5010 solutions, Assignment 13. Chapter 7: 3, 13, 15, 24, 25, 33, 35.

3. First result is obvious. $\operatorname{Var}(X) = E[X^2] - (E[X])^2 = B(a+2,b)/B(a,b) - [B(a+1,b)/B(a,b)]^2$. It is possible to express B(a,b) in terms of the gamma function and further simplify these formulas, but the problem does not require that.

13. $E[e^{tX}] = \int_0^\infty e^{tX} \lambda e^{-\lambda x} dx = \int_0^\infty \lambda e^{-(\lambda-t)x} dx = \lambda/(\lambda-t)$, and this is valid for $t < \lambda$.

15. Let $\mu = E[X]$. Then $E[(X-c)^2] = E[(X-\mu+\mu-c)^2] = E[(X-\mu)^2] + (\mu-c)^2$, showing that the function is minimized at $c = \mu$.

24. This is important in statistics. The density of $Y = X^2$ is the derivative of $P(X^2 \leq y) = 2\Phi(\sqrt{y}) - 1$, which is $f_Y(y) = (2\pi y)^{-1/2} e^{-y/2}$ for y > 0. (See last week's quiz problem.) The mgf is that of the gamma density with parameters $\lambda = 1/2$ and $\alpha = 1/2$, which we derived in class. The result is $(\lambda/(\lambda - t))^{\alpha} = (1 - 2t)^{-1/2}$.

25. This is the density $f_X(x) = (\lambda/2)e^{-\lambda|x|}$, x real. Let Y be exponential (λ) . Then the mgf of X is

$$E[e^{tX}] = \int_{-\infty}^{\infty} e^{tx} (\lambda/2) e^{-\lambda|x|} dx$$

$$= \frac{1}{2} \int_{0}^{\infty} e^{tx} \lambda e^{-\lambda x} dx + \frac{1}{2} \int_{-\infty}^{0} e^{tx} \lambda e^{\lambda x} dx$$

$$= \frac{1}{2} \int_{0}^{\infty} e^{tx} \lambda e^{-\lambda x} dx + \frac{1}{2} \int_{0}^{\infty} e^{-tx} \lambda e^{-\lambda x} dx$$

$$= (1/2)[M_Y(t) + M_Y(-t)]$$

$$= \frac{1}{2} \left[\frac{\lambda}{\lambda - t} + \frac{\lambda}{\lambda + t} \right]$$

$$= \frac{\lambda^2}{\lambda^2 - t^2}.$$

The argument requires $|t| < \lambda$.

33. By Chebyshev, $P(|X-\mu|>a\sigma)\leq \sigma^2/(a^2\sigma^2)=1/a^2.$ Take complements.

Y has mean $\alpha + \gamma$ and variance $\beta^2 \operatorname{Var}(Z) + \gamma^2 \operatorname{Var}(Z^2) + 2\alpha\beta \operatorname{Cov}(Z, Z^2)$. The variance of Z^2 is 2 and the covariance term is 0, so the variance is $\beta^2 + 2\gamma^2$. By Chebyshev, $P(|X - \alpha - \gamma| > \alpha/2) \leq 4\alpha^{-2}(\beta^2 + 2\gamma^2)$. Again, take complements.

35. (a)
$$E[X] = E[\int_0^X 1 \, dx] = E[\int_0^\infty 1_{\{X > x\}} \, dx] = \int_0^\infty E[1_{\{X > x\}}] \, dx = \int_0^\infty P(X > x) \, dx.$$

(b) $E[X^r] = E[\int_0^X rx^{r-1} \, dx] = E[\int_0^\infty rx^{r-1} 1_{\{X > x\}} \, dx] = \int_0^\infty rx^{r-1} E[1_{\{X > x\}}] \, dx = \int_0^\infty rx^{r-1} P(X > x) \, dx.$

$$\begin{array}{l} (\mathbf{c}) \ E[e^{\theta X}] = 1 + E[\int_0^X \theta e^{\theta x} \, dx] = 1 + E[\int_0^\infty \theta e^{\theta x} \mathbf{1}_{\{X > x\}} \, dx] = 1 + \int_0^\infty \theta e^{\theta x} E[\mathbf{1}_{\{X > x\}}] \, dx = 1 + \theta \int_0^\infty e^{\theta x} P(X > x) \, dx. \\ (\mathbf{d}) \ \sum_{k=0}^\infty s^k P(X > k) = E[\sum_0^\infty s^k \mathbf{1}_{\{X > k\}}] = E[\sum_0^{X-1} s^k] = E[(1 - s^X)/(1 - s)] = (1 - G_X(s))/(1 - s). \end{array}$$