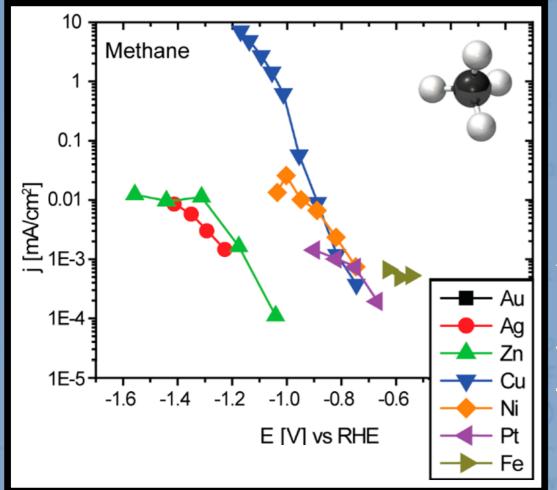
# Using Evolution to Discover Carbon Dioxide Reducing Catalysts

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## Background

Emitted anthropogenic carbon dioxide (CO<sub>2</sub>) must be minimised as it significantly contributes to the greenhouse effect. A partial solution to apply to industrial processes is to capture and convert discharged CO<sub>2</sub> to other products. Copper (Cu) surfaces have shown promise to

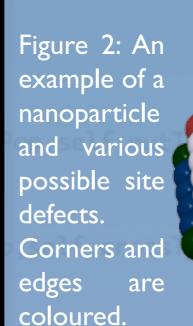


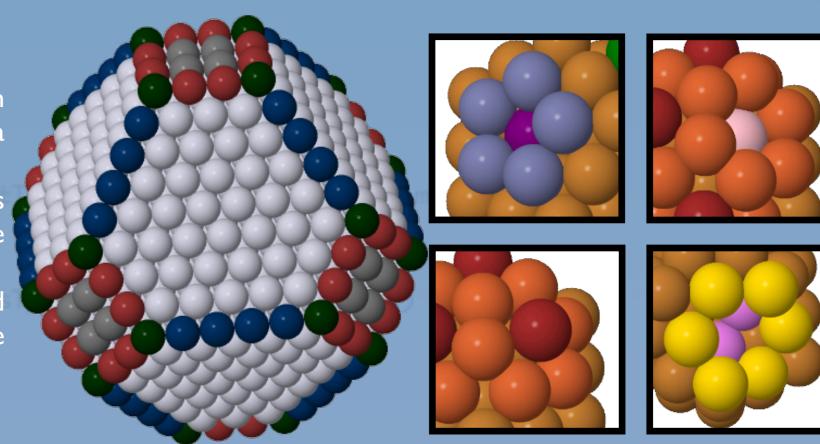
the selectively reduction into hydrocarbons and alcohols; significantly more than other metals (example in Figure 1) However, the performance of the catalyst must still be widespread use. A nanoparticle form of Cu offers such a catalytic pathway improvement.

Figure 1: Current densities (j) for the reduction of CO<sub>2</sub> to methane for a variety of transition metal surfaces across different applied potentials (E). K. P. Kuhl et al., J. Am. Chem. Soc., 2014, 136 (40), pp 14107-14113

### Nanoparticles and their Structure

Nanoparticles have more rich surface features, such as corners, edges, and defects of these sites, compared to bulk surfaces (Figure 2). These can give nanoparticles interesting catalytic properties.





Before the catalytic properties of a nanoparticle can be explored, it is necessary to obtain the stable structures of the nanoparticle that would exist experimentally.

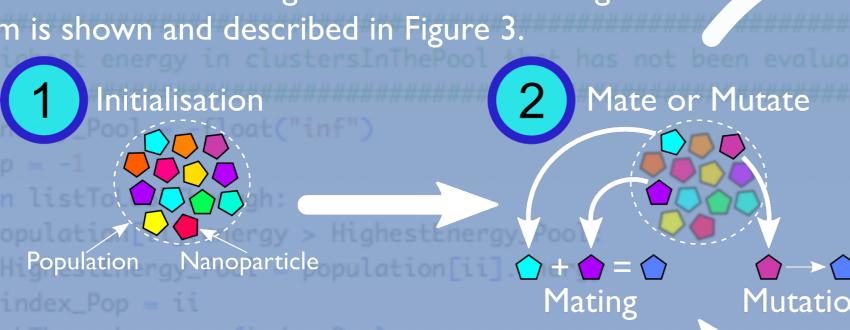
"Natural Selection for Generation " + str(generation\_number)

