

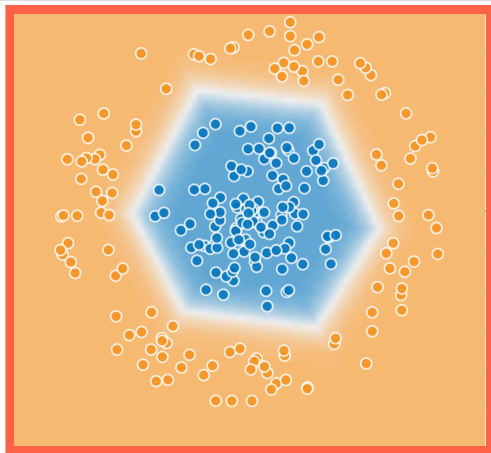
## What is...machine learning in mathematics - part 6?

---

Or: Approximating hard problems

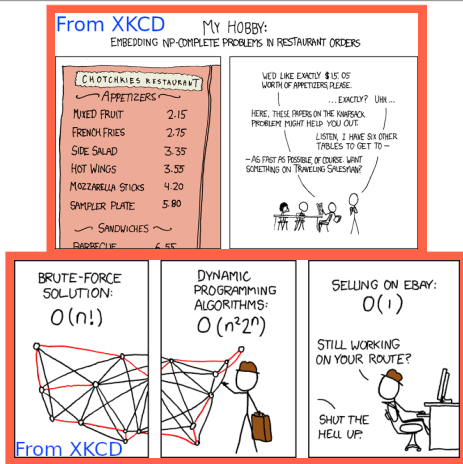
## Neural network (NN) = approximation

---



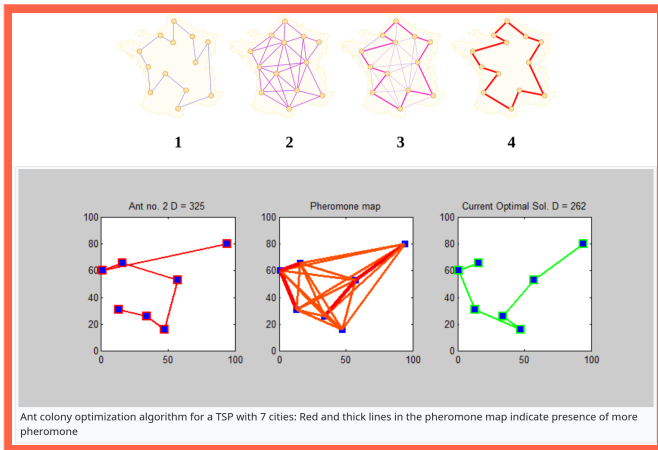
- 
- ▶ Above A NN finds a polygon approximation of a circle
  - ▶ Essentially a NN is a way to approximate a potentially complicated function
  - ▶ A NN approximates by combining simple functions into a complex model

# Hard problems



- ▶ For this video define hard to be 'Any algorithm for computation is nasty'
- ▶ Example NP-complete problems '=' not solvable in polynomial runtime
- ▶ Hard problems often require heuristic or approximate solutions

# Traveling salesperson problem (TSP)



- ▶ **TSP** Find the shortest route visiting each city exactly once and returning home
- ▶ **This is hard** Fastest known algorithm needs  $\approx 2^{\#cities}$  operations
- ▶ **Idea** Use NN to approximate a near-optimal solution efficiently

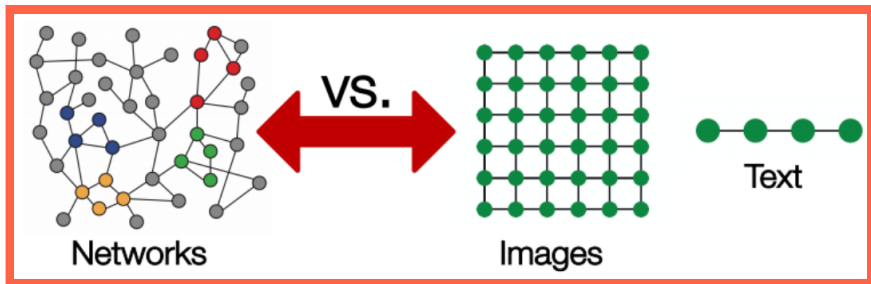
## Enter, the theorem

A 2018 GNN (<https://arxiv.org/abs/1809.02721>) can predict the TSP tour with

high probability and low derivation

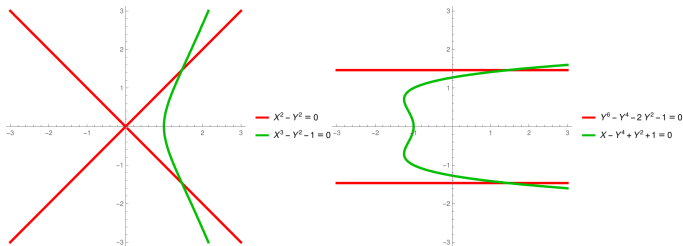
Roughly, 80% probability on a 2% derivation interval

- ▶ **Very impressive** This is close to results heuristics for TSP obtained by 'traditional methods', e.g. multi-fragment algorithms
- ▶ **Graph NN (GNN)** is a generalization of a convolutional NN (CNN)



# Promising!

## The same intersection set in two different ways



Question. How can we algebraically see that the intersections match?

$$(X^2 - Y^2, X^3 - Y^2 - 1) \stackrel{?}{=} (Y^6 - Y^4 - 2Y^2 - 1, X - Y^4 + Y^2 + 1)$$

- ▶ There are actually many other examples along the same lines
- ▶ Explicitly Computation of Gröbner bases is EXPSPACE-hard (doubly exponential in the worst case)
- ▶ Great A NN can do this efficiently, too!

**Thank you for your attention!**

---

I hope that was of some help.