

Exam Optimization of Business Processes

30 May 2017

This exam consists of **5** problems, each consisting of several questions.

All answers should be motivated, including calculations, formulas used, etc.

It is allowed to use 1 sheet of paper (or 2 sheets written on one side) with **hand-written** notes.

The minimal note is 1. All questions (1a, 1b, etc.) give the same number of points.

The use of a calculator and a dictionary are allowed.

A table with the normal distribution and a table with the Poisson distribution are attached.

- 1a. Explain in which order production, demand and consumption are executed in the areas of manufacturing, customer service, and elective health care.
- b. Give an example of each of these three ways of ordering in the area of health care.

2. A flow line with 2 stations has Poisson arrivals with rate 3 and average service times of 0.25 at both stations.

- a. Compute or approximate the expected waiting time at both queues when the service times are exponential in the first and constant in the second queue.
- b. Compute or approximate the expected waiting time at both queues when the service times are constant in the first and exponential in the second queue.
- c. Which one is shorter? Explain intuitively your answer.

3. A project consists of only 3 activities, A, B, and C. A and B can be done in parallel, they have to be finished before C can start. They all have uniform distributions on $[0, b]$ with different values of b : 1 for A, 2 for B, and 3 for C.

- a. What is the earliest finish time of this project, ignoring randomness?
- b. What is the expected earliest finish time of this project, using PERT?
- c. Compute exactly the expected earliest finish time of the project.

4. Consider a single-machine scheduling problem with n jobs, release dates r_i and production times s_i .
 - a. Give a linear programming formulation that minimizes the flowtime.
 - b. The objective is changed to minimizing the makespan. Give a linear programming formulation that minimizes it.

Now we have a second machine, and each job has first to be processed at machine 1 before it goes to machine 2. Service times at machine 2 are s'_i .

- c. Give a linear programming formulation that minimizes the flowtime.
- d. Construct a simple example showing that it might not be optimal to process jobs in the same order on both machines.

- 5a. Consider the dimensioning of hospital wards with the Erlang B model. Explain how you can use the Poisson distribution to compute the occupancy and the rejection probability.
- b. Do this for on average 2 arrivals per day, an ALOS of 5 days, and 12 beds.
- c. Explain what "simple merging" is and give 2 reasons why this might not be a good idea.

Consider merging 2 wards with fully multi-skilled nurses. There are costs c_i for every patients of type i who is rejected. Assume that time is discretized in such a way that every patient of type i leaves with probability μ_i . Assume that $\lambda_1 + \lambda_2 + s \max\{\mu_1, \mu_2\} \leq 1$, with λ_i the probability of an arrival for type i and s the number of beds.

- d. Give a dynamic programming formulation that minimizes the expected total rejection costs over a certain time horizon.

Table with values of $\mathbb{P}(X > k)$ with X a Poisson distributed random variable with mean μ

	values of μ									
values of k	1	2	3	4	5	6	7	8	9	10
0	0.632	0.865	0.950	0.982	0.993	0.998	0.999	1.000	1.000	1.000
1	0.264	0.594	0.801	0.908	0.960	0.983	0.993	0.997	0.999	1.000
2	0.080	0.323	0.577	0.762	0.875	0.938	0.970	0.986	0.994	0.997
3	0.019	0.143	0.353	0.567	0.735	0.849	0.918	0.958	0.979	0.990
4	0.004	0.053	0.185	0.371	0.560	0.715	0.827	0.900	0.945	0.971
5	0.001	0.017	0.084	0.215	0.384	0.554	0.699	0.809	0.884	0.933
6	0.000	0.005	0.034	0.111	0.238	0.394	0.550	0.687	0.793	0.870
7	0.000	0.001	0.012	0.051	0.133	0.256	0.401	0.547	0.676	0.780
8	0.000	0.000	0.004	0.021	0.068	0.153	0.271	0.407	0.544	0.667
9	0.000	0.000	0.001	0.008	0.032	0.084	0.170	0.283	0.413	0.542
10	0.000	0.000	0.000	0.003	0.014	0.043	0.099	0.184	0.294	0.417
11	0.000	0.000	0.000	0.001	0.005	0.020	0.053	0.112	0.197	0.303
12	0.000	0.000	0.000	0.000	0.002	0.009	0.027	0.064	0.124	0.208
13	0.000	0.000	0.000	0.000	0.001	0.004	0.013	0.034	0.074	0.136
14	0.000	0.000	0.000	0.000	0.000	0.001	0.006	0.017	0.041	0.083
15	0.000	0.000	0.000	0.000	0.000	0.001	0.002	0.008	0.022	0.049
16	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.004	0.011	0.027
17	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.002	0.005	0.014
18	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.002	0.007
19	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001	0.003
20	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.002
21	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.001
22	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
23	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
24	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
25	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000

