Class Notes 1-30-2019

• Example: Cards

We draw a card at random from a standard deck of 52 cards. The event A that the card is an ace is independent of the event C that the card is a club. However, this breaks down as soon as the king of diamonds is missing from the deck, but not when all kings are missing.

• Theorem: Independence of *E*, *F* implies independence of *E*, *F^c*, of *E^c*, *F* and of *E^c*, *F^c*. Proof (partial):

$$P(EF^c) + P(EF) = P(E) \Rightarrow P(EF^c) = P(E) - P(E)P(F) = P(E)(1 - P(F)) = P(E)P(F^c)$$

• Assume A and B are independent. Find an expression, in terms of only P(A) and P(B), for the probability of the event C that only exactly one of A and B occur. Solution: $C = AB^c \cup A^cB$ is a disjoint union, thus

$$P(C) = P(AB^{c}) + P(A^{c}B) = P(A)P(B^{c}) + P(A^{c})P(B) = P(A)(1 - P(B)) + (1 - P(A))P(B)$$

where the second = uses the previous theorem.

• Mutually Independent: Events A_1, \ldots, A_n are called *mutually independent* if for every collection A_{i_1}, \ldots, A_{i_k} with $2 \le k \le n$ and $1 \le i_1 < i_2 < \ldots < i_k \le n$ we have

$$P(A_{i_1}A_{i_2}\cdots A_{i_k}) = P(A_{i_1})P(A_{i_2})\cdots P(A_{i_k})$$

• Mutual Independence of 3 Events *A*, *B*, *C* Need to check the truth of

$$P(AB) = P(A)P(B), \ P(AC) = P(A)P(C), \ P(BC) = P(B)P(C), \ P(ABC) = P(A)P(B)P(C)$$

- Mutual Independence Carries over to Complements: If events A_1, \ldots, A_n are *mutually independent* so are A_1^*, \ldots, A_n^* , where A_i^* is either A_i or A_i^c .
- Example of Events not Mutually Independent: Choose a point randomly on the interval $\Omega = [0, 1]$ and consider the events

$$A = \begin{bmatrix} \frac{1}{2}, 1 \end{bmatrix}, \quad B = \begin{bmatrix} \frac{1}{2}, \frac{3}{4} \end{bmatrix} \quad \text{and} \quad C = \begin{bmatrix} \frac{1}{16}, \frac{9}{16} \end{bmatrix} \quad \text{then} \quad ABC = \begin{bmatrix} \frac{1}{2}, \frac{9}{16} \end{bmatrix}$$

and $P(ABC) = \frac{1}{16} = \frac{1}{2} \frac{1}{4} \frac{1}{2} = P(A)P(B)P(C)$ but $P(AB) = \frac{1}{4} \neq \frac{1}{8} = P(A)P(B)$.

• Pairwise Independence:

Events A_1, \ldots, A_n are pairwise independent if any two of its events are independent, i.e., $P(A_iA_j) = P(A_i)P(A_j)$ for any $i \neq j$. This is a weaker form of independence then mutual independence

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• Example of Pairwise but not Mutual Independence:

Flip 3 fair coins and consider the following events. A is the event of exactly one tails in the

first two flips, B is the event of exactly one tails in the last two flips, C is the event of exactly one tails in the first and last flip.

$$A = \{(T, H, H), (T, H, T), (H, T, H), (H, T, T)\} \quad P(A) = \frac{1}{2}$$
$$B = \{(H, T, H), (T, T, H), (H, H, T), (T, H, T)\} \quad P(B) = \frac{1}{2}$$
$$C = \{(T, H, H), (T, T, H), (H, H, T), (H, T, T)\} \quad P(C) = \frac{1}{2}$$

$$\begin{split} AB &= \{ (\mathtt{T}, \mathtt{H}, \mathtt{T}), (\mathtt{H}, \mathtt{T}, \mathtt{H}) \}, \quad AC &= \{ (\mathtt{T}, \mathtt{H}, \mathtt{H}), (\mathtt{H}, \mathtt{T}, \mathtt{T}) \}, \quad BC &= \{ (\mathtt{T}, \mathtt{T}, \mathtt{H}), (\mathtt{H}, \mathtt{H}, \mathtt{T}) \}, \quad ABC = \emptyset \\ P(AB) &= \frac{1}{4} = P(A)P(B), \quad P(AC) = \frac{1}{4} = P(A)P(C), \quad P(BC) = \frac{1}{4} = P(B)P(C) \\ P(ABC) &= 0 \neq P(A)P(B)P(C) = \frac{1}{8} \end{split}$$

• A Reliability Example:

A system functions, event D, as long as one of two subsystems C_1 or C_2 functions. Subsystem C_1 functions as long both its components function, denoted as events S_1 and S_2 . C_2 functions as long as its one component functions, event S_3 . Assume that all components function or fail independently, with $P(S_i) = p_i$

$$D = C_1 \cup C_2 = (S_1 S_2) \cup S_3$$

$$P(D) = P(C_1 \cup C_2) = P(C_1) + P(C_2) - P(C_1C_2) = P(S_1S_2) + P(S_3) - P(S_1S_2S_3)$$

= $p_1p_2 + p_3 - p_1p_2p_3$

Boeing's Scientific Research Laboratory (BSRL) played a big role in developing the field of *Reliability Theory*. Z.W. Birnbaum, founding father of statistics at the UW, was an active contributor to BSRL, as were Barlow, Proschan, Saunders and Pyke.

• Independence of Random Variables:

Random variables X_1, \ldots, X_n , defined on the same probability space Ω , are called independent if

$$P(X_1 \in B_1, \dots, X_n \in B_n) = \prod_{i=1}^n P(X_i \in B_i)$$

for all reasonable subsets B_1, \ldots, B_n on the real line \mathbb{R} .

This is not very practical to check, because of the innumeracy of such subsets. For discrete random variables we have the following equivalence.