Let $X \sim N(\mu, \sigma^2)$

Then $X = \sigma(z) + \mu$

From theorem 4.8 we know:

$$\phi_{aX+b}(t) = e^{ibt}\phi_X(at)$$

The characteristic function standard normal distribution $Z \sim N(0,1)$ is:

$$\phi_Z(t) = e^{-t^2/2}$$

Plugging in $a = \sigma$ and $b = \mu$ into theorem 4.8 we have:

$$\begin{split} \phi_X(t) &= \phi_{\sigma Z + \mu}(t) = e^{i\mu t} \phi_Z(\sigma t) \\ \phi_X(t) &= e^{i\mu t} e^{-(\sigma t)^2/2} = e^{i\mu t - \sigma^2 t^2/2} \end{split}$$

Problem 2

$$p_X(k) = \begin{cases} e^{-\lambda} \lambda^k / k! & k = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

We can write:

$$\frac{P(X=k)}{P(X=k-1)} = \frac{\lambda}{k}$$

 p_X is increasing monotonically for $\frac{\lambda}{k} > 1$

 p_X is decreasing monotonically for $\frac{\lambda}{k} < 1$

Let i = the largest integer value not exceding λ

Thus p_x is increasing for k = 0, 1, ..., i

and p_x is decreasing for k = i + 1, i + 2, ...

Since $p_X(i) \geq p_X(k) \ \forall \ k$ in the support of X

 p_X has a maximum at k=i

Since the Poisson distribution is unimodal, i is also the mode

Problem 3

A car must pass at t=4 seconds so that the pedestrian does not start crossing before 4 seconds and no cars can pass at t=5,6,7 seconds in order for the pedestrian to only have to wait 4 seconds before crossing.

Thus we have: $p(1-p)^3$

Looking at the first 3 seconds, at least one car must pass during the period so the pedestrian does not start crossing immediately.

The probability of this is: $1 - (1 - p)^3$

Multiplying the two probabilities together we have:

$$(1 - (1-p)^3)p(1-p)^3$$

The probability of a success is .8 if the new and old drugs are equally effective. Assuming all trials are independent we can use the binomial distribution

$$P(X \ge 85) = \sum_{85}^{100} {100 \choose k} .8^k .2^{100-k} = .1285055$$

Since the probability of observing 85 or more success is almost 13%, this is not that unlikely so we cannot conclude the new drug is more effective.

(a) There are $\binom{N}{4}$ ways to select 4 packets of cocaine over $\binom{N+M}{4}$ total ways to select 4 packets. The probability of selecting 4 packets of cocaine is $\frac{\binom{N}{4}}{\binom{N+M}{4}}$

There are $\binom{M}{2}$ ways to select 2 non cocaine packets after selecting 4 cocaine packets over $\binom{N+M-4}{2}$ total ways to select 2 packets after having already selected 4 packets.

The probability of selecting 2 non cocaine packets is $\frac{\binom{M}{2}}{\binom{N+M-4}{2}}$

Thus the probability the defendant is innocent equals the above probabilities multiplied together.

$$P(\text{Defendent Innocent}) = \frac{\binom{N}{4}\binom{M}{2}}{\binom{N+M}{4}\binom{N+M-4}{2}}$$

(b) Since M = 496 - N we have:

$$\frac{\binom{N}{4}\binom{496-N}{2}}{\binom{496}{4}\binom{492}{2}} \quad N = 4, \dots, 496$$

$$\frac{P_{N+1}}{P_N} = \frac{\binom{N+1}{4}\binom{496-N-1}{2}}{\binom{N}{4}\binom{496-N}{2}}$$

$$= \frac{(N+1)}{(N-3)}\frac{(494-N)}{(496-N)}$$

Plugging in N = 331 we have

$$\frac{(331+1)(494-331)}{(331-3)(496-331)}\approx 1.00547 > 1$$

Plugging in N = 332 we have :

$$\frac{(332+1)(494-332)}{(332-3)(496-332)}\approx .9998<1$$

Thus N = 331 is a maximum

Plugging in N = 331 into the original equation we get:

$$\frac{\binom{331}{4}\binom{496-331}{2}}{\binom{496}{4}\binom{492}{2}}=.02208168\approx.022$$

The maximum innocence probability is .022 and this is attained at N=331

Homework 8 Problems 6-10 Ty Darnell

Problem 9

WTS: Limiting $r \to \infty$, $p \to 1$, $r(1-p) \to \lambda$

In the Negative Binomial mgf: $M(t) = \left(\frac{p}{1 - (1 - p)e^t}\right)^r$

We have the Poisson mgf: $M(t) = e^{\lambda(e^t - 1)}$

Let a = r(1 - p) then we have:

$$\begin{split} M(t) &= \left(\frac{1 - a/r}{1 - (a/r)e^t}\right)^r \\ &= \left(\frac{1 - (a/r)e^t + (a/r)e^t - (a/r)}{1 - (a/r)e^t}\right)^r \\ &= \left(\frac{(1 - (a/r)e^t) + (a/r)(e^t - 1)}{1 - (a/r)e^t}\right)^r \\ &= \left(\frac{r(1 - (a/r)e^t) + a(e^t - 1)}{r(1 - (a/r)e^t)}\right)^r \\ &= \left(1 + \frac{a(e^t - 1)}{r(1 - (a/r)e^t)}\right)^r \\ &= \left(1 + \frac{a(e^t - 1)}{r(1 - (a/r)e^t)}\right)^r \end{split}$$

