

Last lecture

Poisson process ([Ch 3.5](#))

- Motivation
- Bernoulli process to Poisson process
- Definition
- Properties

Agenda

Gaussian (normal) Distribution (Ch 3.6.2)

- Motivation and Definition
- Examples

The Central Limit Theorem and Gaussian Approximation (Ch 3.6.3)

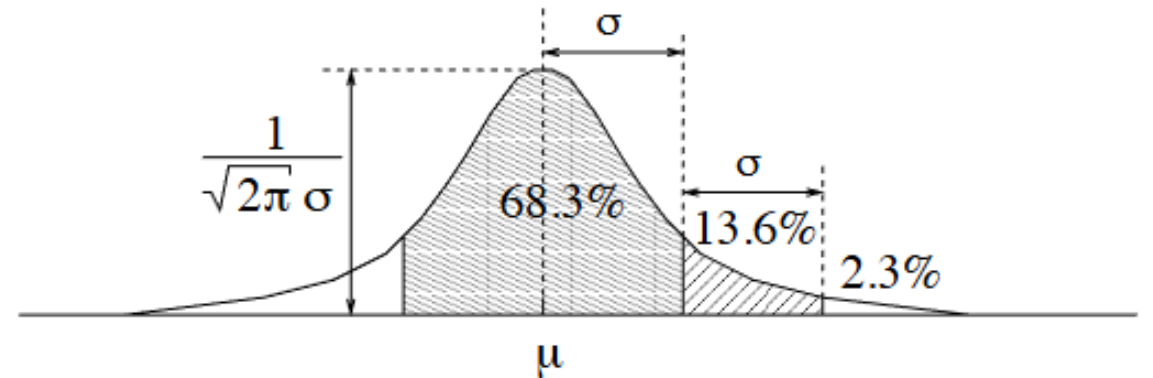
- Definition
- CDF Approximation
- Examples

Gaussian (Normal) Distribution

Definition

A normal distribution is defined by μ_X and σ_X^2 , Let $X \sim N(\mu, \sigma^2)$

- $f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$
- Usage – model “Sum of many small independent events”
- E.g. Sum of many binomial distributions



Standard normal distribution

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right)$$

$$X \sim N(\mu = 0, \sigma^2 = 1) \sim N(0,1)$$

- $\Phi(u) \triangleq F_X(u) =$
- $Q(u) = 1 - \Phi(u)$
- Pre-computed tables!

Φ and Q tables

z	-0.00	-0.01	-0.02	-0.03	-0.04	-0.05	-0.06	-0.07	-0.08	-0.09
-3.9	0.00005	0.00005	0.00004	0.00004	0.00004	0.00004	0.00004	0.00004	0.00003	0.00003
-3.8	0.00007	0.00007	0.00007	0.00006	0.00006	0.00006	0.00006	0.00005	0.00005	0.00005
-3.7	0.00011	0.00010	0.00010	0.00010	0.00009	0.00009	0.00008	0.00008	0.00008	0.00008

⋮

-0.1	0.46017	0.45620	0.45224	0.44828	0.44433	0.44038	0.43644	0.43251	0.42858	0.42465
-0.0	0.50000	0.49601	0.49202	0.48803	0.48405	0.48006	0.47608	0.47210	0.46812	0.46414
z	-0.00	-0.01	-0.02	-0.03	-0.04	-0.05	-0.06	-0.07	-0.08	-0.09

z	+ 0.00	+ 0.01	+ 0.02	+ 0.03	+ 0.04	+ 0.05	+ 0.06	+ 0.07	+ 0.08	+ 0.09
0.0	0.50000	0.50399	0.50798	0.51197	0.51595	0.51994	0.52392	0.52790	0.53188	0.53586
0.1	0.53983	0.54380	0.54776	0.55172	0.55567	0.55962	0.56360	0.56749	0.57142	0.57535

