9.

Intro

In this question we have to exercise our capacity of getting some information about conditional independence and see if we can infer something more based on it. I would like to give a special shout out to github user xulongwu4 and the repository MLaPP from astahlman for pointing out a mistake in item b. The item is now corrected.

Solution

a)

We have two information about conditional independence. The first one is:

$$X \perp W|Z, Y$$

Based on this, we can state that:

$$p(x, w|z, y) = p(x|z, y)p(w|z, y)$$
(1)

The second one is:

$$X \perp Y|Z$$

Based on this, we can state that:

$$p(x,y|z) = p(x|z)p(y|z)$$
(2)

Now, we have to see if $X \perp Y, W|Z$ is true.

$$p(x, y, w|z) = p(x|z)p(y|x, z)p(w|x, y, z) = p(x|z)p(y|z)p(w|y, z) = p(x|z)p(y, w|z)$$
(3)

In (3), the first passage is the chain rule of probability. The second make use of the conditional independences given to us (Y do not depend on X given Z and W do not depend on X given Z and Y). The third passage is putting the joint distribution back together. Therefore, the proposition is **true**.

b)

Let's prove that this proposition is false with a counter example. Start by defining X, Y and Z to be i.i.d. random variables with the following distribution:

$$\begin{cases}
P(X=1) = 0.5 \\
P(X=-1) = 0.5
\end{cases}$$
(4)

Also, define W = XYZ. First, we shall prove that the two conditional independences $X \perp Y|Z$ and $X \perp Y|W$ are true. Let's start with $X \perp Y|Z$:

$$p(x, y|z) = p(x|z)p(y|z, x) = p(x|z)p(y|z)$$
 (5)

where in the last passage we used the fact that X and Y are independent by construction. As for the second conditional independence, start by realizing that the original sample space of (X, Y, Z) is (-1, -1, -1), (-1, -1, 1), (-1, 1, -1), (-1, 1, 1), (1, -1, 1), (1, 1, 1) and that $P(X, Y, Z) = \frac{1}{8}$ for any of those events. Now, when we observe W = 1, our sample space is reduced by half, because only combinations with an even number of -1 will produce this W. Therefore:

$$\begin{cases} P(X,Y,Z|W=1) = 0.25 \text{ for an even number of } -1\\ P(X,Y,Z|W=1) = 0 \text{ for an odd number of } -1 \end{cases} \tag{6}$$

Notice that the same reasoning can be applied to P(X, Y, Z|W = -1), but this time, the new sample space will be the complement of the previous sample space (combinations with an odd number of -1). Based on these results we can perform the following marginalization:

$$P(X = 1|W = 1) = P(X = 1, Y = 1, Z = 1|W = 1) + P(X = 1, Y = -1, Z = -1|W = 1) + P(X = 1, Y = -1, Z = 1|W = 1) + P(X = 1, Y = 1, Z = -1|W = 1) = \frac{1}{4} + \frac{1}{4} + 0 + 0 = \frac{1}{2}$$

$$(7)$$

We can do the same for (X, W) = (1, -1), (-1, 1), (-1, -1) and for (Y, W) also. Thus, $P(X|W) = P(Y|W) = \frac{1}{2}$. Now, let's perform the same marginalization on the joint distribution P(X = 1, Y = 1|W = 1).

$$P(X = 1, Y = 1|W = 1) = P(X = 1, Y = 1, Z = 1|W = 1) + P(X = 1, Y = 1, Z = -1|W = 1) = \frac{1}{4} + 0 = \frac{1}{4}$$
(8)

We can do the same thing for the other combinations of (X,Y,W). Therefore, $P(X,Y|W)=\frac{1}{4}$. Thus, we second proposition $P(X,Y|W)=\frac{1}{4}=\frac{1}{2}\frac{1}{2}=P(X|W)P(Y|W)$ is also true.

As for the final step

$$P(X,Y|Z,W) = P(X|Z,W)P(Y|X,Z,W) = \frac{1}{2}I(Y = \frac{W}{XZ})$$

$$P(X|Z,W)P(Y|Z,W) = \frac{1}{2}\frac{1}{2} = \frac{1}{4}$$
(9)

So, the proposition $X \perp Y|Z,W$ does not hold true.

Conclusion

There is not a major conclusion to be mentioned. As it was said it the begin, I think the main purpose of this exercise is just to exercise the brain in inference involving conditional independence. A special notice to item b shoud be made. The counter example was based on i.i.d. random variables and a "linking" random variable W, which made the random variables dependent in the following circumstance $P(Y|X,Z,W) = I(Y = \frac{W}{XZ})$. Again, I would like to thank xulongwu4 and MLaPP repository for this correction.