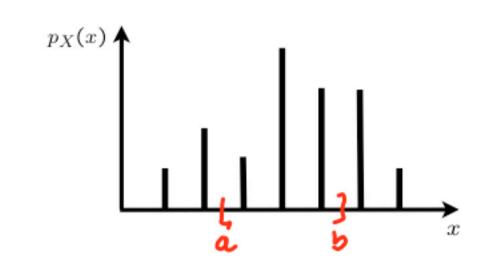
#### LECTURE 8: Continuous random variables and probability density functions

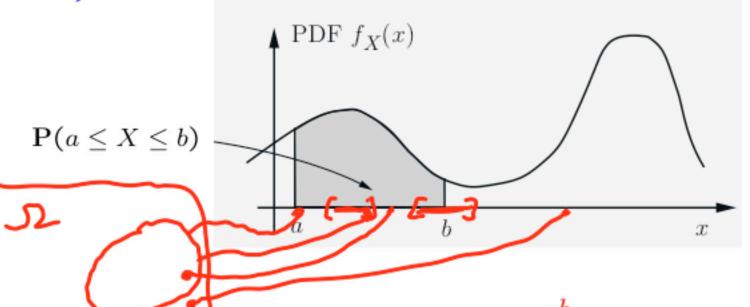
- Probability density functions
  - Properties
  - Examples
- Expectation and its properties
- The expected value rule
- Linearity
- Variance and its properties
- Uniform and exponential random variables
- Cumulative distribution functions
- Normal random variables
  - Expectation and variance
  - Linearity properties
  - Using tables to calculate probabilities

## Probability density functions (PDFs)



$$P(a \le X \le b) = \sum_{x: a \le x \le b} p_X(x)$$

$$p_X(x) \ge 0 \qquad \qquad \sum_x p_X(x) = 1$$



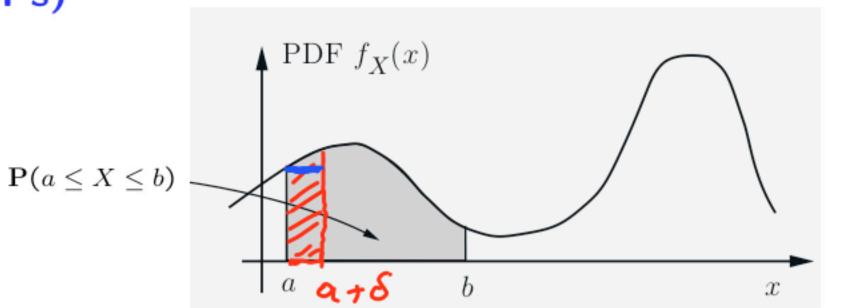
$$P(a \le X \le b) = \int_a^b f_X(x) dx$$

• 
$$f_X(x) \ge 0$$
 •  $\int_{-\infty}^{\infty} f_X(x) dx = 1$ 

Definition: A random variable is continuous if it can be described by a PDF

### Probability density functions (PDFs)

$$\approx f_{x}(a).\delta$$



$$P(a \le X \le a + \delta) \approx f_X(a) \cdot \delta$$

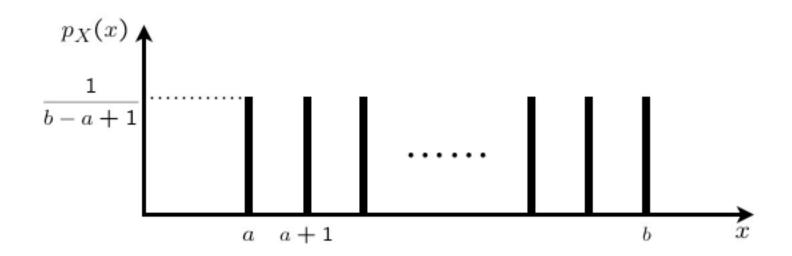
$$P(X=a)=0$$

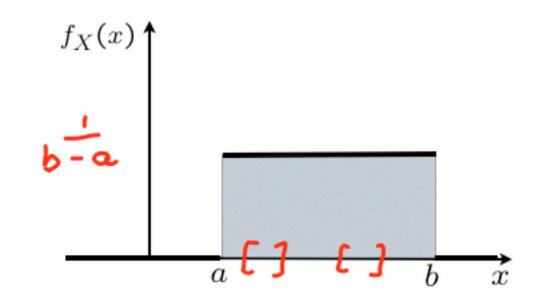
$$\mathbf{P}(a \le X \le b) = \int_a^b f_X(x) \, dx$$

