

INF5820/INF9820

LANGUAGE TECHNOLOGICAL APPLICATIONS

Jan Tore Lønning, 2016 fall

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Probabilities 2

Today

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- **Some maths:**
 - ▣ Notation for sums and products
 - ▣ Exponential function and logarithms
- **More on probabilities**
 - ▣ Conditional probabilities and Bayes' formula
 - ▣ Random variables
 - ▣ Probability distributions – binomial distribution

Notation

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- $\sum_{i=1}^n a_i = a_1 + a_2 + a_3 + \dots + a_n$
- $\sum_{i=1}^7 i = 1 + 2 + 3 + 4 + 5 + 6 + 7 = 28$
- $\sum_{i=2}^5 i^2 = 4 + 9 + 16 + 25 = 54$
- $\prod_{i=1}^n a_i = a_1 * a_2 * a_3 * \dots * a_n$
- $\prod_{i=1}^7 i = 1 * 2 * 3 * 4 * 5 * 6 * 7 = 7! = 5040$

Exponentiation

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- Assume $a, b > 0$
- $a^m \cdot a^n = a^{(m+n)}$
- $a^{-m} = \frac{1}{a^m}$
- $(a^m)^n = a^{(m \cdot n)}$
- $a^{\left(\frac{1}{n}\right)} = \sqrt[n]{a}$ because
- $\left(a^{\left(\frac{1}{n}\right)}\right)^n = a^{\left(\frac{1}{n} \cdot n\right)} = a$

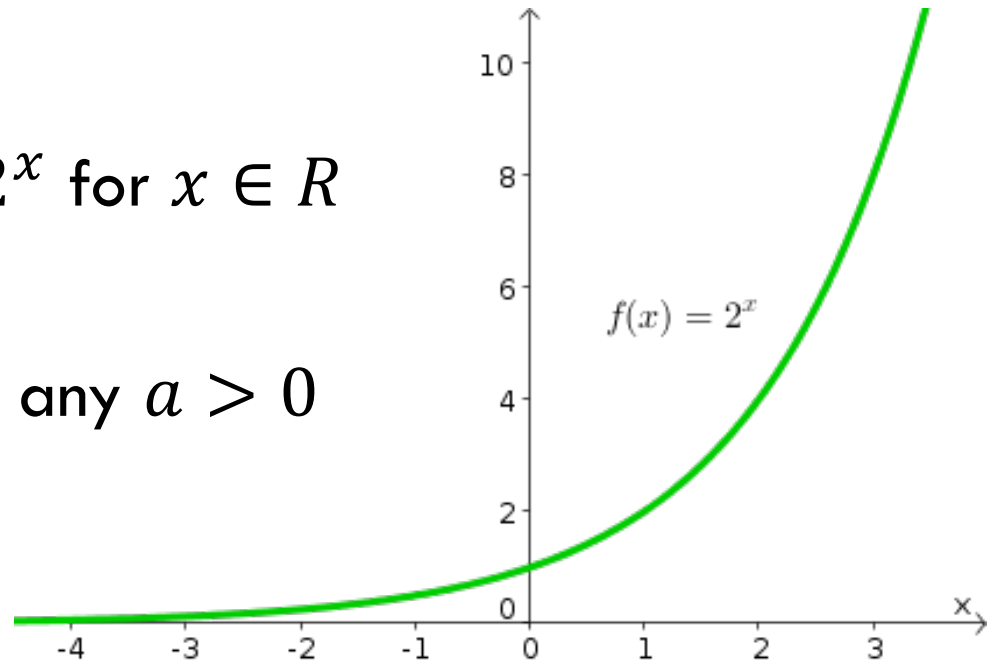
$$\prod_{i=1}^n a^{b_i} = a^{(\sum_{i=1}^n b_i)}$$

Exponential function

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0	1	2	3	4	5	6	7	8	9	10
1	2	4	8	16	32	64	128	256	512	1024

- 2^n for $n = 0, 1, 2, \dots$
- Extends to $f(x) = 2^x$ for $x \in \mathbb{R}$
- Also $f_a(x) = a^x$ for any $a > 0$