⋮

3.8	0.99993	0.99993	0.99993	0.99994	0.99994	0.99994	0.99994	0.99995	0.99995	0.99995
3.9	0.99995	0.99995	0.99996	0.99996	0.99996	0.99996	0.99996	0.99996	0.99997	0.99997
z	+0.00	+0.01	+0.02	+0.03	+0.04	+0.05	+0.06	+0.07	+0.08	+0.09

Scaling the Gaussian RV

$$f(u) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{u^2}{2}\right)$$

$$X \sim N(\mu = 0, \sigma^2 = 1) \sim N(0,1)$$

- $Y = \sigma X + \mu$
- $f_Y(y) =$

Examples

Given $\Phi(u) \triangleq F_X(u)$ and $Q(u) = 1 - \Phi(u)$ for $X \sim N(0,1)$

- Let $Y = N(\mu = 10, \sigma^2 = 16)$
- Find $P\{Y > 15\}$, $P\{Y \leq 5\}$, $P\{Y^2 \geq 400\}$ and $P\{Y = 2\}$ in terms of Φ or Q

Examples

Suppose $\mu_X = 10$ and $\sigma_X^2 = 3$. Compute $P\{X < 10 - \sqrt{3}\}$ if

- X is a Gaussian RV in terms of Q
- X is a uniform RV

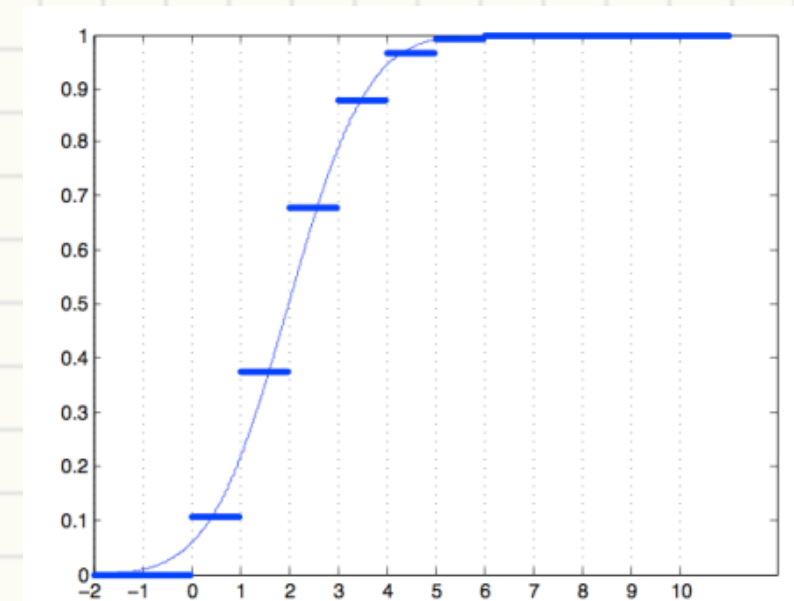
(Hint: $10 - \sqrt{3} \approx 8.27$)

Central Limit Theorem and Gaussian Approximation

Central Limit Theorem (CLT)

If many **independent** random variables are added together, and if each of them is **small** in magnitude compared to the sum, then the **sum** X has an approximately **Gaussian** distribution $\tilde{X}$.

- $P\{X \leq v\} \approx P\{\tilde{X} \leq v\}$
- E.g., $X \sim \text{Binomial}(n, p)$ when np and $n(1 - p)$ are not small
 - $(n, p) = (10, 0.2)$
 - $\mu_X =$
 - $\sigma_X^2 =$
- What if np is small?



Gaussian Approximation - 1

Approximate $X \sim \text{Bin}(10, 0.2)$ with $\tilde{X} \sim N(2, 1.6)$

- $F_X(2) = F_X(2.1) = F_X(2.9)$
- $F_{\tilde{X}}(2) \neq F_{\tilde{X}}(2.9)$
- How should we approximate?

- $P\{X \leq k\} \approx$
- $P\{X < k\} \approx$
- $P\{X \geq k\} \approx$
- $P\{X > k\} \approx$