$$f_X(x) \geq 0$$

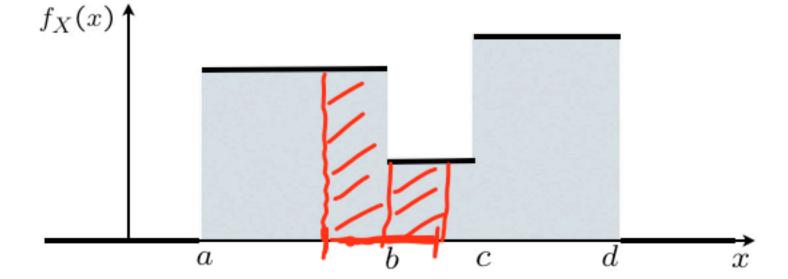
$$\int_{-\infty}^{\infty} f_X(x) \, dx = 1$$

# **Example: continuous uniform PDF**

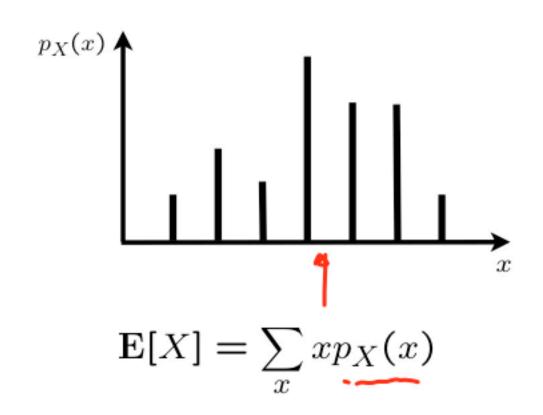


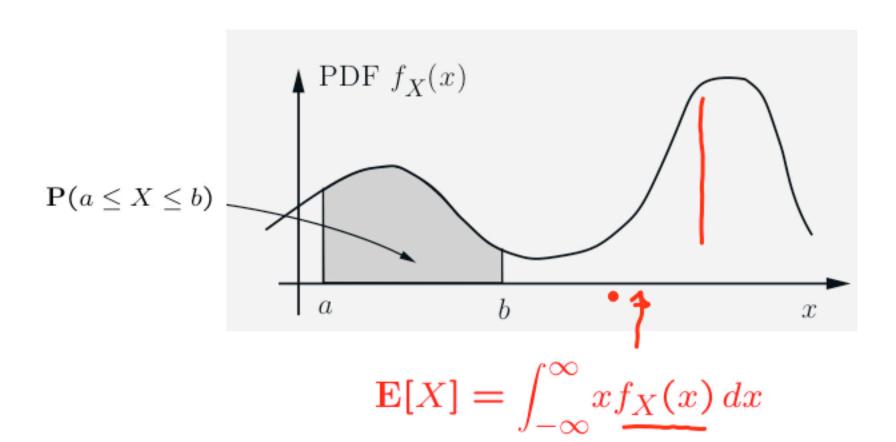


• Generalization: piecewise constant PDF



### Expectation/mean of a continuous random variable





 Interpretation: Average in large number of independent repetitions of the experiment

Fine print: 
$$\operatorname{Assume} \, \int_{-\infty}^{\infty} |x| \, f_X(x) \, dx < \infty$$

