

MATH 180A (Lecture A00)

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Today: Gaussian (Normal) distribution
Normal approximation

Next: ASV 4.1

Week 6:

- Homework 4 due Friday, February 17

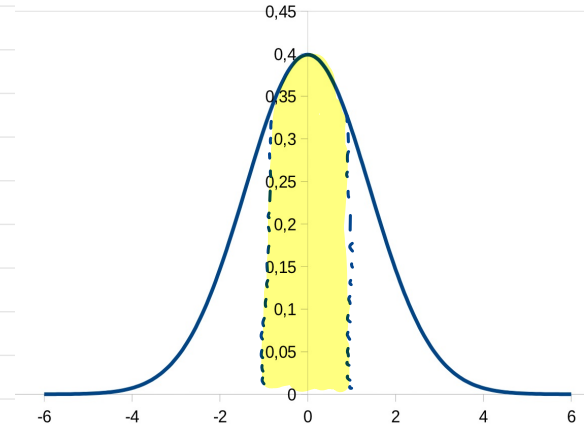
CDF of $N(0,1)$

Suppose $X \sim N(0,1)$. What is $P(|X| \leq 1)$?

$$P(-1 \leq X \leq 1)$$

$$= \int_{-1}^1 \varphi(t) dt = \frac{1}{\sqrt{2\pi}} \int_{-1}^1 e^{-t^2/2} dt$$

Cannot use the polar coordinate trick.



$$\Phi(x) := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{t^2}{2}} dt - \text{CDF of } X \sim N(0,1)$$

- no simple explicit formula
- table of values of $\Phi(x)$ (for $x \geq 0$)

Normal table of values (Appendix E in textbook)

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	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
0.2	0.5793	0.5832	0.5871	0.5910	0.5948	0.5987	0.6026	0.6064	0.6103	0.6141
0.3	0.6179	0.6217	0.6255	0.6293	0.6331	0.6368	0.6406	0.6443	0.6480	0.6517
0.4	0.6554	0.6591	0.6628	0.6664	0.6700	0.6736	0.6772	0.6808	0.6844	0.6879
0.5	0.6915	0.6950	0.6985	0.7019	0.7054	0.7088	0.7123	0.7157	0.7190	0.7224
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	0.7852
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

This table gives $P(Z \leq z)$ where $Z \sim N(0,1)$, $z = x_i + y_j$

Example $\Phi(0.91) = P(Z \leq 0.91) = P(Z \leq 0.9 + 0.01) \approx 0.8186$

Fact:

$$P(Z > 0.24) =$$

$$P(-0.28 < Z < 0.59) =$$

Normal table of values

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	0.5359
0.1	0.5398	0.5438	0.5478	0.5517	0.5557	0.5596	0.5636	0.5675	0.5714	0.5753
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0.7	0.7580	0.7611	0.7642	0.7673	0.7704	0.7734	0.7764	0.7794	0.7823	0.7852
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

Exercise Let $Z \sim N(0, 1)$

Find $x_0 \in \mathbb{R}$ such that $P(|Z| > x_0) \approx 0.704$

$$P(|Z| > x_0) =$$

Mean and variance of $X \sim N(0,1)$

$$E(X) = \int_{-\infty}^{+\infty} t f_X(t) dt =$$

$$\text{Var}(X) = E(X^2) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} t^2 e^{-\frac{t^2}{2}} dt$$

General normal distribution $N(\mu, \sigma^2)$

Def Let $\mu \in \mathbb{R}$ and $\sigma > 0$. Random variable X has normal (Gaussian) distribution with mean μ and variance σ^2 if the PDF of X is given by

$$f_X(x) =$$

We write

Using the density we can compute

$$E(X) = \quad , \quad \text{Var}(X) =$$

"Gaussian distribution" = family of distributions

Relation between $X \sim N(\mu, \sigma)$ and $Z \sim N(0, 1)$

Proposition Let $X \sim N(\mu, \sigma^2)$, $a \neq 0$, $b \in \mathbb{R}$.

Then

Using this proposition any Gaussian random variable can be written as a shifted and rescaled standard normal.

