## Problem 1

Let $A$ and $B$ be two events, and $P(B) \neq 0$. Let $X$ and $Y$ be two discrete random variables, then
i) Bayes' rule for events is

$$
P(A \mid B)=\frac{P(B \mid A) P(A)}{P(B)}
$$

for discrete r.v. is

$$
p_{X \mid Y}(x \mid y)=\frac{p_{Y \mid X}(y \mid x) p_{X}(x)}{p_{Y}(y)}
$$

for $p_{Y}(y) \neq 0$.
ii) Total probability law for events is

$$
P(A)=\sum_{n} P\left(A \mid B_{n}\right) P\left(B_{n}\right)
$$

where $\left(B_{n}\right)$ is a partition of the whole sample space; for discrete r.v. is

$$
p_{X}(x)=\sum_{y} p_{X \mid Y}(x \mid y) p_{Y}(y)
$$

## Problem 2

a) none.
b) $A$ and $B$; $A$ and $C$; $A$ and $D$.
c) $C \subset B \subset D$.
d)


## Problem 3

a) False. A counter example is as follows. Let $X$ and $Y$ be independent r.v.s, and $Z=X+Y$. Then $X$ and $Y$ are independent by our setting, but they are not independent conditional on $Z$. In fact, conditional on $Z, X$ and $Y$ are determined by each other.
(b) False. A counter example is as follows. Let $Z, N_{1}$ and $N_{2}$ be independent r.v.s. Define $X=Z+N_{1}$, and $Y=Z+N_{2}$. Then $X$ and $Y$ are obviously not independent by our setting. However they are independent conditional on $Z$, since then $X$ and $Y$ are determined by $N_{1}$ and $N_{2}$ respectively and $N_{1}$ and $N_{2}$ are independent.

## Problem 4

a)

$$
\begin{aligned}
\operatorname{Var}(X)=E\left[(X-E(X))^{2}\right] & =E\left[X^{2}-2 X E(X)+E(X)^{2}\right] \\
& \left.=E\left(X^{2}\right)-2 E(X) E(X)\right]+E(X)^{2} \\
& =E\left(X^{2}\right)-E(X)^{2}
\end{aligned}
$$

b) Let $X$ be a geometric r.v. with parameter $p, Y$ be indicator r.v. of the result of the first trial, i.e. $Y=1$ represents the first trial successes and $Y=0$ represents the failure. We use "divide and conquer" method to compute $E(X)$ :

$$
E(X)=E(X \mid Y=1) P(Y=1)+E(X \mid Y=0) P(Y=0)
$$

Conditional on $Y=1$, the experiment successes in the first trial, thus the conditional expectation is 1 ; conditional on $Y=0$, the first trial fails, but the later trials are independent of the first one. Thus the experiment 'restarts' again after the first trial, thus the conditional expectation is $1+E(X)$. So we have

$$
\begin{aligned}
E(X) & =1 \cdot p+(1+E(X))(1-p) \\
\Rightarrow \quad E(X) & =\frac{1}{p}
\end{aligned}
$$

c) From part a) we know we need to compute $E\left(X^{2}\right)$ to get $\operatorname{Var}(X)$. To compute $E\left(X^{2}\right)$, we use the same method in part b):

$$
\begin{aligned}
E\left(X^{2}\right) & =1 \cdot p+E\left((1+X)^{2}\right)(1-p) \\
\Rightarrow \quad E\left(X^{2}\right) & =p+E\left(X^{2}+2 X+1\right)(1-p) \\
\Rightarrow \quad E\left(X^{2}\right) & =\frac{2-p}{p^{2}}
\end{aligned}
$$

