

For Normal distribution, it is obvious that $\sqrt{n} Y$ will converge to normal distribution, but Y bar will result in lower variance, which will converge to 0 when n close to infinite.

For t(2), it's variance lies between $\sqrt{n} Y$ and Y bar. As n becomes larger, $\sqrt{n} Y$ will still different from t(2) since it is sharper than normal distribution.

Q2.

(a)

```
Estimate
                    Std. Error
                                  t value
       0.215519353 0.061958219
                                3.4784627 0.0005490618
      -1.167618067 0.927018939 -1.2595407 0.2084322820
xinfl -0.379379508 0.642884239 -0.5901210 0.5553804406
xsvar -0.101604035 0.393862529 -0.2579683 0.7965392642
xtms
      -0.329207402 0.206163991 -1.5968230 0.1109472182
xtbl
      -0.317573893 0.113024303 -2.8097841 0.0051549300
xdfr
       0.275242786 0.148556414
                                1.8527829 0.0645120943
                                3.6664096 0.0002727608
       0.045320259 0.012360937
xdp
xltr
       0.126357857 0.073946585
                                1.7087720 0.0881238502
хер
      -0.002077709 0.008739102 -0.2377485 0.8121751096
      0.028790417 0.032257027 0.8925316 0.3725443638
```

For alpha = 0.05, xdp, xtbl and intecept is rejected

(b)

```
Linear hypothesis test

Hypothesis:
    xones = 0
    xdfy + xinfl = 0

Model 1: restricted model
    Model 2: y ~ (x - 1)

    Res.Df    RSS Df Sum of Sq Chisq Pr(>Chisq)
    1    494    0.97081
    2    492    0.93777    2    0.033039    17.334    0.0001722 ***
```

P-value is smaller than 5 %, meaning the hypothesis is rejected. The result is aligned with (a) since intercept = 0 is reject in single wald test. If we set only dfy and infl to 0. Wald test cannot reject the hypothesis.

Code can be found on https://github.com/YuJu0819/quant-method in folder hw6