# Stats 110 Notes

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1	Lecture 1: Probability and Counting	

# 1

#### Notes 1.1

- A Sample Space is the set of all possible outcomes of an experiment
- An **event** is a subset of the sample space.
- Naive Definition of Probability:

$$P(A) = \frac{\# \ of \ favorable \ outcomes}{\# \ possible \ outcomes}$$

- The Multiplication Rule if you have experiment with n possible outcomes, and for each outcome of the first experiment there are n outcomes for the second experiments, ..., for each  $r^{th}$  experiment there are  $n_r$  outcomes, then there are  $n_1 * n_2 * ... * n_r$ overall possible outcomes. Assumes all outcomes are equally likely within a finite sample space
- Binomial Coefficient:

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

0 if k > n

How do you get there? Number of subsets of size k of group n. Choosing k from n in order: n\*(n-1)(n-2)\*...\*(n-k+1) Now if order does not matter:  $\frac{n*(n-1)(n-2)*...*(n-k+1)}{k!}$  Now through canceling we reach the formula for the binomial coefficient.

## 1.2 Example: Full House, 5 Card Hand

There are 13 suits. We must choose one suit then 3 of the 4 possible cards. Then we must choose another suit of the remaining 12, then select 2 of the 4 possible cards. Then divide by the sample space which is all possible 5 card hands from a deck of 52 cards.

$$\frac{13\binom{4}{3} * 12\binom{4}{2}}{\binom{52}{5}}$$

## 1.3 Sampling Table

Sampling Table			
	Order Matters	Order Doesn't Matter	
Replace	$n^k$	$\binom{n+k-1}{k}$	
Don't Replace	n(n-1)(n-2)(n-k+1)	$\binom{n}{k}$	

# 2 Lecture 2: Story Proofs, Axioms of Probability

#### 2.1 Notes

• Classic and Fundamental Relationship:

$$\binom{n}{k} = \binom{n}{n-k}$$

• Exploring picking k times from set of n objects with replacement where order doesn't matter. Let's try out a few cases:

## Exploration

- Case 1:  $k = 0 \to \binom{n-1}{0} = 1$ 

– Case 2:  $k = 1 \rightarrow \binom{n}{1} = n$ 

- Case 3 (Simplest non-trivial example):  $n = 2 \rightarrow \binom{k+1}{k} = \binom{k+1}{1} = k+1$ 

- This is equivalent to the number of ways to put k indistinguishable particles into n distinguishable boxes, which can be visualized as such:

The | represent partitions for the 'boxes' and the  $\cdot$  is a particle. There are then  $k \cdot$ 's and n-1 |'s. This can be simplified again to be thought of just arranging symbols, and you are choosing to place the symbols. There are k+n-1=n+k-1 symbols, so choose where to place the separators or the particles. Choose where the  $\cdot$ 's are, and the |'s are then positioned, and vice versa.

$$\binom{n+k-1}{k} = \binom{n+k-1}{n-1}$$

• Story Proof: proof by interpretation.

$$n\binom{n-1}{k-1} = k\binom{n}{k}$$

Pick k people out of n with designated as the president, so you can first pick the president out of n, then select who can be in the club. Or you can pick the club k out of n then choose the president from the k people.

- Vandermonde Identity:

$$\binom{m+n}{k} = \sum_{j=0}^{k} \binom{m}{j} \binom{n}{k-j}$$

You have two groups with labeled people containing m and n people. You need to select k people total from both groups. The summation refers to the combinations of selecting j from m and k-j from n such that you have j+k-j=k total people from the m+n total people in two groups. Each instance is multiplied together due to the multiplication rule.

## • The Non-Naive Definition of Probability

- A probability sample consists of **S** and **P** where **S** is a sample space and **P**, a function which takes an event  $A \subseteq S$  as input and returns  $P(A) \in [0,1]$  as output such that:
  - \*  $(1) P(\emptyset) = 0, P(S) = 1$
  - \* (2)

$$P(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} P(A_n)$$

if  $A_1, A_2, ..., A_n$  are disjoint.

#### 3 Lecture 3: