### MVE055 2018 Lecture 3

#### Marco Longfils

Department of Mathematical Sciences, Chalmers University of Technology and University of Gothenburg

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## Continuous random variables

- Continuous random variables are random variables which can assume any value in an interval of the real numbers, or the entire real line. They are useful in many application (e.g. the time an event happens).
- Discrete random variables are characterized by  $\Pr[X=x] > 0$  for at most countably many values x. If  $\Pr[X=x] > 0$  for uncountably many x, then the total probability would be infinite. Thus, continuous random variables are such that  $\Pr[X=x] = 0$  for all x, but  $\Pr[X \in [a,b]] > 0$ .
- ullet We assume there exists a function f, called density function, such that

$$\Pr[X \in [a, b]] = \int_{a}^{b} f(x)dx$$

• As a consequence  $\Pr[X \in [a,b]] = \Pr[X \in (a,b]] = \Pr[X \in [a,b)] = \Pr[X \in (a,b)]$ 

# Density function

- The density function f satisfies
  - 1.  $f(x) \ge 0$  for all  $x \in \mathbb{R}$ ;
  - $2. \int_{-\infty}^{+\infty} f(x)dx = 1.$
- Again, the distribution function is defined as

$$F(x) = \Pr[X \le x]$$

and it can be computed as

$$F(x) = \int_{-\infty}^{x} f(t)dt$$

and in the continuous case we have

$$f(x) = F'(x).$$

## Expected value, variance, and standard deviation

• The expected value of a continuous random variable X is defined as

$$\mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) dx$$

• In general, the expected value of a function of X is given by

$$\mathbb{E}[H(X)] = \int_{-\infty}^{\infty} h(x)f(x)dx$$

• Variance and standard deviation are defined as

$$Var[X] = \mathbb{E}[(X - \mathbb{E}[X])^{2}]$$
$$Sd[X] = \sqrt{Var[X]}$$

• Same properties as in the discrete case.

### Normal distribution

#### Definition (Normal distribution)

A random variable X is said to have a normal distribution with parameters  $\mu \in \mathbb{R}$  and  $\sigma > 0$  if it has density

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{(x-\mu)^2}{2\sigma^2}}$$
, for all  $x \in R$ 

- Notation:  $X \sim N(\mu, \sigma^2)$ .
- $\mathbb{E}[X] = \mu, \operatorname{Var}[X] = \sigma^2$

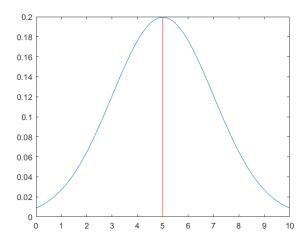


Figure: Density function of a normal random variable with parameters  $\mu=5, \sigma=2.$ 

## Standard normal distribution

#### Definition

A normal random variable  $X \sim N(0,1)$  with parameters  $\mu = 0, \sigma = 1$  is called standard normal.

• Standardization: if  $X \sim N(\mu, \sigma^2)$  then  $\frac{X-\mu}{\sigma} \sim N(0, 1)$ .

•

$$\Pr[\mu - \sigma < X < \mu + \sigma] \approx 0.68$$

$$\Pr[\mu - 2\sigma < X < \mu + 2\sigma] \approx 0.95$$

$$\Pr[\mu - 3\sigma < X < \mu + 3\sigma] \approx 0.997$$

• To compute probabilities like  $\Pr[X \in [a, b]]$  we standardize X and use the tables (see end of the book).

