3.30
$$M'_{X}(t) = \alpha k (1 - \alpha t)^{-k-1} \longrightarrow E(X) = M'_{X}(0) = \alpha k$$
 $M''_{X}(t) = \alpha^{2} (k+1)k (1 - \alpha t)^{-k-2} \longrightarrow E(X^{2}) = M''_{X}(0) = \alpha^{2} k (k+1)$

So, $Var(X) = \alpha^{2} k (k+1) - \alpha^{2} k^{2} = \alpha^{2} k$.

3.41 (a)
$$E(x) = \frac{5}{7}$$
, $Var(x) = \frac{(5)(7)}{((5+7)^2(1+5+7))} = \frac{35}{1872} = 0.01870$.

- (b) gbeta (0.5, 5, 2) = 0.73555
- (c) Pr [X = E(X)] = pbeta (5/7, 5, 7) = 0.4516
- (d) Pr(0.2 < X < 0.4) = pbeta(0.4, 5,2) pbeta(0.2, 5,2) = 0.03936.
- (e) $P_r \left[E(\chi) \sqrt{v_{\alpha r}(\chi)} \le \chi \le E(\chi) + \sqrt{V_{\alpha r}(\chi)} \right] = 0.08336$
- 3.45 Taking into account that points-scored is always an integer, the normal-quantile plot is quite linear in its appearance, making a normal model appear appropriate. At the top end, there may be the suggestion that he should have had a higher personal-season-best.

(b)
$$P_{r}(X < Y) = \int_{0}^{1} \int_{0}^{y} 12 x^{2} y^{3} dx dy = 4 \int_{0}^{1} y^{3} [x^{3}]^{y} dy = 4 \int_{0}^{1} y^{6} dy = \frac{4}{7}.$$

(c)
$$f_{\chi}(x) = 12x^2 \int_0^1 y^3 dy = 3x^2$$
, and $f_{\gamma}(y) = 12y^3 \int_0^1 x^2 dx = 4y^3$.
Note that $f_{\chi,\gamma}(x,y) = f_{\chi}(x) f_{\gamma}(y)$, so χ and χ are independent.

3.55 (a)
$$f_{X,Y}(x,y) = 1$$
 for $(x,y) \in [5,6] \times [5,6]$.

(6)
$$P_r((X,Y) \in [5,5.5] \times [5,5.5]) = \frac{1}{4}$$

(c) For them to arrive within 10 minutes of each other, (X,Y) must be a point in the orange shaded region, $X = \frac{10}{36} \times \frac{5}{6} \times$

3.63
$$\times \sim \text{Pois}(\lambda_1)$$
 and $\times \sim \text{Pois}(\lambda_2)$, so $M_{\chi}(t) = e^{e^t \lambda_1 - \lambda_1}$, and $M_{\chi}(t) = e^{t \lambda_2 - \lambda_2}$.

By independence of \times , \times .

$$M_{X+Y}(t) = M_{X}(t) M_{Y}(t) = e^{e^{t} \lambda_{1}} \cdot e^{-\lambda_{1}} \cdot e^{e^{t} \lambda_{2}} \cdot e^{-\lambda_{2}}$$

$$= e^{e^{t} \lambda_{1}} + e^{t} \lambda_{2} \cdot e^{-(\lambda_{1} + \lambda_{2})} = e^{e^{t} (\lambda_{1} + \lambda_{2})} - (\lambda_{1} + \lambda_{2})$$

This is the mgf for another Poisson r.v. with parameter $\lambda_1 + \lambda_2$. Thus $X + Y \sim Pois(\lambda_1 + \lambda_2)$.

3.64
$$\times \text{ ~~Binom}(n,\pi,)$$
 and has $M_{\chi}(t) = (\pi,e^t+1-\pi,)^n$. As well, $Y \sim \text{Binom}(n,\pi_2)$ and has $M_{\chi}(t) = (\pi_2e^t+1-\pi_2)^n$. By independence of $X \text{ and } Y$, $M_{\chi+\chi}(t) = M_{\chi}(t) M_{\chi}(t) = \left[(\pi,e^t+1-\pi,)(\pi_2e^t+1-\pi_2) \right]^n$

$$= \left[\pi_1\pi_2e^{2t} + (\pi_1+\pi_2)e^t - 2\pi_1\pi_2e^t + 1 - (\pi_1+\pi_2) + \pi_1\pi_2 \right]^n$$

This is not the mgf of a binomial r.v.

4.1 Let $X = \frac{1}{n} \sum X_i$ be the first sample moment about the origin (a.k.a. the sample mean). The population mean for Binom (1, π) is $1 \cdot \pi = \pi$. Our estimate is $\hat{\pi} = X$.

4.4 For
$$X \sim NBinom$$
, $E(X) = \frac{\Delta}{\pi} - \Delta$, So, we set
$$\frac{\Delta}{\hat{\pi}} - \Delta = \overline{X} \qquad \Longrightarrow \qquad \hat{\pi} = \frac{\Delta}{\Delta + \overline{X}}.$$

4.7 favstats reveals $\overline{X} = 0.6091$, S = 0.248, $n = 134 \implies v = 0.06105$. From the formulas:

$$\hat{\alpha} = \overline{x} \left(\frac{\overline{x}(1-\overline{x})}{\sigma} - 1 \right) = 1.7665, \qquad \hat{\beta} = \left(1-\overline{x} \right) \left(\frac{\overline{x}(1-\overline{x})}{\sigma} - 1 \right) = 1.1337$$

The beta distribution using these shape parameters gives a very poor fit to the data. By filtering out the players with FTPct = 0, the new parameter estimates from remaining players are

$$\hat{a} = 4.824, \quad \hat{\beta} = 2.387,$$

and the fit is vastly improved.

4.9 Take
$$\overline{X} = \frac{1}{n} \sum X_i$$
 and $\overline{V} = \frac{1}{n} \sum (X_i - \overline{X})^2$.

For an exponential distribution,
$$\hat{\lambda} = \frac{1}{\bar{x}} = 0.00763$$

$$\frac{\hat{\lambda}}{\hat{\lambda}} = \bar{\lambda} \implies \hat{\lambda} = \hat{\lambda} \bar{x}$$

$$\frac{\hat{x}}{\hat{\lambda}^2} = V \implies \frac{\hat{\lambda} \times}{\hat{\lambda}^2} = \frac{\overline{y}}{\hat{\lambda}} = V$$

$$\Rightarrow \hat{\lambda} = \frac{\overline{x}}{\overline{y}} = 1.858 \quad \text{and} \quad \hat{\alpha} = \hat{\lambda} = \frac{\overline{x}^2}{\overline{y}} = 0.0141.$$