Problem 1.

- (3.6) Let $\{B_i\}_{i\in I}$ be a collection of elements in \mathscr{F} such that $\{B_i\}_{i\in I}$ is a partition of Ω . Then for any $A\in\mathscr{F}$, since $B_i\cap B_j=\emptyset$ for all $i\neq j$, we have that $(A\cap B_i)\cap (A\cap B_j)=A\cap B_i\cap B_j=A\cap\emptyset=\emptyset$ for all $i\neq j$. Furthermore, $\cup_{i\in I}(A\cap B_i)=A$ because $\Omega=\cup_{i\in I}B_i$ and $A\cap\Omega=A$. Therefore, $\{(A\cap B_i)\}_{i\in I}$ forms a partition of A. It follows that $P(A)=P(\cup_{i\in I}(A\cap B_i))=\sum_{i\in I}P(A\cap B_i)$.
- (3.8) Let $\{E_1, E_2, ..., E_n\}$ be a collection of independent events. This implies that the compliments $\{E_1^c, E_2^c, ..., E_n^c\}$ are independent as well. Since the compliments are independent and by De Morgan's laws $\bigcup_{k=1}^n E_k = (\bigcap_{k=1}^n E_k^c)^c$ we have that

$$P(\cup_{k=1}^{n} E_{k}) = P((\cap_{k=1}^{n} E_{k}^{c})^{c}) = 1 - P(\cap_{k=1}^{n} E_{k}^{c}) = 1 - \prod_{k=1}^{n} P(E_{k}^{c}) = 1 - \prod_{k=1}^{n} (1 - P(E_{k}))$$

$$P(s = \text{crime } | s \text{ tested } +) = \frac{P(s \text{ tested } + | s = \text{crime }) * P(s = \text{crime })}{P(s \text{ tested } +)} = \frac{P(s \text{ tested } + | s = \text{crime }) * P(s = \text{crime })}{P(s \text{ tested } + | s = \text{crime }) * P(s = \text{crime })} = \frac{P(s \text{ tested } + | s = \text{crime }) * P(s = \text{crime })}{P(s \text{ tested } + | s = \text{crime }) P(s = \text{crime }) * P(s \neq \text{crime })}$$
We have that

$$P(s \text{ tested } + | s \neq \text{crime }) = \frac{1}{3000000}$$

$$P(s = \text{crime }) = \frac{1}{250000000}$$

$$P(s \neq \text{crime }) = 1 - \frac{1}{250000000}$$

$$P(s \text{ tested } + | s = \text{crime }) = 1$$

Therefore

$$P(s = \text{crime } | s \text{ tested } +) = .0118$$

(3.12) WLOG assume contestant chooses door 1. Denote the event that the car is behind door i for $i \in 1, 2, 3$ as C_1, C_2, C_3 where

$$P(C_i) = \frac{1}{3}, i \in 1, 2, 3$$

Denote the door Monty opens as D_1, D_2, D_3 . Then the probability monty opens a door given that the contestant chooses door 1 and the car is behind door i is:

$$P(D_3|C_1) = 1/2$$

$$P(D_3|C_2) = 1$$

$$P(D_3|C_3) = 0$$

By Bayes' law we have that:

$$P(C_2|D_3) = \frac{P(D_3|C_2)P(C_2)}{P(D_3|C_1)P(C_1) + P(D_3|C_2)P(C_2) + P(D_3|C_3)P(C_3)} = \frac{2}{3}$$

$$P(C_1|D_3) = \frac{P(D_3|C_1)P(C_1)}{P(D_3|C_1)P(C_1) + P(D_3|C_2)P(C_2) + P(D_3|C_3)P(C_3)} = \frac{1}{3}$$

Therefore the probability of switching is higher. Similarly, if there are 10 doors and Monty opens 8, the probability of winning if the contestant switches is $\frac{9}{10}$ and the probability of winning if the contestant doesn't switch is $\frac{1}{10}$.

- (3.16) Let X be a random variable with $E[X] = \mu$ and X^2 be the random variable given by $X^2(\omega) = (X(\omega))^2$. Then $Var[X] = E[(X E[X])^2] = E[X^2 2XE[X] + E[X]^2] = E[X^2] E[2XE[X]] + E[E[X]^2] = E[X^2] 2E[X]E[X] + E[X]^2 = E[X^2] E[X^2] \mu^2$
- (3.33) Let B = B(n, p) be a binomial random variable for n trials with parameter p. Then since $\mu = pn$ and $\sigma^2 = n(p)(1-p)$, for any $\epsilon > 0$ we have by Chebyshev's inequality that

$$P(|\frac{B}{n} - p| \ge \epsilon) = P(|B - pn| \ge n\epsilon) = P(|B - \mu| \ge n\epsilon) \le \frac{p(1 - p)n}{n^2 \epsilon^2} = \frac{p(1 - p)}{n\epsilon^2}$$

