

Discrete distributions worksheet

Name of special distribution (X)	Properties	What X represents and its range (R_X)	Parameters	Probability $P(X = x)$	Expected value $E(X)$	Variance $Var(X)$
Binomial distribution	<ul style="list-style-type: none"> You perform n identical experiments and you know n before starting Each experiment or trial independent of each other The probability of success, p (or seeing a particular event) is same for every trial 	<p>The number of times success is observed</p> <p>$R_X: \{0, 1, 2, \dots, n\}$</p>	n, p	$\binom{n}{x} p^x (1-p)^{n-x}$	np	$np(1-p)$
Hypergeometric distribution	<ul style="list-style-type: none"> I sample k objects from a finite population with b success and r failures Each sample of k objects is equally likely to be chosen 	<p>The number successes observed</p> <p>$R_X: \{0, 1, 2, \dots, \min\{b, k\}\}$</p>	k, b, r	$\frac{\binom{b}{x} \binom{r}{k-x}}{\binom{b+r}{k}}$	$\frac{kb}{b+r}$	$k * \frac{b}{b+r} * \frac{r}{b+r} * \frac{b+r-k}{b+r-1}$
Negative binomial distribution	<ul style="list-style-type: none"> You perform identical experiments and you do not know n before starting, rather you keep performing experiments until you observe m-th success Each experiment or trial independent of each other The probability of success, p is same for every trial 	<p>The number of trials before m-th success is observed</p> <p>$R_X: \{m, m+1, \dots, \infty\}$</p>	m, p	$\begin{cases} \binom{x-1}{m-1} p^m (1-p)^{x-m} \\ 0 \end{cases}$	$\frac{m}{p}$	$\frac{m(1-p)}{p^2}$
Poisson distribution	<ul style="list-style-type: none"> An event occurs at an average rate of λ such that: Occurrences of the event are independent of each other More than one of these events cannot occur simultaneously 	<p>The number of events</p> <p>$R_X: \{0, 1, 2, \dots, \infty\}$</p>	λ	$\frac{e^{-\lambda} \lambda^x}{x!}$	λ	λ

Continuous distributions worksheet

Name of special distribution (X)	Properties	Range (R_X) and Parameters	PDF (it is not probability $X=x$) $f(x)$	CDF $F(x) = P(X \leq x)$	Expected value $E(X)$	Variance $Var(X)$
Uniform distribution	<ul style="list-style-type: none"> If all possible values of X are equally likely to occur And, all values fall in the range $[a, b]$ 	$R_X: [a, b]$ Parameters: a, b	$f(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$	$F(x) = \begin{cases} 0 & x \leq a \\ \frac{x-a}{b-a} & a \leq x \leq b \\ 1 & x \geq b \end{cases}$	$(b+a)/2$	$\frac{(b-a)^2}{12}$
Exponential distribution	<ul style="list-style-type: none"> X represents the waiting time between Poisson distribution events X has property of being memoryless which means $P(X \geq x + x_0 X \geq x_0) = P(X \geq x)$ 	$R_X: [0, \infty)$ Parameters: λ	$f(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$	$F(x) = 1 - e^{-\lambda x}$	$1/\lambda$	$\frac{1}{\lambda^2}$
Normal distribution	<ul style="list-style-type: none"> When the pdf of X has a bell curve shape 	$R_X: (-\infty, \infty)$ Parameters: μ, σ	$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/2\sigma^2}$	Use z-table or R	μ	σ^2
Gamma distribution	<ul style="list-style-type: none"> General case of Exponential distribution 	$R_X: [0, \infty)$ Parameters: α, λ	$f(x) = \begin{cases} \frac{1}{\Gamma(\alpha)} \lambda^\alpha x^{\alpha-1} e^{-\lambda x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$	Use R	α/λ	$\frac{\alpha}{\lambda^2}$

We will learn about two more continuous distributions later in the course: t-distribution and chi-squared distribution

$\Gamma(\cdot)$ is the Gamma function with properties: (a) $\Gamma(\alpha) = (\alpha-1)\Gamma(\alpha-1)$ (b) $\Gamma(n) = (n-1)!$ if n is positive integer (c) $\Gamma(1/2) = \sqrt{\pi}$

Given: X_1, X_2, \dots, X_n are random samples i.e. they are **independent and identically distributed**

A confidence interval (CI) gives a plausible range for a **population** parameter p by using outcomes from a sample

