

# Mathematics for New Technologies in Finance

## Exercise sheet 2

### Exercise 2.1 (Stone-Weierstrass theorem [1])

- (a) Construct a sequence of polynomials converges pointwisely but not uniformly on  $[0, 1]$ .
- (b) Construct a sequence of polynomials converges uniformly to  $x \mapsto |x|$  on  $[-1, 1]$ . (Hint: Corollary 2.3. in [1])
- (c) Prove that ReLU can be approximated uniformly by polynomials on  $[-1, 1]$ .
- (d) Use the universal approximation theory of shallow neural networks on  $[0, 1]$  to prove the Stone-Weierstrass theorem.

### Exercise 2.2 (Networks on discrete path spaces)

- (a) Describe the space of paths  $\omega : \{1, \dots, T\} \rightarrow \mathbb{R}^d$  as  $\mathbb{R}^{dT}$ .
- (b) Describe a shallow neural network, which depends on value at time  $t$  and on path information up to time  $t$ . Formulate a universal approximation theorem in this setting.

**Exercise 2.3 (Backpropagation of neural network)** Let  $\theta = (w, b, a) \in \mathbb{R}^3$  and let  $\sigma$  be the activation function. We consider the shallow neural network  $f_\theta : \mathbb{R} \rightarrow \mathbb{R}$  s.t.

$$f_\theta(x) = a\sigma(wx + b). \quad (1)$$

Then we solve the regression problem with 3 data point  $(x_i, y_i) \in \mathbb{R}^2$ ,  $i = 1, 2, 3$  by minimizing the  $L^2$  loss

$$\mathcal{L}_f = \sum_{i=1,2,3} (f_\theta(x_i) - y_i)^2. \quad (2)$$

- (a) When solving the regression, do we compute  $\nabla_{x_0} \mathcal{L}_f$  or  $\nabla_\theta \mathcal{L}_f$ ?
- (b) Compute  $\partial_w f$  and  $\partial_b f$  by chain rule. Do you find any intermediate value computed twice in both  $\partial_w f$  and  $\partial_b f$ ?
- (c) Consider regression problem as a constrained optimization problem

$$\begin{aligned} \min \quad & \sum_{i=1,2,3} l_i \\ l_i = & (\tilde{y}_i - y_i)^2 \\ \tilde{y}_i = & a\sigma(z_i), \quad i = 1, 2, 3. \\ z_i = & wx_i + b \end{aligned} \quad (3)$$

Solve it by Lagrange multiplier and relate this with backpropagation.

- (d) Generalize this idea to deep neural networks.

**Exercise 2.4 (Functional analysis)** Let  $K$  be a compact subset of  $\mathbb{R}^d$ .

(a) Let  $\mu$  be a finite Borel measure on  $K$ . Prove that

$$\mathcal{L}_\mu(f) := \int_K f(x)\mu(dx) \quad (4)$$

for  $f \in C(K, \mathbb{R})$  is a bounded linear functional.

(b) Let  $\mathcal{L}, C(K, \mathbb{R})$  be a positive linear functional, i.e.  $\mathcal{L}(f) \geq 0$  for  $f \geq 0$ . Then  $\mathcal{L}$  is continuous.

(c) Prove that

$$\mathcal{F} := \left\{ f \mapsto \sum_{i=1}^n \lambda_i f(x_i) \mid \lambda_i \in \mathbb{R}, n \in \mathbb{N}, x_i \in K, i = 1, 2, \dots, n \right\} \quad (5)$$

is point separating and additive.

## References

- [1] SAMEER CHAVAN. Problems and notes: Uniform convergence and polynomial approximation.
- [2] Ricky TQ Chen, Yulia Rubanova, Jesse Bettencourt, and David K Duvenaud. Neural ordinary differential equations. *Advances in neural information processing systems*, 31, 2018.
- [3] Hassan Ismail Fawaz, Germain Forestier, Jonathan Weber, Lhassane Idoumghar, and Pierre-Alain Muller. Deep learning for time series classification: a review. *Data mining and knowledge discovery*, 33(4):917–963, 2019.
- [4] Yann LeCun, D Touresky, G Hinton, and T Sejnowski. A theoretical framework for back-propagation. 1:21–28, 1988.