#### **Properties of expectations**

- If  $X \ge 0$ , then  $\mathbf{E}[X] \ge 0$
- If  $a \le X \le b$ , then  $a \le \mathbf{E}[X] \le b$
- Expected value rule:

$$\mathbf{E}[g(X)] = \sum_{x} g(x) p_{X}(x)$$

$$\mathbf{E}[g(X)] = \int_{-\infty}^{\infty} g(x) f_{X}(x) dx$$

$$\mathbf{E}[X^{2}] = \int_{-\infty}^{\infty} x^{2} \int_{X} (x) dx$$

Linearity

$$\mathbf{E}[aX + b] = a\mathbf{E}[X] + b$$

#### Variance and its properties

• Definition of variance:  $var(X) = E[(X - \mu)^2]$ 

Calculation using the expected value rule,  $\mathbf{E}[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) \, dx$   $(x - \nu)^2 \int_{-\infty}^{\infty} (x) \, dx$   $g(x) = (x - \mu)^2$ 

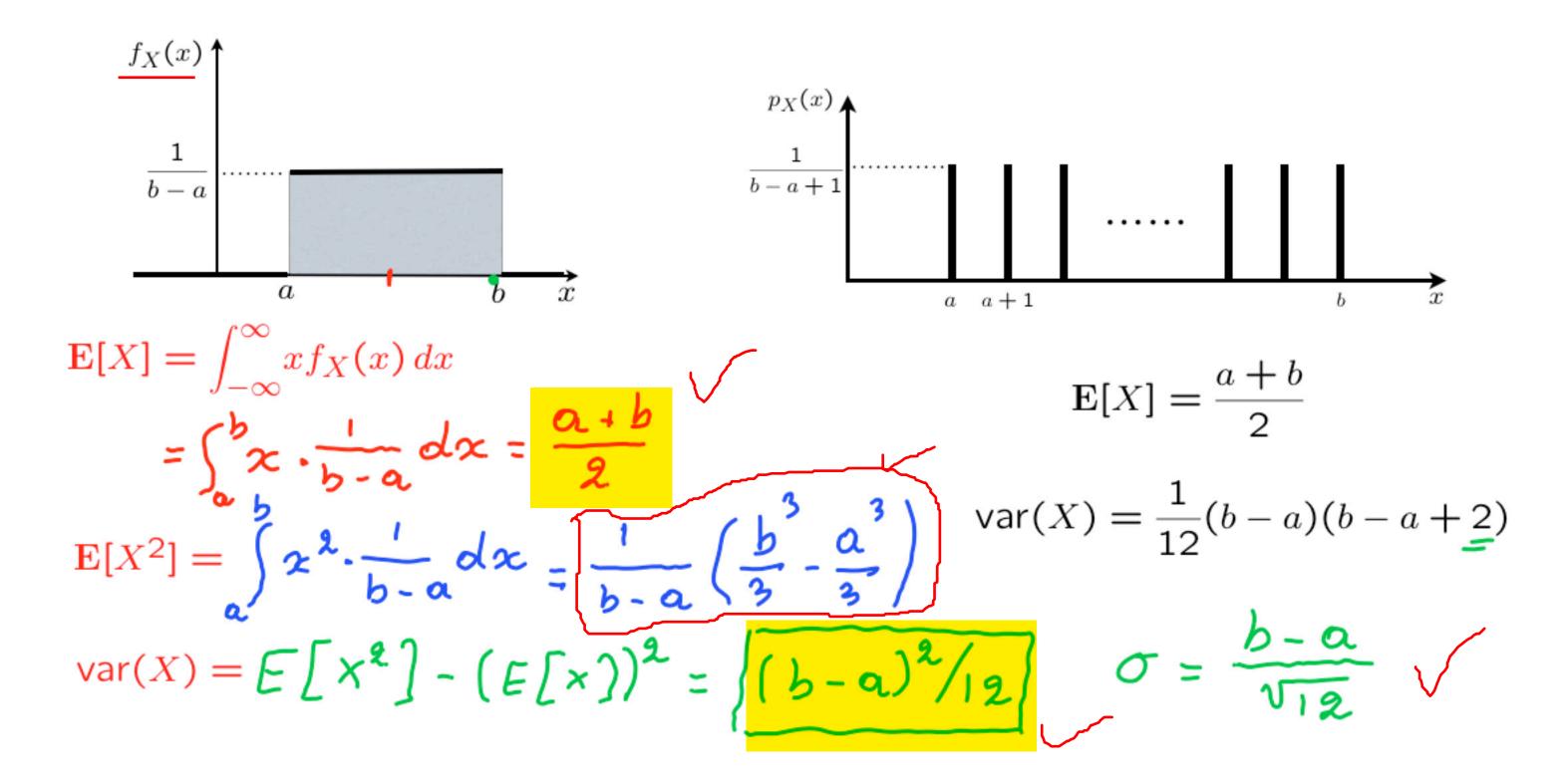
$$var(X) = \int_{-\infty}^{\infty} (x - \mu)^2 f_{x}(x) dx$$