Table with values of $\mathbb{P}(0 < X < x + y)$ with X a random variable with a standard normal distribution

	values of y									
values of x	0	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0	0.000	0.004	0.008	0.012	0.016	0.020	0.024	0.028	0.032	0.036
0.1	0.040	0.044	0.048	0.052	0.056	0.060	0.064	0.067	0.071	0.075
0.2	0.079	0.083	0.087	0.091	0.095	0.099	0.103	0.106	0.110	0.114
0.3	0.118	0.122	0.126	0.129	0.133	0.137	0.141	0.144	0.148	0.152
0.4	0.155	0.159	0.163	0.166	0.170	0.174	0.177	0.181	0.184	0.188
0.5	0.191	0.195	0.198	0.202	0.205	0.209	0.212	0.216	0.219	0.222
0.6	0.226	0.229	0.232	0.236	0.239	0.242	0.245	0.249	0.252	0.255
0.7	0.258	0.261	0.264	0.267	0.270	0.273	0.276	0.279	0.282	0.285
0.8	0.288	0.291	0.294	0.297	0.300	0.302	0.305	0.308	0.311	0.313
0.9	0.316	0.319	0.321	0.324	0.326	0.329	0.331	0.334	0.336	0.339
1	0.341	0.344	0.346	0.348	0.351	0.353	0.355	0.358	0.360	0.362
1.1	0.364	0.367	0.369	0.371	0.373	0.375	0.377	0.379	0.381	0.383
1.2	0.385	0.387	0.389	0.391	0.393	0.394	0.396	0.398	0.400	0.401
1.3	0.403	0.405	0.407	0.408	0.410	0.411	0.413	0.415	0.416	0.418
1.4	0.419	0.421	0.422	0.424	0.425	0.426	0.428	0.429	0.431	0.432
1.5	0.433	0.434	0.436	0.437	0.438	0.439	0.441	0.442	0.443	0.444
1.6	0.445	0.446	0.447	0.448	0.449	0.451	0.452	0.453	0.454	0.454
1.7	0.455	0.456	0.457	0.458	0.459	0.460	0.461	0.462	0.462	0.463
1.8	0.464	0.465	0.466	0.466	0.467	0.468	0.469	0.469	0.470	0.471
1.9	0.471	0.472	0.473	0.473	0.474	0.474	0.475	0.476	0.476	0.477
2	0.477	0.478	0.478	0.479	0.479	0.480	0.480	0.481	0.481	0.482
2.1	0.482	0.483	0.483	0.483	0.484	0.484	0.485	0.485	0.485	0.486
2.2	0.486	0.486	0.487	0.487	0.487	0.488	0.488	0.488	0.489	0.489
2.3	0.489	0.490	0.490	0.490	0.490	0.491	0.491	0.491	0.491	0.492
2.4	0.492	0.492	0.492	0.492	0.493	0.493	0.493	0.493	0.493	0.494
2.5	0.494	0.494	0.494	0.494	0.494	0.495	0.495	0.495	0.495	0.495
2.6	0.495	0.495	0.496	0.496	0.496	0.496	0.496	0.496	0.496	0.496
2.7	0.497	0.497	0.497	0.497	0.497	0.497	0.497	0.497	0.497	0.497
2.8	0.497	0.498	0.498	0.498	0.498	0.498	0.498	0.498	0.498	0.498
2.9	0.498	0.498	0.498	0.498	0.498	0.498	0.498	0.499	0.499	0.499
3	0.499	0.499	0.499	0.499	0.499	0.499	0.499	0.499	0.499	0.499