$F_Z(z) = P[Z \le z]$										
z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
-3.4 -3.3	0.0003	0.0003	0.0003	0.0003	0.0003	0.0003				0.07
-3.3 -3.2	0.0003	0.0003	0.0005	0.0004	0.0004	0.0003				0.0002
-3.1	0.0010	0.0009	0.0006	0.0006	0.0006	0.0004				0.0003
-3.0	0.0013	0.0003	0.0009	0.0009	0.0008	0.0008				0.0005
			0.0013	0.0012	0.0012	0.0011				0.0007
-2.9	0.0019	0.0018	0.0018	0.0017	0.0016				0.0010	0.0010
-2.8	0.0026	0.0025	0.0024	0.0023	0.0016	0.0016				0.0014
-2.7	0.0035	0.0034	0.0033	0.0032	0.0023	0.0022	0.0021			0.0019
-2.6	0.0047	0.0045	0.0044	0.0043	0.0031	0.0030	0.0029			0.0026
-2.5	0.0062	0.0060	0.0059	0.0057	0.0055	0.0040 0.0054	0.0039	0.0038		0.0036
-2.4	0.0082	0.0080	0.0078				0.0052	0.0051	0.0049	0.0048
-2.3	0.0107	0.0104	0.0102	0.0075	0.0073	0.0071	0.0069	0.0068	0.0066	0.0064
-2.2	0.0139	0.0136	0.0102	0.0099	0.0096	0.0094	0.0091	0.0089	0.0087	0.0084
-2.1	0.0179	0.0174	0.0170	0.0129	0.0125	0.0122	0.0119	0.0116	0.0113	0.0110
-2.0	0.0228	0.0222	0.0217	0.0166	0.0162	0.0158	0.0154	0.0150	0.0146	0.0143
					0.0207	0.0202	0.0197	0.0192	0.0188	0.0183
1.9	0.0287	0.0281	0.0274	0.0268	0.0262	0.0256	0.0250	0.0244	0.0239	0.0233
1.8	0.0359	0.0351	0.0344	0.0336	0.0329	0.0322	0.0314	0.0307	0.0301	0.0294
1.7	0.0446	0.0436	0.0427	0.0418	0.0409	0.0401	0.0392	0.0384	0.0375	0.0367
1.6	0.0548	0.0537	0.0526	0.0516	0.0505	0.0495	0.0485	0.0475	0.0465	0.0455
1.5	0.0668	0.0655	0.0643	0.0630	0.0618	0.0606	0.0594	0.0582	0.0571	0.0559
1.4	0.0808	0.0793	0.0778	0.0764	0.0749	0.0735	0.0721	0.0708	0.0694	0.0681
1.3	0.0968	0.0951	0.0934	0.0918	0.0901	0.0885	0.0869	0.0853	0.0838	0.0823
1.2	0.1151	0.1131	0.1112	0.1093	0.1075	0.1056	0.1038	0.1020	0.1003	0.098
1.1	0.1357	0.1335	0.1314	0.1292	0.1271	0.1251	0.1230	0.1210	0.1190	0.117
1.0	0.1587	0.1562	0.1539	0.1515	0.1492	0.1469	0.1446	0.1423	0.1401	0.137
		0.1814	0.1788	0.1762	0.1736	0.1711	0.1685	P(X <= -0.	<b>671</b> 1635	0.161
0.9	0.1841		0.2061	0.2033	0.2005	0.1977	0.1949	0.1	0.1894	0.186
0.8	0.2119	0.2090		0.2327	0.2296	0.2266	0.2236	0.2 .6	0.2177	0.214
0.7	0.2420	0.2389	0.2358		0.2230	0.2578	0.2546	0.2514		0.245
0.6	0.2743	0.2709	0.2676	0.2643		0.2912	0.2877	0.2843		
0.5	0.3085	0.3050	0.3015	0.2981	0.2946	0.2912				
0.4	0.3446	0.3409	0.3372	0.3336	0.3300	0.3264	0.3228	0.3192		
0.3	0.3821	0.3409	0.3745	0.3707	0.3669	0.3632		0.3557		
0.2			0.4129	0.4090	0.4052	0.4013	0.3974			
0.2	0.4207	0.4168		0.4483	0.4443	0.4404				
0.0	0.4602	0.4562	0.4522	0.4483	0.4840	0.4801		0.472	0.468	0.46

TABLE V
Computative distribution: Standard normal (concluded)