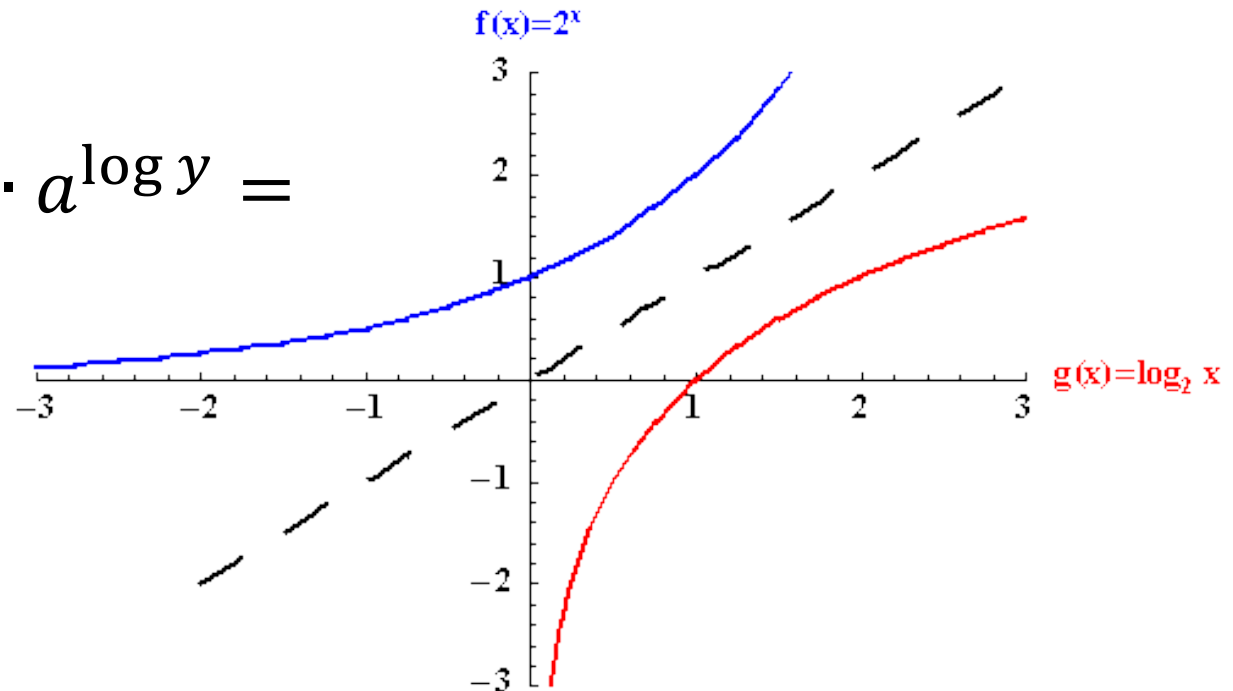


Logarithms

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- When $y = a^x$
- $x = \log_a y$
- Inverse function of exp
- $a^{\log_a x} = x$
- $x \cdot y = a^{\log x} \cdot a^{\log y} = a^{(\log x + \log y)}$

$$\prod_{i=1}^n x_i = \prod_{i=1}^n a^{\log x_i} = a^{(\sum_{i=1}^n \log x_i)}$$



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Basic concepts

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- **Random experiment** (or trial) (no: **forsøk**)
 - ▣ Observing a chance event
- **Outcomes** (**utfallene**)
 - ▣ The possible results of the experiment
- **Sample space** (**utfallsrommet**)
 - ▣ The set of all possible outcomes

Event

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- An **event** (**begivenhet**) is a set of elementary outcomes

	Experiment	Event	Formally
2	Rolling a dice	Getting 5 or 6	$\{5,6\}$
3	Flipping a coin three times	Getting at least two heads	$\{HHH, HHT, HTH, THH\}$
5	The first word you hear tomorrow	You hear a noun	$\{u \mid u \text{ is a noun}\}$
6	Throwing a dice until you get 6	An odd number of throws	$\{1,3,5, \dots\}$
7	The maximum temperature at Blindern	Between 20 and 22	$\{t \mid 20 \leq t \leq 22\}$

Probability measure, sannsynlighetsmål

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□ A probability measure P is a function from events to the interval $[0,1]$ such that:

1. $P(\Omega) = 1$

2. $P(A) \geq 0$

3. If $A \cap B = \emptyset$ then $P(A \cup B) = P(A) + P(B)$

□ And if A_1, A_2, A_3, \dots are disjoint, then
$$P\left(\bigcup_{j=1}^{\infty} A_j\right) = \sum_{j=1}^{\infty} P(A_j)$$

Conditional probability

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- $P(A \cap B)$
 - ▣ Both A and B happens
- **Conditional probability** (betinget sannsynlighet)

$$P(A | B) = \frac{P(A \cap B)}{P(B)}$$

- ▣ The probability of A happens if B happens
- **Multiplication rule** $P(A \cap B) = P(A | B)P(B) = P(B | A)P(A)$
- A and B are **independent** iff $P(A \cap B) = P(A)P(B)$

Bayes theorem

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$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

- Jargon:
 - ▣ $P(A)$ – prior probability
 - ▣ $P(A | B)$ – posterior probability
- Extended form

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)} = \frac{P(B | A)P(A)}{P(B | A)P(A) + P(B | -A)P(-A)}$$

Example

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- Disease: 1 out of 1000 are infected
- Test:
 - ▣ Detects 99% of the infected
 - ▣ 2% of the non-infected get a positive test
- Given a positive test: what is the chance you are ill?

What are probabilities?

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- Example throwing a dice:
 1. Classical view:
 - ▣ The six outcomes are equally likely
 2. Frequentist:
 - ▣ If you throw the dice many, many, many times, the number of 6s approach $16.6666\dots\%$
 3. Bayesian: subjective beliefs

Today

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- Some maths:
 - ▣ Notation for sums and products
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- More on probabilities
 - ▣ Conditional probabilities and Bayes' formula
 - ▣ **Random variables**
 - ▣ Probability distributions – binomial distribution

Random variable

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- A **variable** X in statistics is a property (feature) of an outcome of an experiment.
 - ▣ Formally it is a function from a sample space (**utfallsrom**) Ω to a **value space** Ω_X .
- When the value space Ω_X is numeric (roughly a subset of \mathbb{R}^n), it is called a **random variable**
- There are two kinds:
 - ▣ **Discrete random variables**
 - ▣ **Continuous random variables**
- A third type of variable: **categorical variable**, when Ω_X is nonnumeric
- Motivation for talking about random variables:
 - ▣ It is convenient to consider mathematical concepts, e.g. mean and variance
 - ▣ We can study their distribution abstracting away from what the actual outcomes

Examples

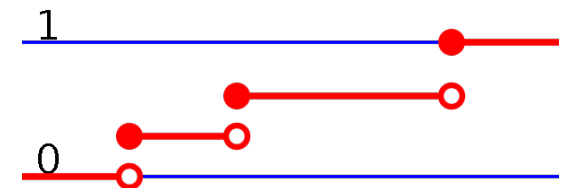
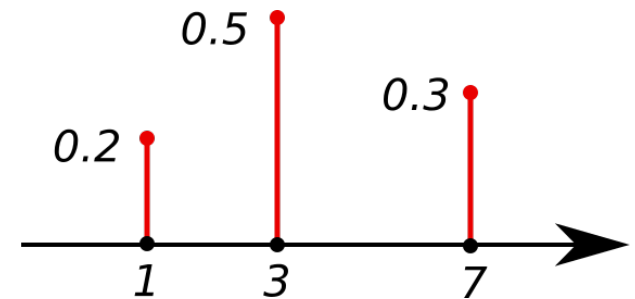
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1. Throwing two dice,
 - ▣ $\Omega = \{(1,1), (1,2), \dots, (1,6), (2,1), \dots, (6,6)\}$
 1. The number of 6s is a random variable X , $\Omega_X = \{0, 1, 2\}$
 2. The number of 5 or 6s is a random variable Y , $\Omega_Y = \Omega_X$
 3. The sum of the two dice, Z , $\Omega_Z = \{2, 3, \dots, 12\}$
2. A random person:
 1. X , the height of the person $\Omega_X = [0, 3]$ (meters)
 2. Y , the gender $\Omega_Y = \{0, 1\}$ (1 for female)
 - ▣ Ex 2.1 is continuous, the other are discrete