(3.36) The probability of each student enrolling is Bernoulli(.801) and the students are i.i.d., we have that if 6242 students are offered admission the expected number of students is 5000. Denote S_{6242} as the number of students enrolled given 6242 admissions, so by the Central Limit Theorem:

$$P(S_{6242} > 5500) = P(\frac{S_{6242} - 5000}{\sqrt{6242(.801)(.199)}} > \frac{5500 - 5000}{\sqrt{6242(.801)(.199)}}) =$$

$$= P(\frac{S_{6242} - 5000}{\sqrt{6242(.801)(.199)}} > 15.85) \approx 1 - \Phi(15.85) = 0$$

Problem 2.

- (a) Events A, B, C must be pairwise independent, but not mutually independent. Consider four equally likely points $\Omega = \{1, 2, 3, 4\}$ and let $A = \{1, 2\}$, $B = \{2, 3\}$, $C = \{3, 1\}$. Then P(A) = P(B) = P(C) = 1/4 and $P(A \cap C) = P(\{1\}) = 1/4 = P(A)P(C)$. The other pairwise probabilities similarly follow because the sets have similar structures. However, $P(A \cap B \cap C) = P(\{\emptyset\}) = 0 \neq 1/8 = P(A)P(B)P(C)$.
- (b) Now let events A, B, C must be mutually independent, but not pairwise independent in B and C. Consider eight equally likely points $\Omega = \{1, 2, ..., 8\}$ and let $A = \{1, 2, 3, 4\}$, $B = \{3, 4, 5, 6\}$, $C = \{1, 3, 6, 7\}$. Then Then P(A) = P(B) = P(C) = 1/2 and $P(A \cap B) \cap C = P(\{3\}) = 1/8 = P(A)P(B)P(C)$. Additionally, $P(A \cap C) = P(\{1, 3\}) = 1/4 = P(A)P(C)$ and $P(A \cap B) = P(\{3, 4\}) = 1/4 = P(A)P(B)$ but $P(B \cap C) = P(\{\emptyset\}) = 0 \neq 1/4 = P(B)P(C)$.

Problem 3. In Benford's law, $P(d) = \log_{10}(1+1/d)$ for d = 1, 2, ..., 9. Then

$$\sum_{d=1}^{9} P(d) = \sum_{d=1}^{9} \log_{10}(1+1/d) = \log_{10}(\Pi_{d=1}^{9}(1+1/d)) = \log_{10}(2\frac{3}{2}\frac{4}{3}\frac{5}{4}...\frac{10}{9}) = \log_{10}(10) = 1$$

Problem 4.

(a) $E[X] = \sum_{n=1}^{\infty} 2^n P(\text{first tail on flip } n) = \sum_{n=1}^{\infty} 2^n (\frac{1}{2})^n = \sum_{n=1}^{\infty} (\frac{2}{2})^n = \sum_{n=1}^{\infty} 1 = +\infty$

(b) $E[\ln X] = \sum_{n=1}^{\infty} \ln(2^n) P(\text{first tail on flip } n) = \sum_{n=1}^{\infty} n \ln(2) (\frac{1}{2})^n = \ln(2) \sum_{n=1}^{\infty} n (\frac{1}{2})^n = \ln($

$$= \ln(2) * \frac{\frac{1}{2}}{(1 - \frac{1}{2})^2} = 2\ln(2)$$

Problem 5. Let r be the risk free interest rate in both US and Switzerland. Then if the investors invest n in the assets of their respective countries, their returns will be (1+r)n since the interest rate is risk free. If they invest in a foreign country then

E[foreign investment] = 0.5[1.25(1+r)n + (1+r)n/1.25] = (1+r)n(1.025) > (1+r)nso the investors should invest in the asset of the foreign country.

Problem 6. Let $\Omega = [0, 1]$.

(a) Let X be the map from $\Omega = [0,1]$ to the Pareto distribution on $\mathbb R$ where parameter $x_m = 1$. Then

$$f_X(x) = \begin{cases} \frac{\alpha x_m^{\alpha}}{x^{\alpha+1}} & \text{for } x \le 0\\ 0 & \text{for } x < x_m \end{cases}$$

When $\alpha \in (1,2]$ the expectation is finite $E[X] = \frac{\alpha x_m}{\alpha - 1}$ and the variance is infinite so $E[X^2] = \infty$.