z	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.5000	0.5040	0.5080	0.5120	0.5160	0.5199	0.5239	0.5279	0.5319	0.535
0.0	0.5000	0.5438	0.5478	0.5120	0.5557	0.5596	0.5636	0.5675	0.5714	0.575
0.1	0.5398	0.5832	0.5478	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.614
0.3	0.6179	0.5852	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.651
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.687
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.722
0.6	0.7257	0.7291	0.7324	0.7357	0.7389	0.7422	0.7454	0.7486	0.7517	0.7549
0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.785
0.8	0.7881	0.7910	0.7939	0.7967	0.7995	0.8023	0.8051	0.8078	0.8106	0.8133
0.9	0.8159	0.8186	0.8212	0.8238	0.8264	0.8289	0.8315	0.8340	0.8365	0.8389
1.0	0.8413	0.8438	0.8461	0.8485	0.8508	0.8531	0.8554	0.8577	0.8599	0.8621
1.1	0.8643	0.8665	0.8686	0.8708	0.8729	0.8749	0.8770	0.8790	0.8810	0.8830
1.2	0.8849	0.8869	0.8888	0.8907	0.8925	0.8944	0.8962	0.8980	0.8997	0.9015
1.3	0.9032	0.9049	0.9066	0.9082	0.9099	0.9115	0.9131	0.9147	0.9162	0.9177
1.4	0.9192	0.9207	0.9222	0.9236	0.9251	0.9265	0.9279	0.9292	0.9306	0.9319
1.5	0.9332	0.9345	0.9357	0.9370	0.9382	0.9394	0.9406	0.9418	0.9429	0.9441
1.6	0.9452	0.9463	0.9474	0.9484	0.9495	0.9505	0.9515	0.9525	0.9535	0.9545
1.7	0.9554	0.9564	0.9573	0.9582	0.9591	0.9599	0.9608	0.9616	0.9625	0.9633
1.8	0.9641	0.9649	0.9656	0.9664	0.9671	0.9678	0.9686	0.9693	0.9699	0.9706
1.9	0.9713	0.9719	0.9726	0.9732	0.9738	0.9744	0.9750	0.9756	0.9761	0.9767
2.0	0.9772	0.9778	0.9783	0.9788	0.9793	0.9798	0.9803	0.9808	0.9812	0.9817
2.1	0.9821	0.9826	0.9830	0.9834	0.9838	0.9842	0.9846	0.9850	0.9854	0.9857
2.2	0.9861	0.9864	0.9868	0.9871	0.9875	0.9878	0.9881	0.9884	0.9887	0.9890
2.3	0.9893	0.9896	0.9898	0.9901	0.9904	0.9906	0.9909	0.9911	0.9913	0.9916
2.4	0.9918	0.9920	0.9922	0.9925	0.9927	0.9929	0.9931	0.9932	0.9934	0.9936
2.5	0.9938	0.9940	0.9941	0.9943	0.9945	0.9946	0.9948	0,9949	0.9951	0.9952
2.6	0.9953	0.9955	0.9956	0.9957	0.9959	0.9960	0.9961	0.9962	0.9963	0.9964
2.7	0.9965	0.9966	0.9967	0.9968	0.9969	0.9970	0.9971	0.9972	0.9903	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9972	0.9973	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9979	0.9980	0.9986
3.0	0.9987	0.9987	0.9987	0.9988	0.9988	0.9989	0.9989			0.9990
3.1	0.9990	0.9991	0.9991	0.9991	0.9992	0.9992		0.9989	0.9990	
3.2	0.9993	0.9993	0.9994	0.9994	0.9994	0.9992	0.9992	0.9992	0.9993	0.9993
3.3	0.9995	0.9995	0.9995	0.9996	0.9996	0.9994	0.9994	0.9995	0.9995	0.9995
3.4	0.9997	0.9997	0.9997	0.9997	0.9997	0.9996	0.9996	0.9996	0.9996	0.9997
					0,2331	0.999/	0.9997	0.9997	0.9997	0.9998

#### Normal distribution

- The normal distribution can model well many measurement arising in experiments (N.B: many ≠ all);
- We will discuss more about the normal distribution later in the course (CLT, linear regression).
- It can be used to approximate other distribution.

# Normal approximation to the binomial distribution

• Recall: the binomial distribution has frequency function

$$\Pr[X=k] = \binom{n}{k} p^k (1-p)^{n-k}$$

and it has  $\mathbb{E}[X] = np \operatorname{Var}[X] = np(1-p)$ .

• When n is large it is hard to compute something like

$$\Pr[X \le k] = \sum_{i=0}^{k} \binom{n}{i} p^{i} (1-p)^{n-i}$$

- $\rightarrow$  Approximate the binomial random variable X with a normal random variable  $Y \sim N(np, np(1-p))$  when n is large and p is far from 0 and 1.
- Continuity correction:  $\Pr[X \le k] \approx \Pr[Y \le k + \frac{1}{2}]$  and  $\Pr[X < k] \approx \Pr[Y \le k \frac{1}{2}]$

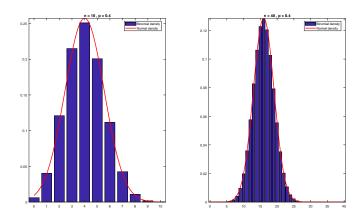


Figure: Left is borderline to be acceptable, while the one on the right would provide a good approximation.

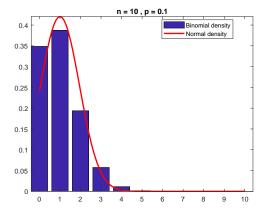


Figure: Empirical rule for good approximation: either  $p \le 0.5$  and np > 5 or p > 0.5 and n(1-p) > 5.

#### Transformation of variables

#### Theorem

Let X be a continuous random variable with density  $f_X$ . Moreover, let Y = g(X) where g is a strictly monotonic and differentiable function. The density  $f_Y$  of Y is given by

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dg^{-1}(y)}{dy} \right|$$

example: if  $X \sim N(0,1)$  and  $Y = \sigma X + \mu$  then

$$f_y(y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{(y-\mu)^2}{2\sigma^2}}$$

$$\rightarrow Y \sim N(\mu, \sigma^2)$$