Standard deviation:  $\sigma_X = \sqrt{\text{var}(X)}$ 

 $var(aX + b) = a^2 var(X)$ 

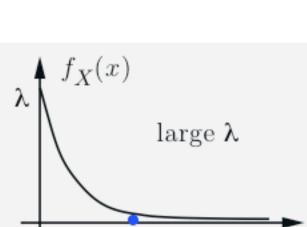
A useful formula:  $var(X) = \mathbf{E}[X^2] - (\mathbf{E}[X])^2$ 

## Continuous uniform random variable; parameters a, b



## **Exponential random variable**; parameter $\lambda > 0$

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & x \ge 0 \\ 0, & x < 0 \end{cases} \qquad \begin{cases} f_X(x) dx = 1 \\ f_X(x) dx = 1 \end{cases}$$



$$E[X] = \int_{0}^{\infty} x \cdot \lambda e^{-\lambda x} dx = \frac{1}{\lambda}$$

$$E[X^{2}] = \int_{0}^{\infty} x^{2} \lambda e^{-\lambda x} dx = \frac{2}{\lambda^{2}}$$

$$Var(X) = E[X^{2}] - (E[X])^{2} = \frac{1}{\lambda^{2}}$$

$$E[x] = \frac{1}{p}$$

$$(1-p)^{k-1}p$$

$$F(X \ge a) = \begin{cases} 00 \\ \lambda e^{-\lambda x} dx \end{cases}$$

$$= \lambda \cdot (-\frac{1}{\lambda}) e^{-\lambda x}$$

$$= -e^{-\lambda \cdot 00} + e^{-\lambda a} = e^{-\lambda a}$$

## **Cumulative distribution function (CDF)**

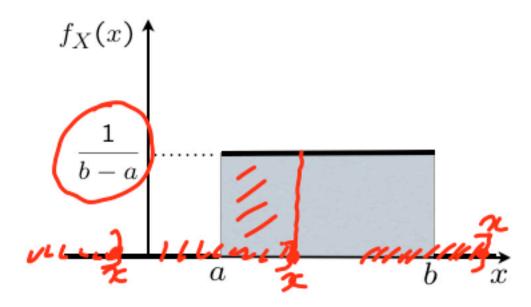
CDF definition: 
$$F_X(x) = P(X \le x)$$

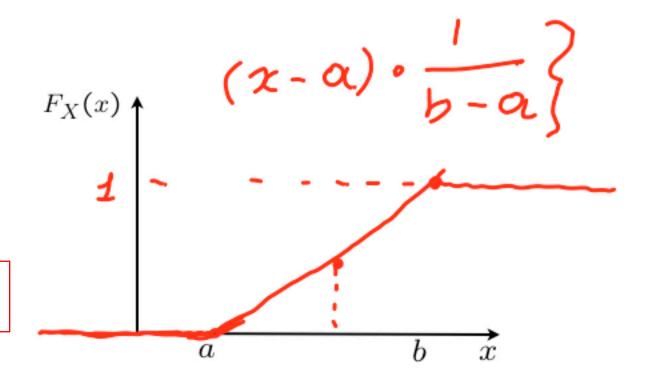
Continuous random variables:

$$F_X(x) = \mathbf{P}(X \le x) = \int_{-\infty}^x f_X(t) dt$$

$$\frac{df_x}{dx}(x) = f_x(x)$$

To generate the pdf from the cdf.