Finally, we get

$$
\operatorname{Var}(X)=E\left(X^{2}\right)-E(X)^{2}=\frac{1-p}{p^{2}}
$$

## Problem 5

## Part I:

a) Let $W$ represent a packet is transmitted successfully, and $L$ represent a packet is erased. $\Omega=$ $\{W W, W L, L W, L L\}$
b)

$$
P(W W)=(1-p)^{2}, \quad P(W L)=P(L W)=p(1-p), \quad P(L L)=p^{2}
$$

c) Yes, it is enough to specify the probability law, as follows:

$$
P(W W)=1-p-p(1-q), \quad P(W L)=P(L W)=p(1-q), \quad P(L L)=p q
$$

d) Let $S$ be the event that packet $A$ is successfully communicated. Then for the model in b), $S=\{L L\}^{c}$ and thus $P(S)=1-p^{2}$. For the model in c), $P(S)=1-p q$.
e) From description, the pair far apart in time can be considered to be transmitted independently. By part d), we know that if $q>p$, i.e. the loss of successive packets are highly positively dependent, then scheme (ii) is better, resulting in higher successfully communication probability.
f) A natural decoding scheme for packet $A$ (and so for packet $B$ ) is as follows. If packet $A$ is successfully transmitted, then $A$ is already there and no need for further decoding. If $A$ is erased but $B$ and $C$ are successfully transmitted, then we decode $A=B \bigoplus C$. If none of above events happens, we declare $A$ can not be decoded, and the communications of $A$ fails.
Thus the probability of successfully decoding packet $A$ is just the summation of the probability of two events that result in successful decoding:

$$
P(S)=1-p+p(1-p)^{2}=1-p^{2}-\left(p^{2}-p^{3}\right)
$$

Compared to the duplication scheme, we can see the duplication scheme has a higher successful transmission probability. More generally, this result is of two folds. First, for small $p$, the successful transmission probability is much better than no coding but not as good as the duplication scheme; but the data rate of this scheme is 1.5 times of that of duplication scheme.

## Part II:

a) In this part, we assume the feedback channel is perfect, i.e. no feedback packet gets lost. Define a r.v. $X$ as the number of packet transmissions, then $X$ follows the geometric distribution. Then the PMF is

$$
p_{X}(x)=p^{x-1}(1-p), \quad x \in \mathcal{N}
$$

and the expectation is

$$
E(X)=\frac{1}{1-p}
$$

b) By setting,

$$
\begin{aligned}
P(X=x)= & P(1 \text { st packet lost, } 2 \text { nd packet lost, } \ldots, \text { xth packet is received }) \\
= & P(1 \text { st packet lost }) P(2 \text { nd packet lost } \mid 1 \text { st packet lost }) \cdots \\
& P(\text { xth packet is received } \mid \text { st }, 2 \text { nd, } \ldots, \text { packets lost }) \\
= & \begin{cases}1-p, & x=1 \\
p q^{x-2}(1-q), & x \geq 2\end{cases}
\end{aligned}
$$

Thus the expectation is

$$
\begin{aligned}
E(X) & =\sum_{x=1}^{\infty} x p_{X}(x) \\
& =1-p+\left(\frac{p}{q} \sum_{x=1}^{\infty} x q^{x-1}(1-q)\right)-1 \cdot \frac{p}{q}(1-q) \\
& =1-p+\frac{p}{q} \cdot \frac{1}{1-q}-\frac{p}{q}(1-q) \\
& =1+\frac{p}{1-q}
\end{aligned}
$$

Compared to the expectation in a), we conclude if we want the expected time to be longer, the following must be true:

$$
1+\frac{p}{1-q}>\frac{1}{1-p} \Leftrightarrow q>p
$$

c) Under this situation, the number of packet transmissions $X$ still follows geometric distribution, but with parameter $(1-p)(1-r)$ since a transmission is successful if and only if both forward transmission and feedback are successful. Let $p_{1}=(1-p)(1-r)$, then we have PMF as

$$
p_{X}(x)=\left(1-p_{1}\right)^{x-1} p_{1}, \quad x \in \mathcal{N}
$$

and expectation as

$$
E(X)=\frac{1}{p_{1}}=\frac{1}{(1-p)(1-r)}
$$

