

Design a differentially private average consensus scheme

The goal of this exercise is to implement a distributed average consensus scheme using a differentially private projected gradient descent algorithm.

You can use the algorithms discussed in Section 4 of “Privacy in Control and Dynamical Systems” by S. Han and G. J. Pappas for this assignment.

The use of MATLAB is recommended. The following tools are recommended to solve the assignment:

- 1) MATLAB `spmd` function to simulate parallel communicating agents*
- 2) A function to generate noise according to a Laplace distribution (provided in the appendix).*

[10 pts]

Solve the following distributed consensus problem

$$\min_{x_i \in \mathcal{X}} \sum_{i=1}^n \|x_i - v_i\|^2$$

Where $\mathcal{X} := \{x \in \mathbb{R} | -1 \leq x \leq 1\}$, $n = 4$, and private $v_1 = 0.1, v_2 = 0.5, v_3 = 0.4, v_4 = 0.2$.

[2pts] Implement Algorithm 1 in Section 4 of the paper above using the following tuning parameters: $T = 50, q = 0.6, c = 1$. If you are using MATLAB use the `spmd` functionality to run code in parallel on 4 workers. Plot the trajectories of the four agents obtained using Algorithm 1. Do they reach consensus? Provide the consensus value.

[4pts] Implement Algorithm 2 in Section 4 of the paper above using the following tuning parameters: $T = 50, q = 0.6, c = 1, \epsilon = 0.001$. If you use MATLAB, you can use the `randlap` MATLAB function provided in the appendix to generate Laplace noise. Explain how you selected the value of the parameter b_t . Plot the trajectories of the four agents obtained using Algorithm 2. Do they reach consensus? What happens if you increase $\epsilon = 0.01$?

[2pts] Now consider that $n = 8$, and $v_1 = v_5 = 0.1, v_2 = v_6 = 0.5, v_3 = v_7 = 0.4, v_4 = v_8 = 0.2$. If you are using MATLAB you can use the `spmd` functionality to run code in parallel on 8 workers. Use Algorithms 1 to check whether the agents still reach consensus. Plot the trajectories of the agents and their consensus value.

[2pts] Use Algorithms 2 to check whether the 8 agents still reach consensus for $\epsilon = 0.001$ and for $\epsilon = 0.01$. Plot the trajectories of the agents and their consensus value. Does a larger number of agents improve the performance of algorithm under differential privacy?

Submit the assignment together with the code you used to generate the results and the plots. Add a readme file with instructions to run the code.

Appendix: MATLAB function to generate Laplace noise

```
function x = randlap(siz,lambda)
% RANDL random numbers distributed according to the Laplace distribution
% RANDL(N) will return an NxN matrix containing pseudo-random values
% drawn from a Laplace distribution with zero mean and standard deviation
% one. RAND([M,N,...,P]) will return an MxNx...xP matrix.
% RANDL([M,N,...,P],lambda) will return an MxNx...xP matrix of
% pseudo-random numbers with parameter lambda. CAUTION: the pdf is
% assumed as
%      pdf = lambda/2 * exp(-lambda*abs(x))
%
% The Laplace random numbers are generated using the the RAND function to
% generate uniformly distributed numbers in (0,1) and then the probability
% integral transformation, i.e. using the fact that
% if Fx(x) is the cdf of a random variable x, then the RV z=Fx(x) is
% uniformly distributed in (0,1).
%
% In order to generate random numbers with mean mu and variance v, then
% generate the random numbers X with mean 0 and variance 1 and then
%      X = X*sqrt(v)+mu;
%
% C. Saragiotis, Oct 2008
if nargin==1
    lambda = sqrt(2); % this gives a std=var=1.
end
z = rand(siz);
x = zeros(siz);
in = z<=.5;
ip = z> .5;
x(in) = 1/lambda *log(2*z(in));
x(ip) = -1/lambda *log(2*(1-z(ip)));
```