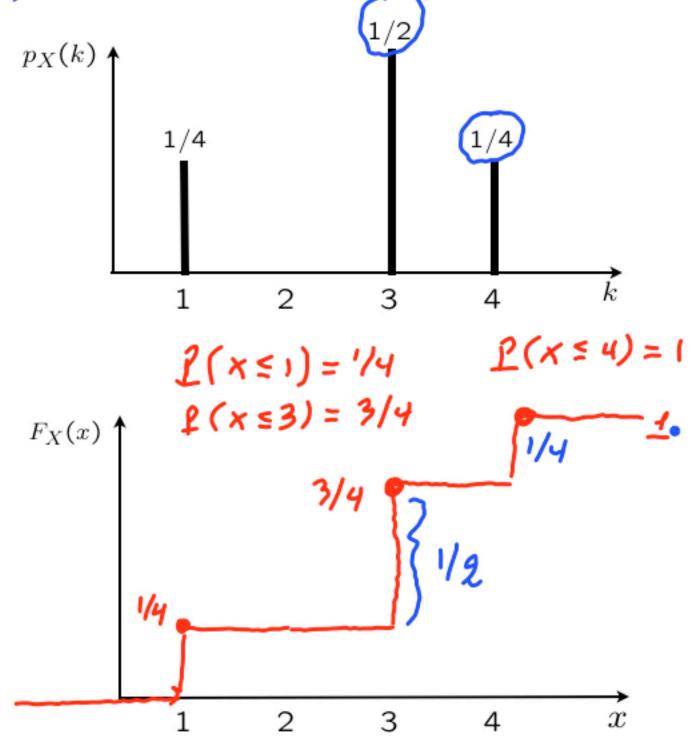


## **Cumulative distribution function (CDF)**

CDF definition: 
$$F_X(x) = P(X \le x)$$

Discrete random variables:

$$F_X(x) = \mathbf{P}(X \le x) = \sum_{k \le x} p_X(k)$$



#### **General CDF properties**

$$F_X(x) = \mathbf{P}(X \le x)$$

- Non-decreasing If  $y \ge x \implies F_x(y) \ge F_x(x)$
- $F_X(x)$  tends to 1, as  $x \to \infty$
- $F_X(x)$  tends to 0, as  $x \to -\infty$

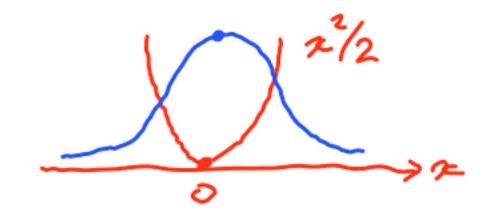
## Normal (Gaussian) random variables

- Important in the theory of probability
  - Central limit theorem

- Prevalent in applications
  - Convenient analytical properties
  - Model of noise consisting of many, small independent noise terms

## Standard normal (Gaussian) random variables

• Standard normal N(0,1):  $f_X(x) = \frac{1}{\sqrt{2\pi}}e^{-x^2/2}$ 



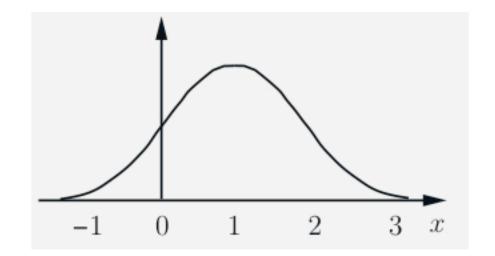
$$\frac{2^{2}/2}{2^{2}} \qquad \text{calculus:}$$