Discrete random variable

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- The value space is a finite or a countable infinite set of numbers $\{x_1, x_2, \dots, x_n, \dots\}$
- The **probability mass function**, pmf, p , which to each value yields
 - ▣ $p(x_i) = p(X=x_i) = P(\{\omega \in \Omega \mid X(\omega)=x\})$
- The **cumulative distribution function**, cdf,
 - ▣ $F(x_i) = p(X \leq x_i) = P(\{\omega \in \Omega \mid X(\omega) \leq x_i\})$

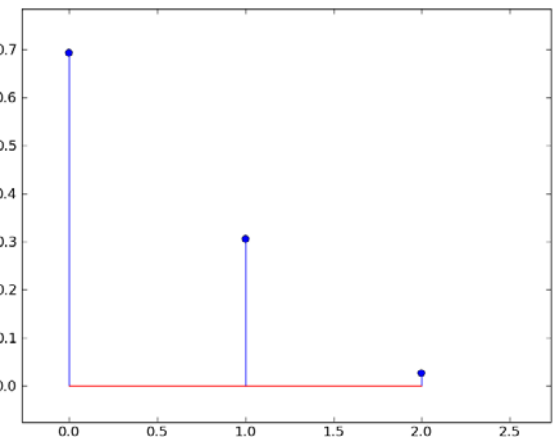
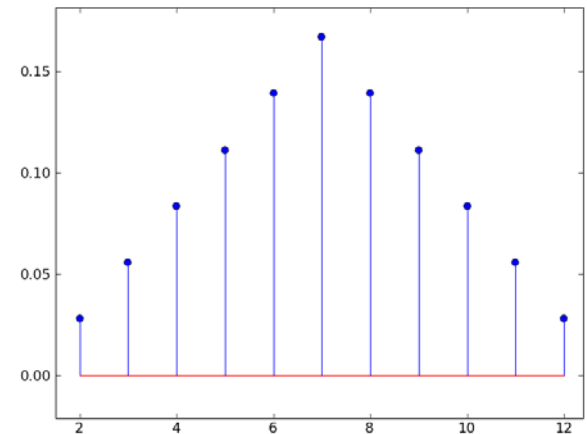


Diagrams: Wikipedia

Examples

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- Throwing two dice,
 - ▣ $\Omega = \{(1,1), (1,2), \dots, (1,6), (2,1), \dots, (6,6)\}$
 - ▣ (1.3) The sum of the two dice, Z ,
 $\Omega_Z = \{2, 3, \dots, 12\}$
 - $p_Z(2) = P(\{(1,1)\}) = 1/36$
 - $p_Z(7) = 6/36$
 - $F_Z(7) = 1+2+\dots+6 = 21/36$
 - ▣ (1.1) The number of 6s X , $\Omega_X = \{0, 1, 2\}$
 - $p_X(2) = P(\{(6,6)\}) = 1/36$
 - $p_X(1) = P(\{(6,x) \mid x \neq 6\}) + P(\{(x,6) \mid x \neq 6\}) = 10/36$
 - $p_X(0) = 25/36$



Mean – example

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- Throwing two dice, what is the mean value of their sum?
- $(2+3+\dots+7+$
 $3+4+\dots+8+$
 $4+\dots+9+$
 $5+\dots+10+$
 $6+\dots+11+$
 $7+\dots+12)/36=$
- $(2+2*3+3*4+4*5+5*6+6*7+5*8+\dots+2*11+12)/36=$
- $(1/36)2+(2/36)*3+(3/36)*4+\dots+(1/36)*12 =$
- $p(2)*2 + p(3)*3 + p(4)*4 + \dots p(12)*12 =$
- $\sum p(x)*x$

Mean of a discrete random variable

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- The **mean** (or **expectation**) (**forventningsverdi**) of a discrete random variable X :

$$\mu_X = E(X) = \sum_x p(x)x$$

- Useful to remember

$$\mu_{(X+Y)} = \mu_X + \mu_Y$$

$$\mu_{(a+bX)} = a + b\mu_x$$

Examples:
One dice: 3.5
Two dices: 7
Ten dices: 35

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Examples of distributions

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- (1.3) The sum of the two dice, Z , i.e.