ain a list of clusters in order of lowest to highest energy for clusters

# Method: The Genetic Algorithm

Natural Selection

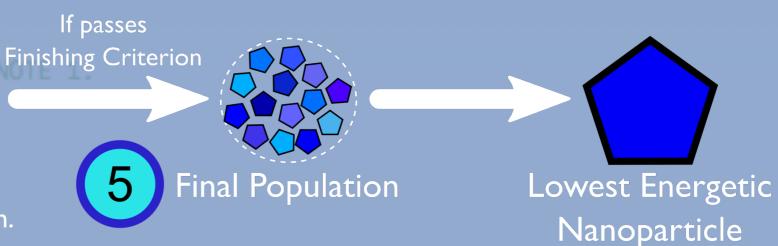
The genetic algorithm is a global optimisation technique based on Darwin's theory of inherited traits. The genetic algorithm rearranges atoms within a nanoparticle in search of the lowest energetic structure. The genetic algorithm is shown and described in Figure 3.



- population of randomly constructed nanoparticles is created.
- Offspring nanoparticles are formed by mixing two parent nanoparticles (mating) modifying an existing nanoparticle (mutation).
- Mutation Criterion Finishing Criterion

Figure 3: Diagram of the genetic algorithm.

- 3. The fitness of both existing and offspring nanoparticles are assessed. Here, fitter nanoparticles are those with lower energies. Less fit individuals in the population are replaced by fitter offspring.
- The population evolves through many generations by repeating steps 2 and 3 until a finishing criterion is met.
- 5. The genetic algorithm ends and the lowest energetic nanoparticle structure is obtained from the population.

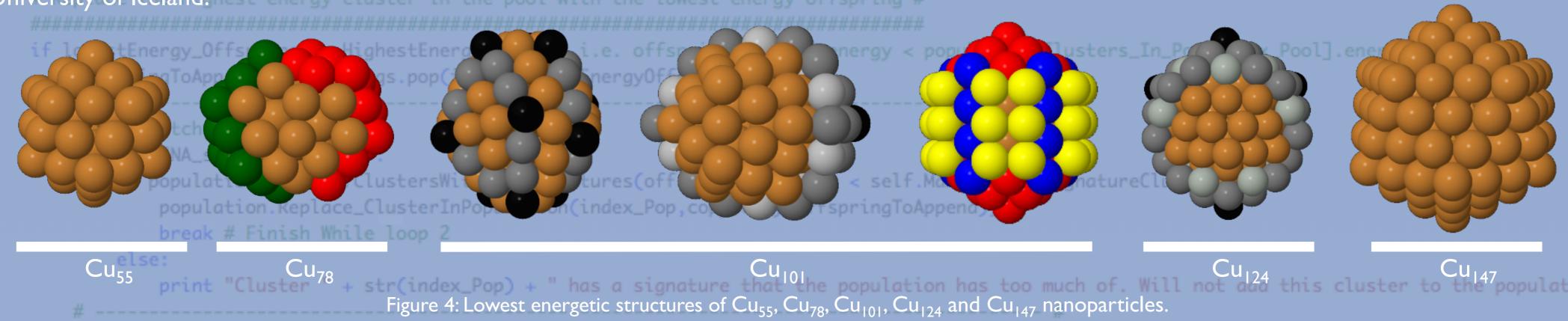


# find the next lowest energetic cluster in the population.

#### lowestEnergy Offsprings - float("inf")

# Results and Future Directions

The lowest energetic structures of Cu<sub>55</sub>, Cu<sub>78</sub>, Cu<sub>101</sub>, Cu<sub>124</sub>, and Cu<sub>147</sub> were obtained using the genetic algorithm (Figure 4). These structures contain interesting surfaces features, such as the twisted icosahedron structure exhibited by Cu<sub>78</sub> and the intersection of the shell and the core seen in Cu<sub>101</sub> middle) and Cu<sub>124</sub>, that could exhibit potential catalytic properties. Catalytic studies of these structures are currently being performed at the Jniversity of Iceland, hest energy cluster in the pool with the lowest energy offspring #



population.Replace\_ClusterInPopulation(index\_Pop,copy.copy(offspringToAppend))

Acknowledgments: Thanks to Dr Anna Garden and the Garden gnomes. Acknowledgements to the University of Otago and the Marsden Fund for providing PhD funding, the New Zealand eScience Infrastructure for providing computational facilities and support, and the MacDiarmid Institute for their supportes in offsprings have a lower energy than clusters in the curren









EnakeplacingClustersInroolwltnUffspringProcess = True # kequirea to br break # Break While Loop 2