$\mathbf{Q}\mathbf{1}$

for MoM:
$$\int_0^1 x f(x) dx = \int_0^1 x \theta x^{\theta - 1} dx = \theta \int_0^1 x \theta dx = \theta \left(\frac{x^{\theta - 1}}{\theta + 1}\right) \text{ for x 1 to 0}$$

$$\theta \left(\frac{1^{\theta + 1}}{\theta + 1}\right) = \frac{\theta}{\theta + 1} = m_1$$

$$\theta = \frac{m}{1 - m}$$

m=E(x), mean

for X we have 0.65 as mean. If we plug 0.65 as m in the theta function we will get 1.85 as the MoM estimator.

for MLE:

$$f(x) = \theta x^{\theta-1}$$

$$lnf(x) = ln\theta + (\theta - 1)lnx$$

$$lnf(x) = \sum_{i=1}^{n} (ln\theta + (\theta - 1)lnx_i)$$

$$lnf(x) = nln\theta + (\theta - 1)\sum_{i=1}^{n} (lnx_i)$$

$$\frac{\partial lnf(x)}{\partial \theta} = \frac{n}{\theta} + \sum_{i=1}^{n} (ln_i) = 0$$

$$\theta = \frac{-n}{\sum_{i=1}^{n} (lnx_i)}$$

If we plug X samples' values to this function we would $get\theta = -1.97$ as the MLE estimator.

$\mathbf{Q2}$

For the inverse function of f(x) = $2.4x^{1.4}$ $f^{-1}(x)=(\frac{x}{2.4})^{1.4}$ From our sample set, the values we would get are: 0.054, 0.143, 0.214, 0.253

$\mathbf{Q3}$

Variance of the estimators gets smaller when N increases which is expected since an increase in sample size when taking samples from same interval(P list only contains elements from the inverse function with interval $0_{i \times i}$ 1) will make the variance smaller. Mean converges to the expected value as N increases since while increasing sample size average value converges to the mean with more and more elements are added.

With the increase of N we can much clearly see which values the estimators converge to. Which are roughly 0.4 and 0.1.

I would prefer the Method of Moments on this experiments since even at small sample sizes, variance is much smaller than what we would get in Maximum Likelihood Method.