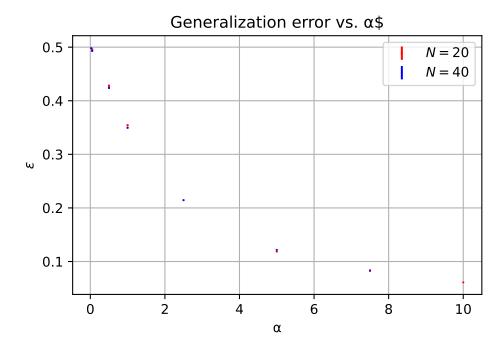
Exercise 04: The annealed solution to randomised perceptron learning

We now take the numerical results of Exercise 3 and compare them to the analytical behaviour that we would expect. In particular, we are studying the behaviour of the generalization error ϵ as $\alpha = P/N$ grows.

The analytic expression is the following

$$\alpha = \frac{(1 - \epsilon)\pi}{\tan(\pi \epsilon)}$$

The data given by the previous exercise are plotted here.

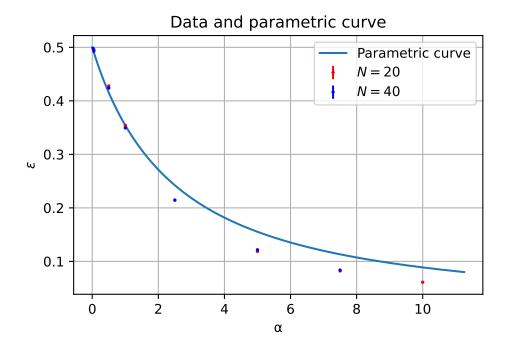


Step 1: Plot the parametric curve

We want parametrize $\alpha = \alpha(t)$ and $\epsilon = \epsilon(t)$ with parameter $t \in [0, 0.5]$. Let's take

$$\alpha(t) = \frac{(1-t)\pi}{\tan(\pi t)}$$

and $\epsilon(t) = t$.



Step 2: Consider a strategy to solve iteratively the eqn for fixed α

Let us devise an iterative algorithm in order to find a solution for a certain α .

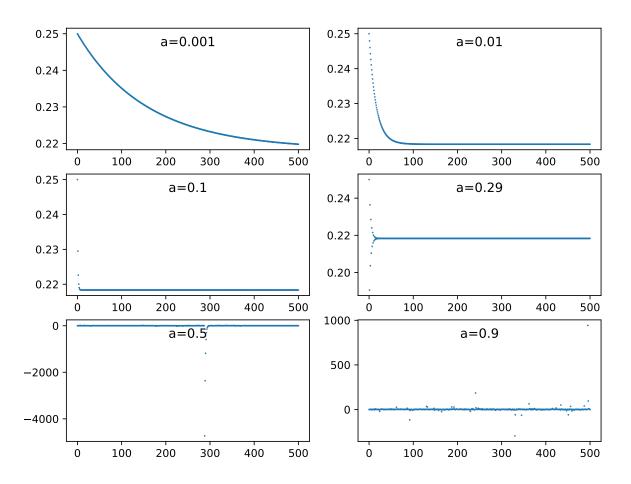
Say, for example, that $\alpha = 5$. Given a self-consistent function of the form $\epsilon = f(\epsilon, a)$, we can:

- Set an initial value $\epsilon_0 = 0.25$, for example, which is a good value in the codomain that we expect.
- Evaluate iteratively $\epsilon_{i+1} = (1-a)\epsilon_i + af(\epsilon_i,a)$
- Plot how ϵ_i changes with i to check for convergence at different settings of a.

Here we have chosen, for example:

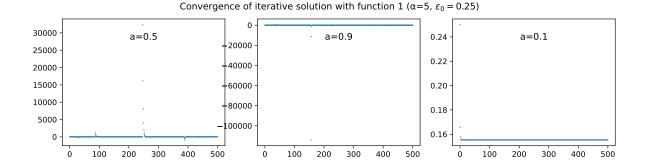
$$f_2(\epsilon,\alpha) = 1 - \frac{\alpha \tan(\pi \epsilon)}{\pi}$$

Convergence of iterative solution (α =5, ϵ_0 = 0.25)



Step 3: Check that a possible choice is $f=1-\alpha\tan(\pi\epsilon)/\pi$

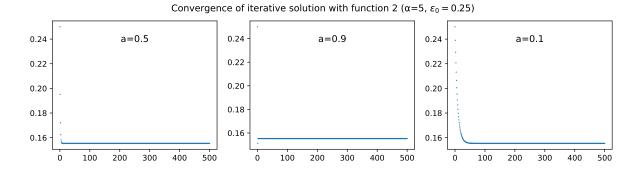
This is similar to the previous case, apart from the choice of α . Let's check for the specific plots we are asked for.



Step 4: Check that a possible choice is $f = \arctan(\pi(1-\epsilon)/\alpha)/\pi$

Let's see what happens with the following choice of the function, and the same parameters.

$$f_1(\epsilon,\alpha) = \frac{1}{\pi}\arctan(\frac{\pi(1-\epsilon)}{\alpha})$$



Step 5 and 6: How do you explain the results?