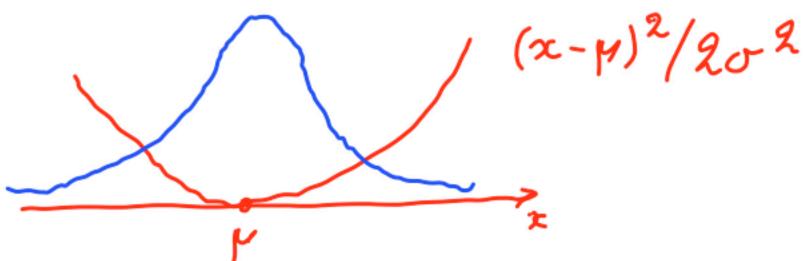
$$\frac{\cos(-2^{2}/2)}{\cos(-2^{2}/2)} dx = \sqrt{2\pi}$$

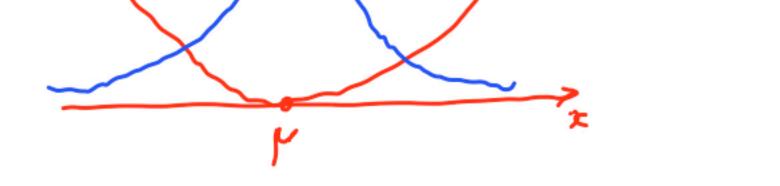
- $\bullet$   $\mathbf{E}[X] = \mathbf{O}$
- var(X) = 1 integrate by parts

## General normal (Gaussian) random variables

• General normal  $N(\mu, \sigma^2)$ :  $f_X(x) = \frac{1}{\sigma\sqrt{2\pi}}e^{-(x-\mu)^2/2\sigma^2}$ 5>0

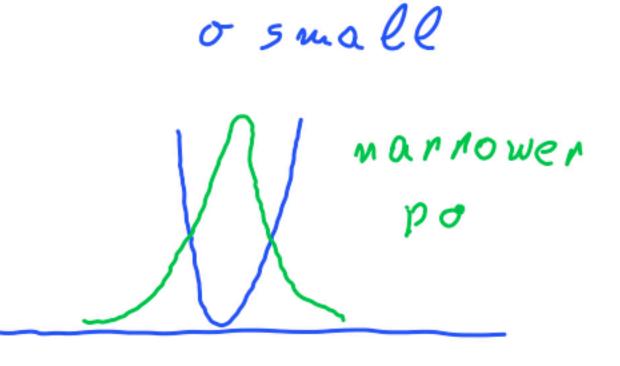








•  $\operatorname{var}(X) = \sigma^2$ 



#### Linear functions of a normal random variable

• Let 
$$Y = aX + b$$
  $X \sim N(\mu, \sigma^2)$ 

$$\mathbf{E}[Y] = \mathbf{a} + \mathbf{b}$$

$$Var(Y) = \alpha^2 \sigma^2$$

Fact (will prove later in this course):

$$Y \sim N(a\mu + b, a^2\sigma^2)$$

• Special case: a = 0?

$$Y=b$$
 oliscrete  
 $N(b,0)$ 

This table given only < less than value of the probability.

Don't be consider > probability.