- (b) Let X be a uniform distribution on [7.5, 7.51] so E[X] > 7.5 and let Y be the polynomial distribution on [0, 10] with pdf $3/1000x^2$ so E[Y] = 7.5. Since P(Y > 7.51) > 0.57 we have that P(Y > X) > 1/2.
- (c) Let X be a uniform distribution on [-3,3]; Y be a uniform distribution on [-2,2] and let Z be a uniform distribution on [-1,1]. Then E[X] = E[Y] = E[Z] = 0 and P(X > Y)P(Y > Z)P(X > Z) > 0 since each probability is greater than 0.

Problem 7. Let random variables X, Z be independent and $X \sim N(0, 1)$ and $P(Z = 1) = P(Z = -1) = \frac{1}{2}$. Let Y = XZ.

- (a) True, $f_Y(x) = f_{XZ}(x) = \frac{1}{2}(f_X(x) + f_X(-x)) = \frac{1}{2}\left[\frac{1}{\sqrt{2\pi}}e^{-x^2/2} + \frac{1}{\sqrt{2\pi}}e^{-(-x)^2/2}\right] = \frac{1}{2}2\frac{1}{\sqrt{2\pi}}e^{-x^2/2} = \frac{1}{\sqrt{2\pi}}e^{-x^2/2}$ so we have that $Y \sim N(0, 1)$.
- (b) True, |Y| = |XZ| = |X| so P(|X| = |Y|) = 1.
- (c) True, we easily show this by counter example: $P(Y > 0|X = 0) = 0 \neq P(Y > 0)$.
- (d) True, $Cov(X,Y) = E[XY] E[X]E[Y] = E[XXZ] E[X]E[XZ] = E[X^2]E[Z] 0*0 = E[X^2]*0 = 0.$
- (e) False, we just showed using part(d) that it is possible for Cov(X, Y) = 0 even when X and Y are not independent.

Problem 8. Let the random variables X_i , i = 1, 2, ..., n be i.i.d and $X_i \sim U[0, 1]$. Let m and M be random variables such that $m = \min\{X_1, X_2, ..., X_n\}$ and $M = \max\{X_1, X_2, ..., X_n\}$. Then for $x \in [0, 1]$ we have

$$P(M \le x) = P(x \ge \max\{X_1, X_2, ..., X_n\}) = P(x \ge X_1, x \ge X_2, ..., x \ge X_n) = \Pi_{i=1}^n P(x \ge X_i) = (x)^n$$

$$P(m \le x) = 1 - P(m \ge x) = 1 - \Pi_{i=1}^n P(X_i \ge x) = 1 - \Pi_{i=1}^n (1 - P(X_i \le x)) = 1 - (1 - x)^n$$

$$F(m) = \begin{cases} 1 & \text{for } x \le 0 \\ 1 - (1 - x)^n & \text{for } x \in (0, 1) \\ 0 & \text{for } x \ge 1 \end{cases} \qquad F(M) = \begin{cases} 0 & \text{for } x \le 0 \\ (x)^n & \text{for } x \in (0, 1) \\ 1 & \text{for } x \ge 1 \end{cases}$$

$$f_m(x) = n(1-x)^{n-1}$$

 $f_M(x) = n(x)^{n-1}$

$$E[m] = \int_0^1 x n(1-x)^{n-1} dx = \frac{1}{n+1}$$
$$E[M] = \int_0^1 x n(x)^{n-1} dx = \frac{n}{n+1}$$

Problem 9.

(a) We know that $X_i \sim \text{Bernoulli}(.5)$ where the X_i are i.i.d with $\mu = 0.5$ and variance p(1-p) = 0.25 so

$$P(490 < S_1000 < 510) = P(\frac{-10}{\sqrt{1000 * 0.5 * 0.5}} < \frac{S_1000 - 500}{\sqrt{1000 * 0.5 * 0.5}}) < \frac{10}{\sqrt{1000 * 0.5 * 0.5}})$$

$$\approx 2\Phi(.632) - 1 = 0.47$$

(b) By the Weak Law of Large Numbers,

$$P(\left|\frac{S_{1000}}{1000} - \frac{1}{2}\right| \ge 0.005) \le \frac{1}{4n * 0.005^2} \le 0.01$$

Therefore we have that $n \ge 1000000$.

Problem 10. Let E[X] < 0 and $\theta \neq 0$ such that $E[e^{\theta X}] = 1$. Because $g(x) = e^x$ is a convex differentiable function, by Jensen's inequality, $E[e^{\theta X}] \geq e^{E[\theta X]} = e^{\theta E[X]}$. Taking ln of both sides, we get that $0 = \ln(1) \geq \theta E[X]$ and E[X] < 0 and $\theta \neq 0$ implies that $0 > \theta E[X]$ so $\theta > 0$.