- $p_Z(2) = 1/36, \dots, p_Z(7) = 6/36$ etc

- (3.2) p_2 given by:

- $p_2(7)=1$

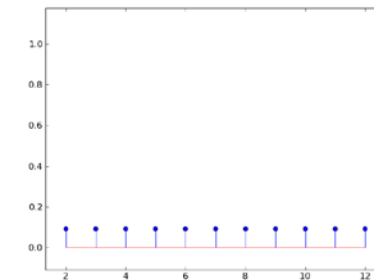
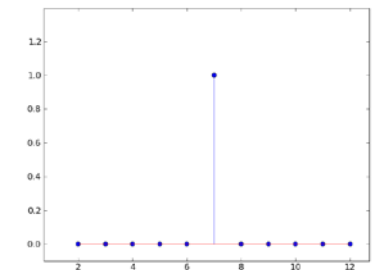
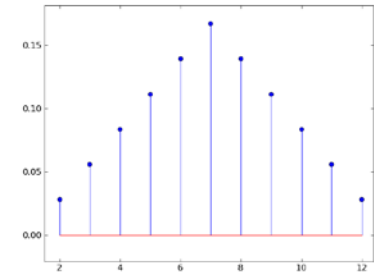
- $p_2(x)=0$ for $x \neq 7$

- (3.3) p_3 given by:

- $p_3(x)=1/11$ for $x = 2,3,\dots,12$

-

The three have the same mean but are very different



Think about

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- Flip a coin 10 times, count the number of heads
- You expect 5 heads, but not exactly 5
 - ▣ 6 is OK
- When do you start to worry whether the coin is unfair?
 - ▣ 8 heads?
 - ▣ 9 heads?

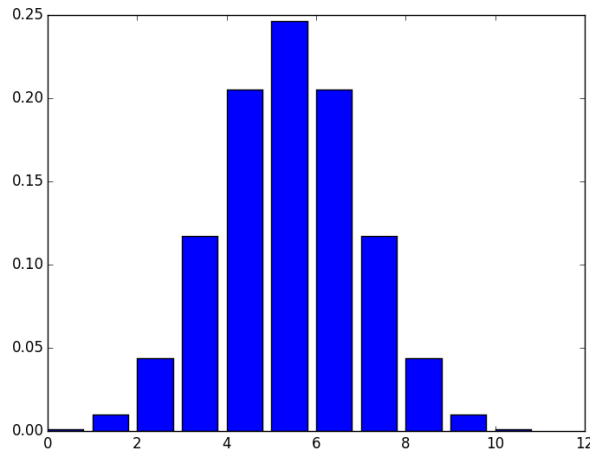
- This is the task for inferential statistics

Tossing a fair(?) coin

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- The cumulative distribution function:
``How likely is it to get N or fewer tails?``

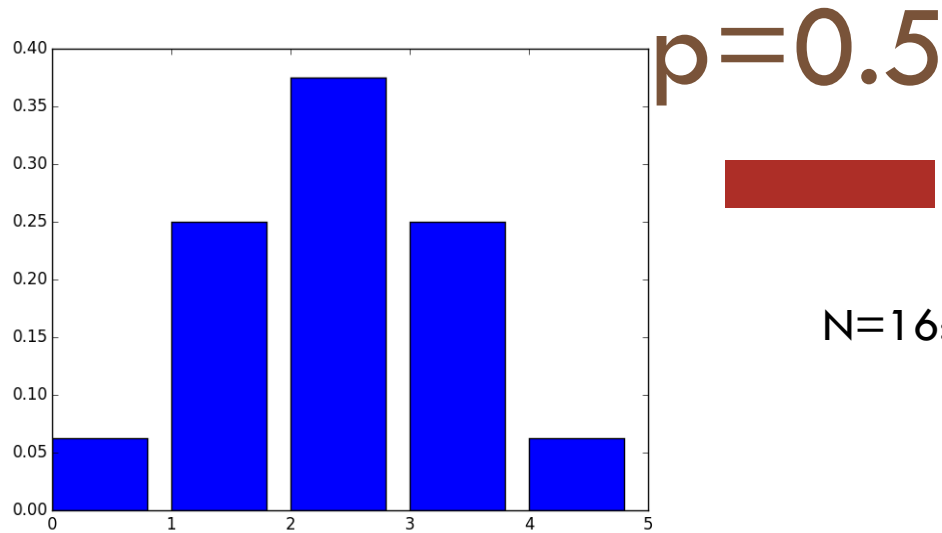
10:



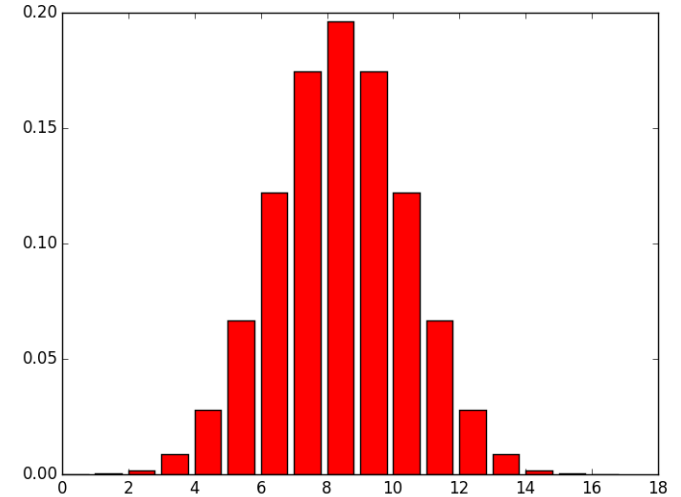
N	pmf(N)	cdf(N)
0	0.001	0.001
1	0.010	0.011
2	0.044	0.055
3	0.117	0.172
4	0.205	0.377
5	0.246	0.623
6	0.205	0.828
7	0.117	0.945
8	0.044	0.989
9	0.010	0.999
10	0.001	1.000

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N=4:

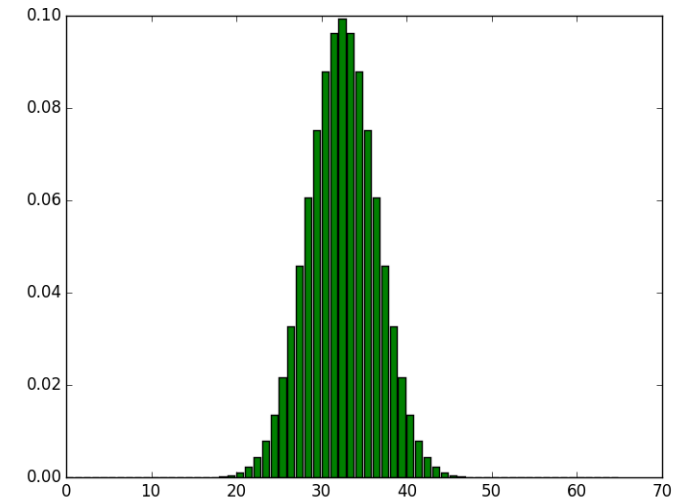


N=16:



N	1	4	16	64	256
σ^2	0.25	1	4	16	64
σ	0.5	1	2	4	8

N=64:



- The relative variation gets smaller with growing N
- The pmf graph approaches a bell shape

Bernoulli trial

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- One experiment, two outcomes
- $\Omega_x = \{0, 1\}$
- Write p for $p(1)$
- Then $p(0) = 1 - p$

Examples:

- Flipping a fair coin, $p = 1/2$
- Rolling a dice, getting a 6, $p = 1/6$

- The mean/expectation: $0 * p(0) + 1 * p(1) = 0 + p = p$

Sampling

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Ordered sequences:

- Choose k items from a population of n items with replacement: n^k
- Without replacement (permutation):
 - ▣ : $n(n-1)(n-2)\dots(n-k+1) = \frac{n!}{(n-k)!}$

Unordered sequences

- Without replac.: $\frac{1}{k!} \left(\frac{n!}{(n-k)!} \right) = \left(\frac{n!}{k!(n-k)!} \right) = \binom{n}{k}$
 - ▣ = the number of ordered sequences/
The number of ordered sequences containing the same k elements

Binomial distribution

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- **Binomial distribution** (binomisk fordeling)
- Conducting n Bernoulli trials with the same probability and counting the number of successes