#### Standard normal tables

No closed form available for CDF

but have tables, for the standard normal

$$\frac{9(-2)}{1-9(2)} = 1-.9772$$

	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
0.0	5000	.5040	.5080	.5120	.5160	.5199	.5239	.5279	.5319	.5359
0.1	.5398	.5438	.5478	.5517	.5557	.5596	.5636	.5675	.5714	.5753
0.2	.5793	.5832	.5871	.5910	.5948	.5987	.6026	.6064	.6103	.6141
0.3	.6179	.6217	.6255	.6293	.6331	.6368	.6406	.6443	.6480	.6517
0.4	.6554	.6591	.6628	.6664	.6700	.6736	.6772	.6808	.6844	.6879
0.5	.6915	.6950	.6985	.7019	.7054	.7088	.7123	.7157	.7190	.7224
0.6	.7257	.7291	.7324	.7357	.7389	.7422	.7454	.7486	.7517	.7549
0.7	.7580	.7611	.7642	.7673	.7704	.7734	.7764	.7794	.7823	.7852
0.8	.7881	.7910	.7939	.7967	.7995	.8023	.8051	.8078	.8106	.8133
0.9	.8159	.8186	.8212	.8238	.8264	.8289	.8315	.8340	.8365	.8389
1.0	.8413	.8438	.8461	.8485	.8508	.8531	.8554	.8577	.8599	.8621
1.1	.8643	.8665	.8686	.8708	.8729	.8749	.8770	.8790	.8810	.8830
1.2	.8849	.8869	.8888	.8907	.8925	.8944	.8962	.8980	.8997	.9015
1.3	.9032	.9049	.9066	.9082	.9099	.9115	.9131	.9147	.9162	.9177
1.4	.9192	.9207	.9222	.9236	.9251	.9265	.9279	.9292	.9306	.9319
1.5	.9332	.9345	.9357	.9370	.9382	.9394	.9406	.9418	.9429	.9441
1.6	.9452	.9463	.9474	.9484	.9495	.9505	.9515	.9525	.9535	.9545
1.7	.9554	.9564	.9573	.9582	.9591	.9599	.9608	.9616	.9625	.9633
1.8	.9641	.9649	.9656	.9664	.9671	.9678	.9686	.9693	.9699	.9706
1.9	.9713	.9719	.9726	.9732	.9738	.9744	.9750	.9756	.9761	.9767
2.0	.9772	.9778	.9783	.9788	.9793	.9798	.9803	.9808	.9812	.9817
2.1	.9821	.9826	.9830	.9834	.9838	.9842	.9846	.9850	.9854	.9857
2.2	.9861	.9864	.9868	.9871	.9875	.9878	.9881	.9884	.9887	.9890
2.3	.9893	.9896	.9898	.9901	.9904	.9906	.9909	.9911	.9913	.9916
2.4	.9918	.9920	.9922	.9925	.9927	.9929	.9931	.9932	.9934	.9936
2.5	.9938	.9940	.9941	.9943	.9945	.9946	.9948	.9949	.9951	.9952
2.6	.9953	.9955	.9956	.9957	.9959	.9960	.9961	.9962	.9963	.9964
2.7	.9965	.9966	.9967	.9968	.9969	.9970	.9971	.9972	.9973	.9974
2.8	.9974	.9975	.9976	.9977	.9977	.9978	.9979	.9979	.9980	.9981
2.9	.9981	.9982	.9982	.9983	.9984	.9984	.9985	.9985	.9986	.9986

# Standardizing a random variable

• Let X have mean  $\mu$  and variance  $\sigma^2 > 0$ 

• Let 
$$Y = \frac{X - \mu}{\sigma}$$
  $E[Y] = 0$   $Var(Y) = \frac{1}{\sigma^2} Var(x) = 1$ 

• If also X is normal, then:  $Y \sim \mathcal{N}(0, 1)$ 

#### Calculating normal probabilities

 Express an event of interest in terms of standard normal

	.00	.01	.02	.03	.04	.05	.06	.07	.08	.09
0.0	.5000	.5040	.5080	.5120	.5160	.5199	.5239	.5279	.5319	.5359
0.1	.5398	.5438	.5478	.5517	.5557	.5596	.5636	.5675	.5714	.5753
0.2	.5793	.5832	.5871	.5910	.5948	.5987	.6026	.6064	.6103	.6141
0.3	.6179	.6217	.6255	.6293	.6331	.6368	.6406	.6443	.6480	.6517
0.4	.6554	.6591	.6628	.6664	.6700	.6736	.6772	.6808	.6844	.6879
0.5	.6915	.6950	.6985	.7019	.7054	.7088	.7123	.7157	.7190	.7224
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2.4	.9918	.9920	.9922	.9925	.9927	.9929	.9931	.9932	.9934	.9936
2.5	.9938	.9940	.9941	.9943	.9945	.9946	.9948	.9949	.9951	.9952
2.6	.9953	.9955	.9956	.9957	.9959	.9960	.9961	.9962	.9963	.9964
2.7	.9965	.9966	.9967	.9968	.9969	.9970	.9971	.9972	.9973	.9974
2.8	.9974	.9975	.9976	.9977	.9977	.9978	.9979	.9979	.9980	.9981
2.9	.9981	.9982	.9982	.9983	.9984	.9984	.9985	.9985	.9986	.9986