Let
$$a_n = \frac{a(e^t - 1)}{1 - (a/r)e^t}$$

applying the limits:

$$a_n$$
 goes to
$$\frac{\lambda(e^t - 1)}{1 - (1 - 1)e^t}$$
$$= \lambda(e^t - 1)$$

And since we know from Lemma(2.3.14) that:

$$\lim_{n\to\infty} (1+\frac{a_n}{n})^n = e^a$$

where a_n is a sequence where $\lim_{n\to\infty} a_n = a$

$$\left(1 + \frac{a_n}{r}\right)^r$$
 goes to $e^{\lambda(e^t - 1)}$

Which is the Poisson mgf

Thus the Negative Binomial mgf converges to the Poisson mgf

(a)

WTS:
$$\Gamma(a+1) = a\Gamma(a)$$

$$\Gamma(a+1) = \int_0^\infty t^a e^{-t} dt$$

$$= \Big|_0^\infty - t^a e^{-t} + a \int_0^\infty t^{a-1} e^{-t} dt$$
since $\Gamma(a) = \int_0^\infty t^{a-1} e^{-t} dt$ we have:

$$= 0 + a\Gamma(a) = a\Gamma(a)$$

(b)

$$\Gamma(1/2) = \int_0^\infty t^{1/2-1} e^{-t} dt = \int_0^\infty t^{-1/2} e^{-t}$$

doing a change of variables where $u=\sqrt{2t},\ t=u^2/2,\ du=t^{-1/2}/\sqrt{2}$ we have:

$$\int_0^\infty \sqrt{2}e^{-u^2/2}\ du$$

from (3.3.14) we know:

$$\int_0^\infty e^{-u^2/2} \ du = \sqrt{\pi/2} \text{ so we have:}$$

$$\sqrt{\pi/2}\sqrt{2} = \sqrt{\pi}$$

Let $X \sim binom(1000,1/6)$ Using the normal approximation to the binomial p=1/6 n=1000 $\mu=np=1000*1/6=500/3$ $\sigma=\sqrt{(500/3)(5/6)}=\sqrt{1250}/3$ then we have $Y \sim normal(500/3,\sqrt{1250}/3)$ $P(150 \leq Y \leq 200) = P(Y \leq 200) - P(Y \leq 150)$ $= P(Z \leq (200 + .5 - 500/3)/\sqrt{1250}/3) - P(Z \leq (150 - .5 - 500/3)/\sqrt{1250}/3)$ $= \Phi(2.87) - \Phi(-1.46) = \Phi(2.87) - 1 + \Phi(-1.46) = .9258$

Problem 14

WTS: If
$$X \sim exp(\lambda)$$
, $cX \sim exp(\lambda/c)$ $c > 0$

$$f_X(x) = \lambda e^{-\lambda x} \ x \ge 0$$

$$P(X \le x) = F(x) = 1 - e^{-\lambda x} \ x \ge 0$$
Let $Y = cX$
Then $P(Y \le y) = P(cX \le y) = P(X \le y/c) = F_X(y/c) = 1 - e^{-\lambda y/c}$

$$f_Y(y) = \frac{d}{dy} 1 - e^{-\lambda y/c} = \frac{\lambda}{c} e^{-\lambda y/c} \ y \ge 0$$
Thus $cX \sim exp(\lambda/c)$

Problem 15

$$F(s) = 1 - \exp(-(\frac{s - v}{\alpha})^{\beta}) \quad s \ge v = 0$$

$$1 - F(s) = \exp(-(\frac{s - 0}{\alpha})^{\beta})$$

$$\log(1 - F(s)) = \log(\exp(-(\frac{s}{\alpha})^{\beta})) = (-\frac{s}{\alpha})^{\beta}$$

$$\log(1 - F(s))^{-1} = -(\frac{s}{\alpha})^{-\beta}$$

$$\log(\log(1 - F(s))^{-1}) = \log(-(\frac{s}{\alpha})^{-\beta})$$

$$= -\beta \log(-(\frac{s}{\alpha}))$$

$$= \beta \log(s) - \beta \log(\alpha)$$

Plotting this against $\log(s)$ we have a linear function

$$P(s \le \alpha) = 1 - \exp(-(\frac{\alpha}{\alpha})^{\beta})$$

= 1 - \exp(-(1)^{\beta}) = 1 - \exp(-1)
= .6321206