# Stat 5102 Notes: ARE of Method of Moments Estimators for the Poisson Distribution 

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What is the ARE of the two method of moments estimators compared on slides 61-62. deck 2.

These are $\bar{X}_{n}$ and $V_{n}$ considered as estimators of the mean of the Poisson distribution. The asymptotic distributions are

$$
\begin{aligned}
\bar{X}_{n} & \approx \mathcal{N}\left(\mu, \frac{\mu}{n}\right) \\
V_{n} & \approx \mathcal{N}\left(\mu, \frac{\mu_{4}-\mu^{2}}{n}\right)
\end{aligned}
$$

In order to figure out the asymptotic variance of the latter we need to calculate the fourth central moment of the Poisson distribution. We start with the moment generating function.

$$
\begin{aligned}
\varphi(t) & =E\left(e^{t X}\right) \\
& =\sum_{x=0}^{\infty} e^{x t} \frac{\mu^{x}}{x!} e^{-\mu} \\
& =\sum_{x=0}^{\infty} \frac{\left(e^{t} \mu\right)^{x}}{x!} e^{-\mu} \\
& =e^{\mu\left(e^{t}-1\right)}
\end{aligned}
$$

and this has derivatives

$$
\begin{aligned}
\varphi^{\prime}(t)= & e^{\mu\left(e^{t}-1\right)} \mu e^{t} \\
\varphi^{\prime \prime}(t)= & e^{\mu\left(e^{t}-1\right)}\left(\mu e^{t}\right)^{2}+e^{\mu\left(e^{t}-1\right)} \mu e^{t} \\
= & e^{\mu\left(e^{t}-1\right)}\left[\mu^{2} e^{2 t}+\mu e^{t}\right] \\
\varphi^{\prime \prime \prime}(t)= & e^{\mu\left(e^{t}-1\right)}\left[\mu^{2} e^{2 t}+\mu e^{t}\right] \mu e^{t}+e^{\mu\left(e^{t}-1\right)}\left[2 \mu^{2} e^{2 t}+\mu e^{t}\right] \\
= & e^{\mu\left(e^{t}-1\right)}\left[\mu^{3} e^{3 t}+3 \mu^{2} e^{2 t}+\mu e^{t}\right] \\
\varphi^{\prime \prime \prime \prime}(t)= & e^{\mu\left(e^{t}-1\right)}\left[\mu^{3} e^{3 t}+3 \mu^{2} e^{2 t}+\mu e^{t}\right] \mu e^{t} \\
& \quad+e^{\mu\left(e^{t}-1\right)}\left[3 \mu^{3} e^{3 t}+6 \mu^{2} e^{2 t}+\mu e^{t}\right] \\
= & e^{\mu\left(e^{t}-1\right)}\left[\mu^{4} e^{4 t}+6 \mu^{3} e^{3 t}+7 \mu^{2} e^{2 t}+\mu e^{t}\right]
\end{aligned}
$$

and this gives ordinary moments

$$
\begin{aligned}
\alpha_{1}=E(X)=\varphi^{\prime}(0) & =\mu \\
\alpha_{2}=E\left(X^{2}\right)=\varphi^{\prime \prime}(0) & =\mu^{2}+\mu \\
\alpha_{3}=E\left(X^{3}\right)=\varphi^{\prime \prime \prime}(0) & =\mu^{3}+3 \mu^{2}+\mu \\
\alpha_{4}=E\left(X^{4}\right)=\varphi^{\prime \prime \prime \prime}(0) & =\mu^{4}+6 \mu^{3}+7 \mu^{2}+\mu
\end{aligned}
$$

So, finally,

$$
\begin{aligned}
\mu_{4} & =E\left\{(X-\mu)^{4}\right\} \\
& =E\left(X^{4}\right)-4 \mu E\left(X^{3}\right)+6 \mu^{2} E\left(X^{2}\right)-4 \mu^{3} E(X)+\mu^{4} \\
& =\alpha_{4}-4 \mu \alpha_{3}+6 \mu^{2} \alpha_{2}-4 \mu^{3} \alpha_{1}+\mu^{4} \\
& =\left(\mu^{4}+6 \mu^{3}+7 \mu^{2}+\mu\right)-4 \mu\left(\mu^{3}+3 \mu^{2}+\mu\right)+6 \mu^{2}\left(\mu^{2}+\mu\right)-4 \mu^{3} \mu+\mu^{4} \\
& =3 \mu^{2}+\mu
\end{aligned}
$$

and the asymptotic variance of $V_{n}$ is

$$
\mu_{4}-\mu_{2}^{2}=3 \mu^{2}+\mu-\mu_{2}=2 \mu^{2}+\mu
$$

So $\bar{X}_{n}$ has smaller asymptotic variance than $V_{n}$ (for all values of $\mu$ ) and the ARE is

$$
\frac{\mu}{\mu+2 \mu^{2}}
$$

Note that the ARE goes to zero as $\mu$ goes to infinity, so $V_{n}$ gets arbitrarily bad for very large $\mu$. Thus $\bar{X}_{n}$ is not only the more obvious method of moments estimator but also the better one.

