

# Prior and Posterior Predictive Checks

**Frank Schorfheide**

Department of Economics, University of Pennsylvania

## Predictive Checks

- See, for instance, Gelman, Carlin, Stern, and Rubin (1995), Lancaster (2003), Geweke (2005).
- Prior predictive check: does the model have a chance explaining salient features of the data?
- Posterior predictive check: tries to assess the “absolute” fit of the model – similar to classical specification test.
- Recall: posterior odds are designed for relative model comparisons.

## Prior Predictive Checks

- Let  $Y^{rep}$  be a sample of observations of length  $T$  that we could have observed in the past or that we might observe in the future.
- Let's construct a predictive distribution based on our prior knowledge for  $Y^{rep}$ :

$$p(Y^{rep}) = \int p(Y^{rep}|\theta) \underbrace{p(\theta)}_{\text{Prior}} d\theta$$

- Let  $\mathcal{S}(Y)$  be a sample statistic of interest. From  $p(Y^{rep})$  we can derive the predictive distribution of  $p(\mathcal{S})$ .
- Compute the observed value of  $\mathcal{S}$  based on the actual data and assess how far it lies in the tails of its predictive distribution.

## Posterior Predictive Checks

- Let  $Y^{rep}$  be a sample of observations of length  $T$  that we could have observed in the past or that we might observe in the future.
- Let's construct a predictive distribution based on our posterior knowledge for  $Y^{rep}$ :

$$p(Y^{rep}) = \int p(Y^{rep}|\theta) \underbrace{p(\theta|Y)}_{\text{Posterior}} d\theta$$

- Let  $\mathcal{S}(Y)$  be a sample statistic of interest. From  $p(Y^{rep})$  we can derive the predictive distribution of  $p(\mathcal{S})$ .
- Compute the observed value of  $\mathcal{S}$  based on the actual data and assess how far it lies in the tails of its predictive distribution.

## Prior (Posterior) Predictive Checks

- Implementation: for  $s = 1$  to  $n_{sim}$ :
  1. Generate a draw  $\theta^{(s)}$  from prior (posterior).
  2. Simulate data  $Y^{(s)}$  from model conditional on  $\theta^{(s)}$ .
  3. Compute  $\mathcal{S}(Y^{(s)})$ .

## Posterior Predictive Checks

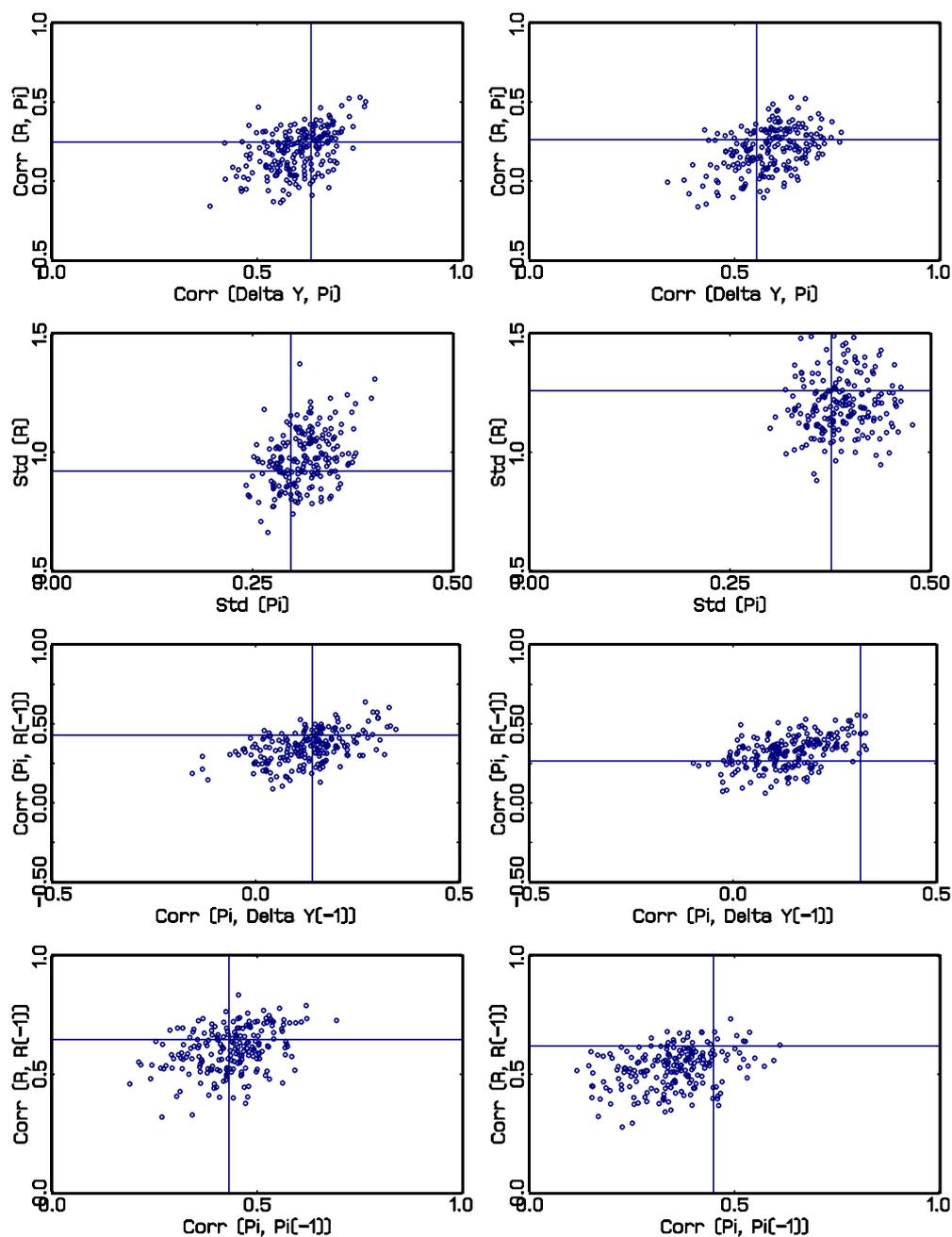
- Let  $Y^{rep}$  be a sample of observations of length  $T$  that we could have observed in the past or that we might observe in the future.
- Derive the sampling distribution of  $Y^{rep}$  given the current state of knowledge:

$$p(Y^{rep}|Y) = \int p(Y^{rep}|\theta) \underbrace{p(\theta|Y)}_{\text{Posterior}} d\theta. \quad (1)$$

- Let  $\mathcal{S}(Y)$  be a test quantity and compute predictive distribution for  $\mathcal{S}(Y)$ .
- Implementation: same as for prior predictive check – replace  $\theta^{(s)}$  draws from prior with draws from posterior.

(insert figures)

Figure 8: POSTERIOR PREDICTIVE CHECK



*Notes:* Output gap rule specification  $\mathcal{M}_1$ . We plot 200 draws from the posterior predictive distribution of various sample moments. Intersections of solid lines signify the observed sample moments. Left panels: Data Set 1- $\mathcal{M}_1$ , no misspecification. Right panels: Data Set 2, misspecification.