

Recall: . sampling with replacement — Bernoulli trials, independence
 . Sampling without replacement : $m < r < N$ Hypergeometric (N, r, m)
 $S = \{0, \dots, m\}$

$$IP(\{k\}) = \frac{\binom{r}{k} \binom{N-r}{m-k}}{\binom{N}{m}} = \frac{\frac{r!}{k!} \frac{N-r!}{m-k!} \frac{1}{r-k!} \cdot \frac{1}{N-r-(m-k)!}}{\frac{N!}{m!(N-m)!}}$$

$$IP(\{k\}) = \frac{m!}{k! (m-k)!} \frac{\frac{r(r-1)(r-2) \dots (r-(k-1))}{N(N-1) \dots (N-(k-1))} \frac{(N-r)(N-r-1) \dots (N-r-(m-k))}{(N-k) \dots (N-(m-1))}}{\frac{N!}{m!(N-m)!}}$$

$$p = \frac{r}{N}, \quad p_1 = \frac{r-k}{N-k}, \quad p_2 = \frac{r-k}{N-m}$$

$$\binom{m}{k} p_1^k (1-p_2)^{m-k} \leq IP(\{k\}) \leq \binom{m}{k} p^k (1-p_1)^{m-k}$$

Theorem 2.3.3 — proof of this is done above.

Conclusion: - $m \ll N, m \ll r \Rightarrow p_1 \approx p_2 \approx p$
 $IP(\{k\}) \approx \binom{m}{k} p^k (1-p)^{m-k}$

$\Rightarrow$ Sampling with replacement $\approx$ Sampling without replacement

3 Discrete Random Variables :-

- Various Experiments - giving rise to Bernoulli trials
 - Each time we created a new sample space, for a new distribution.
 - Can we do all this on one sample space.
- To do this :- Define functions on the space that describe the output of interest
- Random Variables -

Example 3.1.1 : Suppose a coin is flipped 3 times.

Q1: # of heads in 3 tosses?

Q2: Trial at which 1st head shows up?

$$S = \{ hhh, hht, hth, htt, thh, tht, tth, ttt \}$$

$X: S \rightarrow \{0, 1, 2, 3\}$ by $X(s) = \# \text{ of heads in } s$

$Y: S \rightarrow \{1, 2, 3, \text{None}\}$ by $Y(s) = \text{Trial at which head appears in } s$

s	$X(s)$	$Y(s)$
hhh	3	1
hht	2	1
hth	2	1
htt	1	1
thh	2	2
tht	1	2
tth	1	3
ttt	0	None

$$\begin{aligned} \mathbb{P}(2 \text{ heads in 3 tosses}) &= \mathbb{P}(X^{-1}(\{2\})) = \mathbb{P}(\{hht, hth, tht\}) \\ &= 3/8 \end{aligned}$$

$$\begin{aligned}
 \mathbb{P}(\text{first head occurs in 1st toss}) &= \mathbb{P}(\bar{Y}^{-1}(\{1\})) \\
 &= \mathbb{P}(\{hhh, hht, hth, htt\}) \\
 &= 4/8
 \end{aligned}$$

Similarly one can calculate:

$$\begin{aligned}
 \bullet \quad \mathbb{P}(X=0) &= \frac{1}{8}, \mathbb{P}(X=1) = \frac{3}{8}, \mathbb{P}(X=2) = \frac{3}{8}, \mathbb{P}(X=3) = \frac{1}{8} \\
 \bullet \quad \mathbb{P}(Y=1) &= \frac{1}{2}, \mathbb{P}(Y=2) = \frac{1}{4}, \mathbb{P}(Y=3) = \frac{1}{8}, \mathbb{P}(Y=\text{None}) = \frac{1}{8}
 \end{aligned}$$

Give the complete "distribution of X and Y ."

- they describe how X and Y distribute the

Probabilities on their respective ranges.

S - finite or Countable.

Theorem 3.1.2 let S be a sample space with Probability $\mathbb{P}$ and let $X: S \rightarrow T$ be a function. Then X generates a Probability $\mathbb{Q}$ on T given by

$$\mathbb{Q}(B) = \mathbb{P}(X^{-1}(B))$$

The probability $\mathbb{Q}$ is called the "distribution of X ", since it describes how X distributes the probability from S onto T .

