

**Take Home Exam, Part 1:
Statistical Information Theory (MT7037)**

Please read the following instructions carefully before starting:

- Deadline **February 23, 2021, at 8 pm.**
- Credit points will be given based on clear logical explanation and steps leading to the final solution, so do not simply state the final answer.
- Credit for the course is earned by writing reports for **two take-home exams.**
Obtainable grades together for two take-home exams:
 $A(\geq 90 \text{ points}), B(\geq 80 \text{ points}), C(\geq 70 \text{ points}), D(\geq 60 \text{ points}), \geq (50 \text{ points}), Fx \text{ and } F.$
Note that we reserve the right to rescale the points for grades by a factor c ($0.9 < c < 1$) stated depending on the outcome of the exam.
- **The report must be completed independently!** Plagiarism or other forms of cheating is a serious act - to underline this your report must as cover page contain the signed "confirmation sheet" that your work is made in accordance with the rules for written exams at Stockholm University.

Exercise 1 (16 points)

- (6 points) Let X and Y be discrete random variables. Show that $H((X + Y)|X) = H(Y|X)$. Argue that if X and Y are independent then $H(X + Y) \geq H(Y)$.
- (10 points) Show that among all N -valued random variables (i.e. with values $k = 1, 2, \dots$) with expected value μ , the geometric distributed random variable with expected value μ has the maximum value of Shannon entropy.

Reminder: The probability function in a geometric distribution is $p(k) = p(1 - p)^{k-1}, k = 1, 2, \dots$

Exercise 2 (9 points)

- (3 points) Show that the difference entropy is translation invariant i.e. that for a constant c , $h(X + c) = h(X)$, where X is a continuous random variable.
- (6 points) Suppose that (X, Y, Z) are jointly normal distributed and that $X \rightarrow Y \rightarrow Z$ forms a Markov chain. Let X and Y have the correlation coefficient equal to ρ_{XY} , and let Y and Z have the correlation coefficient equal to ρ_{YZ} .
Show that mutual information $I(X; Z) = -\frac{1}{2} \log(1 - \rho_{XY}^2 \rho_{YZ}^2)$.

Exercise 3 (25 points)

- (5 points) Define the transfer entropy, starting from Schreiber's definition in *Phys. Rev. Lett., Vol. 85, Number 2, 461-464, 2000*, in terms of mutual information and Shannon entropy.
- (20 points) To complete this part, you need to read the paper by Seghouane and Amari in *Neural Computation 24, 1722-1739, 2012 "Identification of Directed Influence: Granger Causality, Kullback-Leibler Divergence, and Complexity"* and answer the following questions:
 - (5 points) Look at the formula (2.1) in the paper. Is it equivalent to any of the concepts you have learned during the course? If yes, then to which ones?

- (ii) (5 points) What is the difference between the directed influence measure defined in the formula (2.2) and transfer entropy definition by Schreiber?
- (iii) (5 points) Comment on the formula (4.5) in terms of what you have learned during the course.
- (iv) (5 points) Please provide your opinion about the directed influence measures presented in the paper (the potential advantages and/or disadvantages).

Good Luck!