E.g., if $\sigma > 0$, $\mu \in \mathbb{R}$ and $Z \sim N(0, 1)$, then

If $X \sim N(\mu, \sigma^2)$, then $E(X) =$; $\text{Var}(X) =$

If $X \sim N(\mu, \sigma^2)$, then

Example

Let $X \sim N(-3, 4)$

Find $P(X < 0.91)$; $P(X > 0.82)$; $P(-0.24 < X < 0.88)$

If $X \sim N(-3, 4)$, then , so

$$P(X < 0.91) =$$

$$P(-0.24 < X < 0.88) =$$

The message :

If we have independent and identically distributed random variables X_1, X_2, \dots, X_n with

$E(X_i) = \mu$, $\text{Var}(X_i) = \sigma^2$, then for any $a < b$

Today: $X_i \sim \text{Ber}(p)$; Last lecture: general case

CLT for Bernoulli distribution (approximation of Bin)

If $X_i \sim \text{Ber}(p)$ are independent, then $X_1 + \dots + X_n \sim \text{Bin}(n, p)$

$$E(X_1) = \quad \text{Var}(X_1) =$$

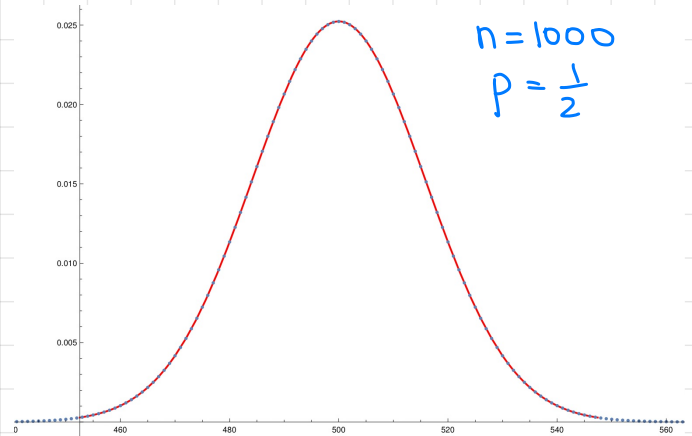
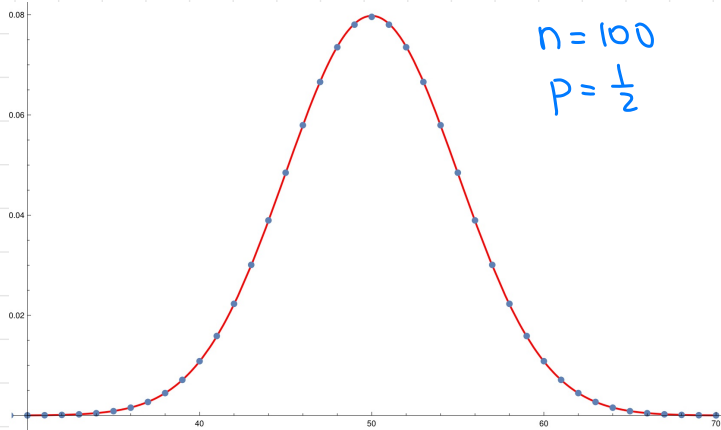
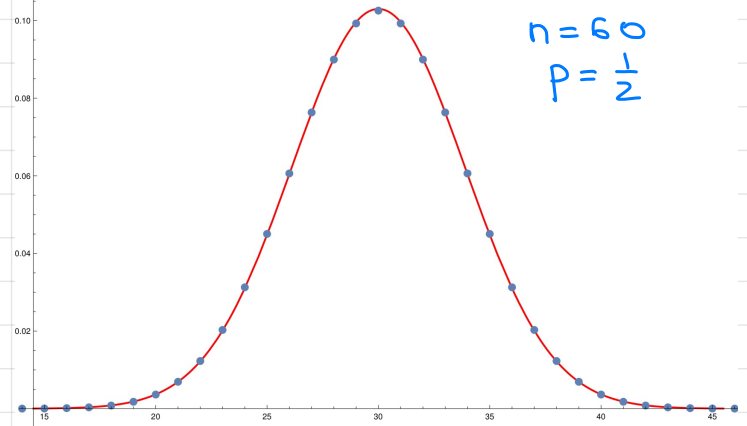
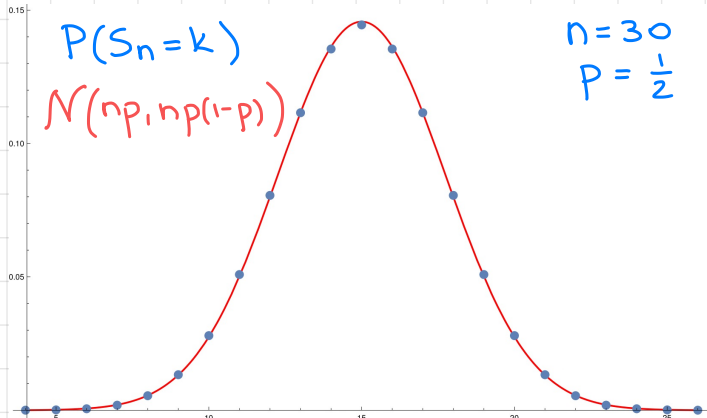
CLT for Bernoulli distribution:

Let $S_n \sim \text{Bin}(n, p)$, let $a < b$. Then

We can rewrite (*) using $\bar{S}_n := \frac{S_n}{n}$

CLT, approximation of Binomial distribution

Some numerics



Normal approximation. 3-sigma rule

We use the approximation of Bin(n,p) by the normal distribution if

In this case we can take

$$P\left(a \leq \frac{S_n - np}{\sqrt{np(1-p)}} \leq b\right) \approx \Phi(b) - \Phi(a)$$

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	0.5359
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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	0.9685	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	0.9924	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	0.9958	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	0.9973	0.9974
2.8	0.9974	0.9975	0.9976	0.9977	0.9977	0.9978	0.9979	0.9979	0.9980	0.9981
2.9	0.9981	0.9982	0.9982	0.9983	0.9984	0.9984	0.9985	0.9985	0.9986	0.9986

In particular, in this case

- $P(|S_n - np| < 1) \approx \Phi(1) - \Phi(-1) = 2\Phi(1) - 1 = 0.68$
- $P(|S_n - np| < 2) \approx \Phi(2) - \Phi(-2) = 2\Phi(2) - 1 = 0.95$
- $P(|S_n - np| < 3) \approx \Phi(3) - \Phi(-3) = 2\Phi(3) - 1 = 0.99$

CLT. Examples

Flipping a fair coin 10000 times

X = number of tails

Find (approximately) $P(4950 \leq X \leq 5050)$

$$X \sim \text{Bin}(10000, \frac{1}{2})$$

$$E(X) =$$

$$\sigma(X) =$$

$$P(4950 \leq X \leq 5050) =$$

Z	0.00	0.01	0.02
0.0	0.5000	0.5040	0.5080
0.1	0.5398	0.5438	0.5478
0.2	0.5793	0.5832	0.5871
0.3	0.6179	0.6217	0.6255
0.4	0.6554	0.6591	0.6628
0.5	0.6915	0.6950	0.6985
0.6	0.7257	0.7291	0.7324
0.7	0.7580	0.7611	0.7642
0.8	0.7881	0.7910	0.7939
0.9	0.8159	0.8186	0.8212
1.0	0.8413	0.8438	0.8461
1.1	0.8643	0.8665	0.8686

CLT. Examples

You win \$9 with probability $\frac{1}{20}$, lose \$1 with prob. $\frac{19}{20}$

Approximate the probability that you lost < 100 \$ after 400 games.

Denote by X the number of wins after 400 games

$$X \sim \text{Bin}(400, \frac{1}{20}). \quad n \cdot p \cdot (1-p) =$$

Total winnings after 400 games:

We have to compute

$$P(9X - (400 - X) > -100) =$$

Law of Large Numbers

Let X_1, X_2, \dots, X_n be independent and identically distributed, and let $E(X_1) = \mu \in \mathbb{R}$. Then

for any $\varepsilon > 0$

In particular, for $X_1 \sim \text{Ber}(p)$