Proof:-

T , Events $\equiv$ subsets of $T \equiv \mathcal{T}$ $B \in \mathcal{T}$

$$Q(B) = P(\bar{X}^{-1}(B))$$

$$\boxed{\begin{array}{l} X: S \rightarrow T \\ P - \text{Probability} \\ \text{on } S \end{array}}$$

Verify the two Axioms of Probability

$$(1) \quad Q(T) = 1$$

Now

$$\bar{X}^{-1}(T) = \{s \in S \mid X(s) \in T\} = S$$

Proof: $Q(T) = P(\bar{X}^{-1}(T)) = P(S) = 1$

(2) $B_1, B_2, \dots, B_n, \dots$ be a sequence of disjoint events in T

$$Q\left(\bigcup_{i=1}^{\infty} B_i\right) = \sum_{i=1}^{\infty} Q(B_i)$$

Proof:- $Q\left(\bigcup_{i=1}^{\infty} B_i\right) = P\left(\bar{X}^{-1}\left(\bigcup_{i=1}^{\infty} B_i\right)\right)$

$$\textcircled{*} \quad \bar{X}^{-1}\left(\bigcup_{i=1}^{\infty} B_i\right) \stackrel{\text{Ex}}{=} \bigcup_{i=1}^{\infty} \bar{X}^{-1}(B_i), \quad \text{Set Theory exercise}$$

$$Q\left(\bigcup_{i=1}^{\infty} B_i\right) \stackrel{(*)}{=} P\left(\bigcup_{i=1}^{\infty} \bar{X}^{-1}(B_i)\right)$$

$\textcircled{**}$ $\{B_i\}_{i=1}^{\infty}$ are disjoint events in T $\Rightarrow$ $\{\bar{X}^{-1}(B_i)\}_{i=1}^{\infty}$ are also disjoint events in S
 Ex

Since P is a Probability on S

$$Q\left(\bigcup_{i=1}^{\infty} B_i\right) = P\left(\bigcup_{i=1}^{\infty} \bar{X}^{-1}(B_i)\right) \stackrel{(**)}{=} \sum_{i=1}^{\infty} P(\bar{X}^{-1}(B_i))$$

$$= \sum_{i=1}^{\infty} Q(B_i)$$

$\Rightarrow Q$ is a Probability on T . $\square$

Definition 3.1.5 : A discrete random variable is a function $X: S \rightarrow T$ where S is a sample space equipped with a Probability $\mathbb{P}$ and T is a countable (or finite) subset of real numbers.

Define: $f_X: T \rightarrow [0, 1]$

$$f_X(t) = \mathbb{P}(X=t)$$

$f_X(\cdot)$ is called the Probability mass function of X .

Then, $A \subseteq T$

$$\mathbb{P}(X \in A) = \sum_{t \in A} \mathbb{P}(X=t) = \sum_{t \in A} f_X(t)$$

Example : Say $X \sim \text{Bernoulli}(p)$

$$X: S \rightarrow \{0, 1\}$$

$\mathbb{P}$ is a Probability on S

$$\mathbb{P}(X=0) = 1-p \quad \& \quad \mathbb{P}(X=1) = p.$$

$$\text{or } f_X(0) = 1-p \quad f_X(1) = p$$

Definition 3.1.6 let $X: S \rightarrow T$ and $Y: S \rightarrow T$ be discrete random variables. we say X and Y have equal distribution provided
(same)

$$\mathbb{P}(X=t) = \mathbb{P}(Y=t) \quad \text{for all } t \in T.$$

Common distributions

Definition 3.1.7 :

(a) $X \sim \text{Uniform } \{1, \dots, n\}$ for some $n \geq 1$ if X is a discrete random variable such that

$$P(X=k) = \frac{1}{n} \quad 1 \leq k \leq n$$

we say X is a uniform random variable on the set $\{1, \dots, n\}$.

(b) $X \sim \text{Bernoulli}(p)$ for some $0 \leq p \leq 1$ if X is a discrete random variable such that

$$P(X=0) = 1-p \quad \text{and} \quad P(X=1) = p$$

we say X is a Bernoulli random variable with parameter p .

(c) $X \sim \text{Binomial}(n, p)$ $0 \leq p \leq 1$, $n \geq 1$
 X is a discrete random variable such that

$$P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$

3.2.1 Independent Random variables : $X: S \rightarrow T_1$, $Y: S \rightarrow T_2$

Definition 3.2.1 : Two random variables X and Y are independent if $\{X \in A\}$ and $\{Y \in B\}$ are independent for every event $A \subseteq T_1$ in the range of X and every event $B \subseteq T_2$ in the range of Y .

Equivalently :- • X and Y are independent-

$$\text{if } P(X \in A, Y \in B) = P(X \in A) P(Y \in B) \quad \text{for all} \\ A \subseteq T_1, B \subseteq T_2$$

• (Exercise) X and Y are independent if

$$P(X=t, Y=s) = P(X=t) P(Y=s) \quad \begin{array}{l} \forall t \in T_1 \\ \forall s \in T_2 \end{array}$$