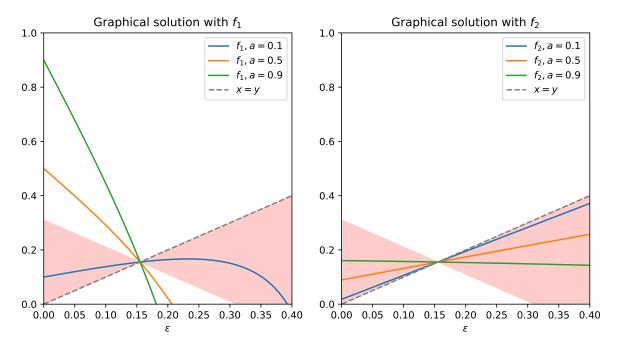
We see that in both cases the iterative algorithm reaches convergence, but not for all values of a. For f_1 , for example, we reach convergence only for the lowest value (a = 1). Meanwhile in the f_2 case we reach convergence in all the three cases. So we deem f_2 to be a better candidate when looking for the solutions of the equation.

We know from the **Banach theorem** that we have guaranteed convergence over a recursive function x = G(x) towards a fixed point if the fixed point x^* exists, and if the absolute value of the derivative of the function is such that $\left|\frac{dG(x)}{dx}\right|_{x=x^*} < 1$.

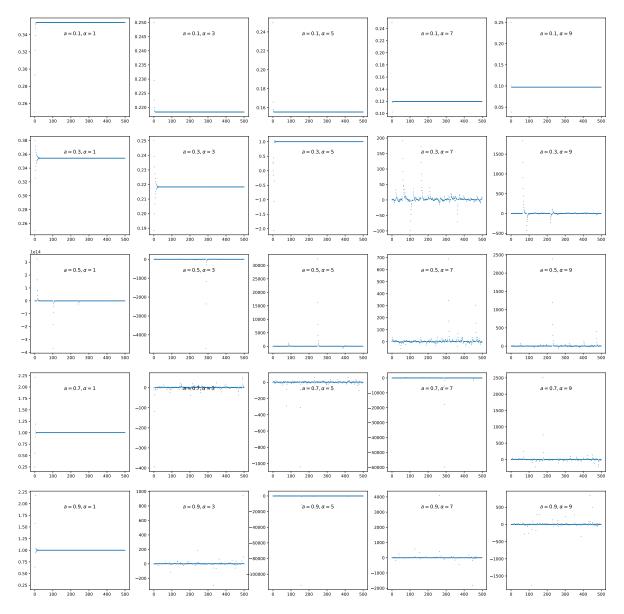
In our case, if recursive functions are such that $\epsilon = G(\epsilon)$, the G functions are, respectively

$$\begin{split} G_1(\epsilon) &= (1-a)\epsilon + a(1-\frac{\alpha(\tan(\pi\epsilon))}{\pi})\\ G_2(\epsilon) &= (1-a)\epsilon + a(\frac{1}{\pi}\arctan(\frac{\pi(1-\epsilon)}{\alpha})) \end{split}$$

We can check analytically by computing the derivative in the fixed point, or graphically see which of the choices of a yield to solutions that respect the condition in the neighbourhood of the fixed point.

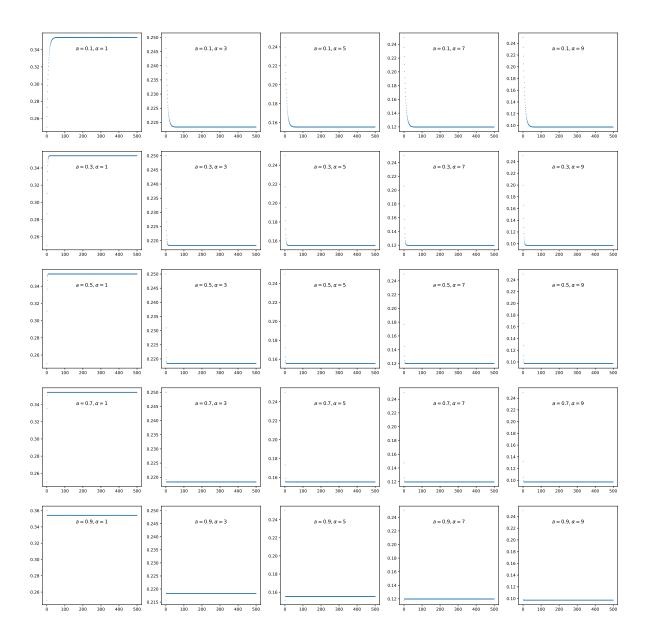


We can perform a grid search over the parameter space of the f_1 convergence. The grid represents convergence plots, going from the left to the right with α values 1, 3, 5, 7, 9 and going from the top to the bottom with a values 0.1, 0.3, 0.5, 0.7, 0.9.

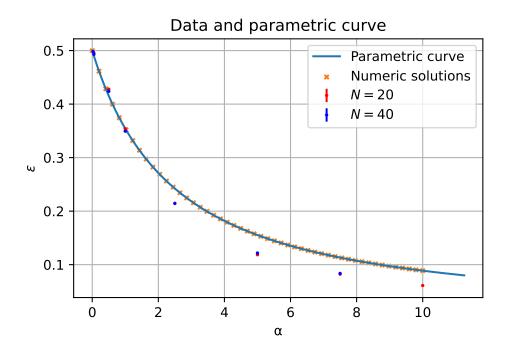


As a grows, we can see the onset of non-convergent behaviour on progressively lower values of α

We can check by doing a grid search over the parameter space that f_2 works much better. See the following grid: we have convergence for all the α 's, for all settings of a. The grid represents convergence plots, going from the left to the right with α values 1, 3, 5, 7, 9 and going from the top to the bottom with a values 0.1, 0.3, 0.5, 0.7, 0.9.



Step 7 and 8: Use the chosen function to find the value of the function for many ϵ 's and plot the final curves



The parametric curve and the numeric solutions agree perfectly, while the numeric solutions lie well below the formers.