammaR{:d}".format(i)] @ GD["GD{:d}".format(i)]

GammaL["GammaL{:d}".format(i)] @ GD["GD{:d}".format(i)].conj(interaction of the conjection of the conjecti
```

Getting Transmission



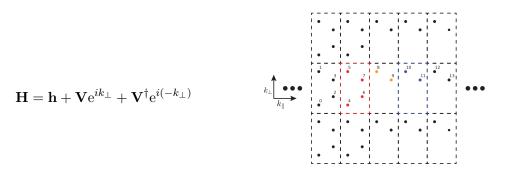
- Text2
- ► Text3



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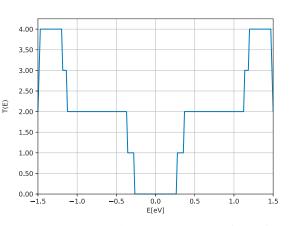
Transmission in 2D

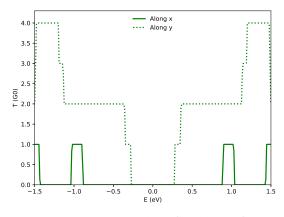


```
h, p = Onsite(xyz=xyz, Vppi=-1, f=1) # Calculate onsite hops
250
         V = Hop(xyz=xyz, xyz1=xyz + np.array([0, UY, 0]), Vppi=-1) # Calculate hops
251
         print('Number of hopping elements: {}'.format(np.sum(np.abs(V)))) # Number of hops
252
         Ham = h + V * np.exp(1j * i) + np.transpose(V) * np.exp(-1j * i) # Hamiltonian for ith kp
253
```

Code validity

$$k = \frac{\pi}{2}$$





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The Hamiltonian

Green's functions and recursion 000

Transmission

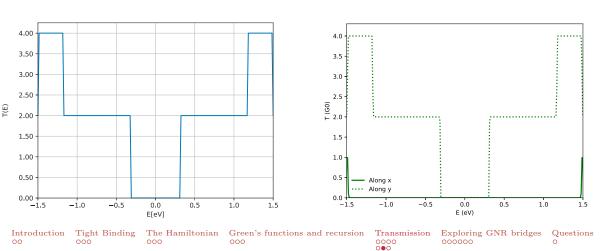
Exploring GNR bridges 000000

Questions

Code validity

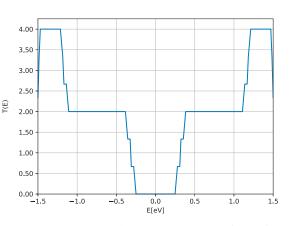
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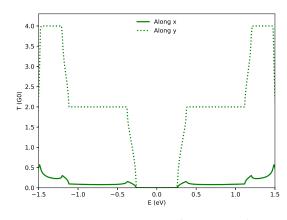
$$k = \pi$$



Code validity

$$k = AVG\left(0, \frac{\pi}{2}, \pi\right)$$





Introduction Tight Binding

The Hamiltonian

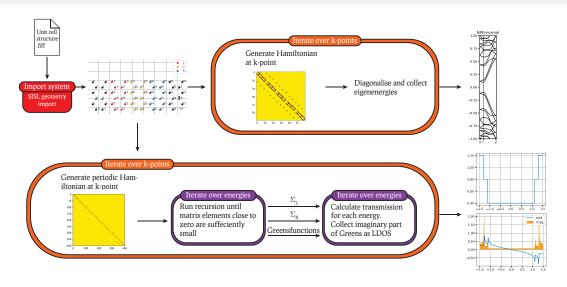
Green's functions and recursion 000

Transmission

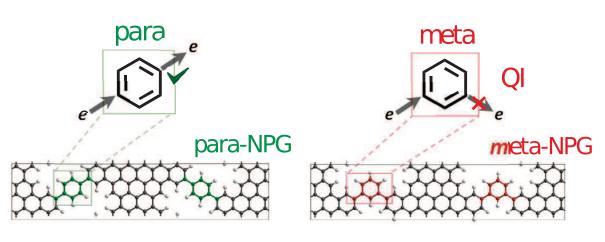
Exploring GNR bridges 000000

Questions

Summary of code structure

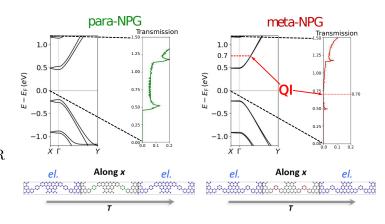


Para and meta bridges

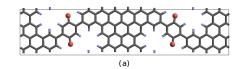


Para and meta bridges

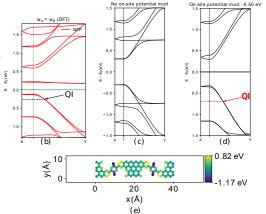
- Path length difference
- Quantum interference
- Transmission
- coupling/decoupling of GNR



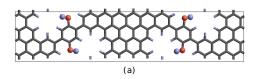
Para-O₄-NPG



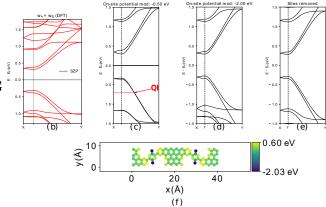
- Functionalisation with oxygen
- Quantum inteference at valence band
- Reproducing DFT result
- Decoupling of GNR
- Bands at Fermi level



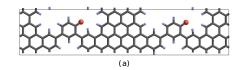
Para-(OH)₄-NPG



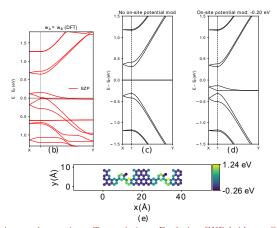
- ► Hydrogenation of oxygen
- ▶ Band splitting of DFT
- ► GNR coupling and resemblance with Para-NPG
- ► Reproducing DFT results



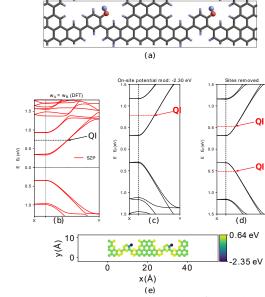
Meta-O₂-NPG



- Split in the valence for DFT
- Similarity in valence/conduction difference for meta/para
- Trying different potentials
- Flat bands at Fermi level



Meta-(OH)₂-NPG



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The Hamiltonian Green's functions and recursion 000

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Questions ŏ

Conclusion

"Development of tight-binding routines in Python in order to understand electron transport in novel nanoporous graphene devices (NPGs)"

Questions

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Energy and velocity fits

► Appendix