A prediction interval (PI) gives a plausible range for a **single future prediction value**

A tolerance interval (TI) gives a plausible range which **contains at least k%** of the entire population

Assumption	X_i 's are from normal distribution with unknown mean and unknown variance	X_i 's are from normal distribution with unknown mean and unknown variance	X_i 's are from normal distribution with unknown mean and unknown variance and $n < 40$
Interval	Confidence interval on population variance	Prediction interval for X of a single individual	Tolerance interval containing at least $k\%$ of population
Notations	n : $\chi^2_{\alpha, n-1}$: s :	\bar{x} : n : s : $t_{\frac{\alpha}{2}, n-1}$:	\bar{x} : n : s : $C_{\alpha, k}$:
$(1 - \alpha)100\%$ two-sided interval	Lower limit: $(n - 1)s^2 / \chi^2_{\frac{\alpha}{2}, n-1}$ Upper limit: $(n - 1)s^2 / \chi^2_{1-\frac{\alpha}{2}, n-1}$	Lower limit: $\bar{x} - t_{\frac{\alpha}{2}, n-1} * s * \sqrt{1 + \frac{1}{n}}$ Upper limit: $\bar{x} + t_{\frac{\alpha}{2}, n-1} * s * \sqrt{1 + \frac{1}{n}}$	Lower limit: $\bar{x} - C_{\alpha, k} s$ Upper limit: $\bar{x} + C_{\alpha, k} s$ (two sided C values)
$(1 - \alpha)100\%$ one-sided upper bound interval	Lower limit: $-\infty$ Upper limit: $(n - 1)s^2 / \chi^2_{1-\alpha, n-1}$	Lower limit: $-\infty$ Upper limit: $\bar{x} + t_{\alpha, n-1} * s * \sqrt{1 + \frac{1}{n}}$	Lower limit: $-\infty$ Upper limit: $\bar{x} + C_{\alpha, k} s$ One sided C value
$(1 - \alpha)100\%$ one-sided lower bound interval	Lower limit: $(n - 1)s^2 / \chi^2_{\alpha, n-1}$ Upper limit: $+\infty$	Lower limit: $\bar{x} - t_{\alpha, n-1} * s * \sqrt{1 + \frac{1}{n}}$ Upper limit: $+\infty$	Lower limit: $\bar{x} - C_{\alpha, k} s$ Upper limit: $+\infty$ One sided C value

Population Parameter	Population mean
Null hypothesis	$\mu = \mu_0$
Test statistic	$\frac{x-\mu_0}{s/\sqrt{n}}$ OR $\frac{x-\mu_0}{\sigma/\sqrt{n}}$ based on if population variance is known
Case 1	$n > 40$ OR $n < 40$ AND population distribution is normal AND population variance is known
Case 2	$n < 40$ AND population distribution is normal AND variance not known
Otherwise	Consult a knowledgeable statistician

Alternate Hypothesis test type	Rejection region Case 1	Rejection region Case 2
$H_a: \mu > \mu_0$ Or $H_a: p > p_0$	(z_α, ∞)	$(t_{\alpha, n-1}, \infty)$
$H_a: \mu < \mu_0$ Or $H_a: p < p_0$	$(-\infty, -z_\alpha)$	$(-\infty, -t_{\alpha, n-1})$
$H_a: \mu \neq \mu_0$ Or $H_a: p \neq p_0$	$(-\infty, -z_{\frac{\alpha}{2}}) \cup (z_{\frac{\alpha}{2}}, \infty)$	$(-\infty, -t_{\frac{\alpha}{2}, n-1}) \cup (t_{\frac{\alpha}{2}, n-1}, \infty)$

Reject if test statistic belongs to rejection region

Population Parameter	Sample proportion
Null hypothesis	$p = p_0$
Test statistic	$\frac{p' - p_0}{\sqrt{p_0(1-p_0)/n}}$
Case 1	$np_0 > 10$ AND $n(1-p_0) > 10$
Otherwise	Consult a knowledgeable statistician

Hypothesis test type	P-values Case 1
$H_a: \mu > \mu_0$ Or $H_a: p > p_0$	$1 - \phi(z)$
$H_a: \mu < \mu_0$ Or $H_a: p < p_0$	$\phi(z)$
$H_a: \mu \neq \mu_0$ Or $H_a: p \neq p_0$	$2[1 - \phi(z)]$

Reject if $\alpha > P$ value