- Example flipping a fair coin n times, $p(k)$:
 - $n=2$: $p(0)=1/4$, $p(1)=1/2$, $p(2)=1/4$
 - $n=3$: $p(0)=1/8$, $p(1)=3/8$, $p(2)=3/8$, $p(3)=1/8$
 - $n=4$: $(1,4,6,4,1)/16$
 - $n=5$: $(1,5,10,5,1)/32$

- n :
$$p(k) = \binom{n}{k} \left(\frac{1}{2}\right)^n$$

where

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

□

Binomial distribution

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- **Binomial distribution** (binomisk fordeling)
- **General form:**
 - $0 < p < 1$
 - n a natural number

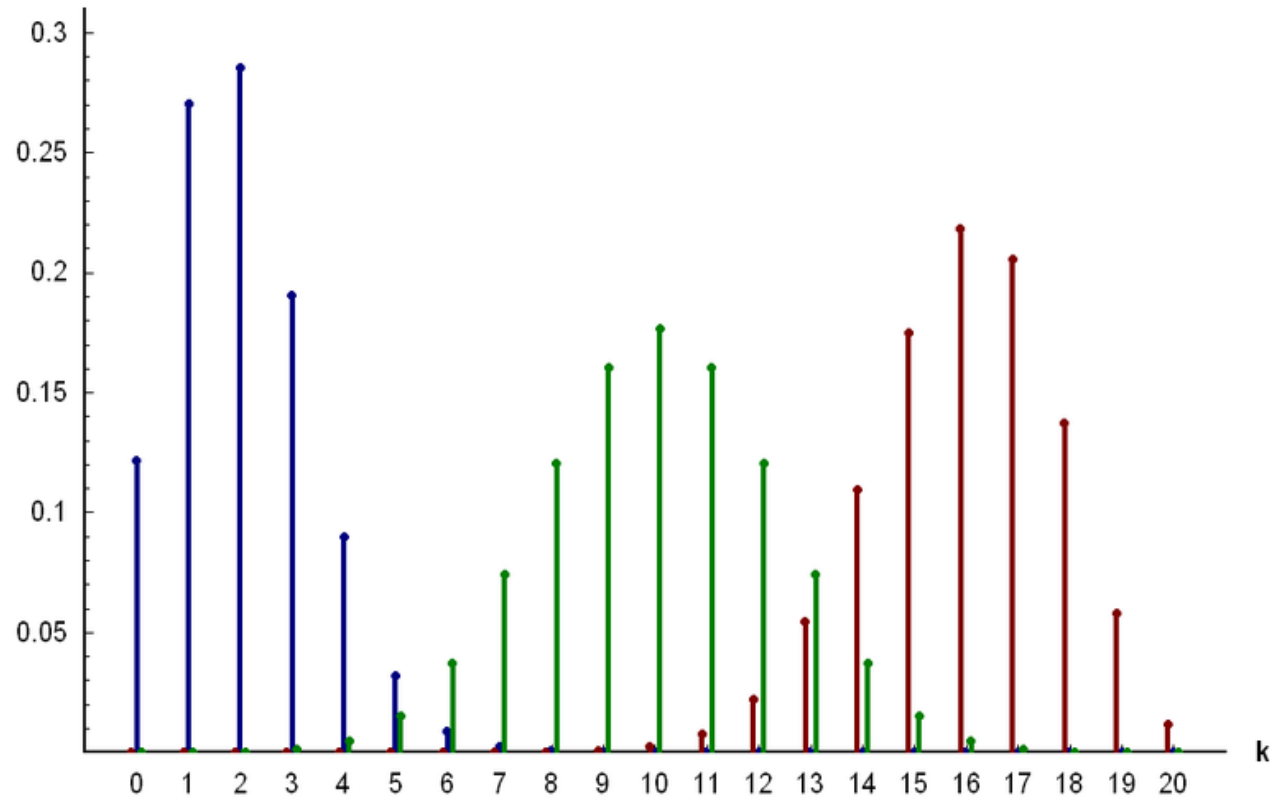
- **B(n,p)** is given by $b(k; n, p) = \binom{n}{k} p^k (1-p)^{(n-k)}$

for $k = 0, 1, \dots, n$, where
$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

Binomial distribution

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Wahrscheinlichkeit



- $n = 20$
- $p = 0.1$ (blue), $p = 0.5$ (green) and $p = 0.8$ (red)