# Business Cycles: Empirics and Theory

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### 1 Introduction

Some jargons:

- $\bullet$  ar = annual rate
- $\bullet$  na = not adjusted
- sa = seasonally adjusted (Swiss case: sa = sport-event adjusted)
- q/q sa quarterly seasonally adjusted growth
- q/q csa quarterly seasonally and calendar adjusted growth
- q/q saar = quarterly seasonally growth expressed at an annual rate

$$q/q \text{ saar} = (1 + q/q \text{ sa})^4 - 1$$

• y/y = growth relative to the same quarter of the previous year (usually calculated using na series)

Some examples:

- $Y_{\mathrm{Q2~2018}} = \text{level of real GDP in Q2~2018}$
- $4 \times Y_{\rm Q2~2018}$  = level of real GDP in Q2 2018, expressed at an annual rate
- $Y_{\mathrm{O2\ 2018}}^{sa} = \text{seasonally adjusted level of real GDP in Q2\ 2018}$

• q/q growth in Q2 2018 (sa) = 
$$\left(\frac{Y_{\text{Q2 2018}}^{sa}}{Y_{\text{Q1 2018}}^{sa}}\right)^4 - 1$$

• y/y growth = 
$$\frac{Y_{\text{Q2 2018}}}{Y_{\text{Q2 2017}}} - 1$$

$$\bullet \text{ annual growth} = \frac{Y_{\mathrm{Q1\ 2018}} + Y_{\mathrm{Q2\ 2018}} + Y_{\mathrm{Q3\ 2018}} + Y_{\mathrm{Q4\ 2018}}}{Y_{\mathrm{Q1\ 2017}} + Y_{\mathrm{Q2\ 2017}} + Y_{\mathrm{Q3\ 2017}} + Y_{\mathrm{Q4\ 2017}}} - 1$$

Some notations:

- $X_t$  = level in period t, e.g., GDP during Q1 2018
- $x_t = \log(X_t) = \text{natural logarithm of } X_t$
- $\bullet \ \Delta x_t = x_t x_{t-1}$
- The growth rate of  $X_t$

$$\frac{X_t - X_{t-1}}{X_{t-1}} \simeq \log \frac{X_t}{X_{t-1}} = \Delta x_t$$

Some useful approximations:

- y/y growth = 4 quarter moving average of q/q ar growth = sum of q/q growth rates =  $\Delta x_t + \Delta x_{t-1} + \Delta x_{t-2} + \Delta x_{t-3}$
- Some definitions to write down approximations to annual growth rates
  - $-\ g_{t,q}^{q/q} \equiv {\bf q}/{\bf q}$  growth rate in  $q{\bf th}$  quarter of the year t
  - $-g_{t,q}^{y/y} \equiv y/y$  growth rate in qth quarter of the year t
  - $-g_{t,q}^{A} \equiv \text{annual growth rate in year } t$
- The following approximations hold

$$-\ g_t^A \simeq \tfrac{1}{4} (g_{t,4}^{q/q} + 2g_{t,3}^{q/q} + 3g_{t,2}^{q/q} + 4g_{t,1}^{q/q} + 3g_{t-1,4}^{q/q} + 2g_{t-1,3}^{q/q} + g_{t-1,2}^{q/q})$$

$$-g_t^A \simeq \frac{1}{4}(g_{t,4}^{y/y} + g_{t,3}^{y/y} + g_{t,2}^{y/y} + g_{t,1}^{y/y})$$

Important implications of approximations to annual growth rate

- Annual growth is not the average of the quarterly growth rates of the respective year.
- Annual growth depends also on quarterly growth in the previous year.
- In fact, the previous year's Q4 growth rate is three times as important for this year's annual growth than this year's Q4 growth rate.
- Once you know growth up to Q2, you can already make a very precise forecast for annual growth (you already know 13/16 of the annual growth rate).
- Similar effect for y/y growth rates (you always know already 3/4 of the next y/y growth rate).

# 2 Detrending

#### Linear trend

$$\log(Y_t) = \beta_0 + \beta_1 t + \varepsilon_t$$

- $\beta_1$  is the trend growth rate.
- $\varepsilon_t$  are the deviations from the trend and therewith the cycle.

#### Time-varying trend

$$c_t \equiv \frac{Y_t - Z_t}{Z_t}$$

- $Z_t$ : the "trend level" of GDP.
- $c_t$ : the percentage deviation of GDP from trend by  $c_t$ .

Using log-approximations:

$$c_t = \frac{Y_t}{Z_t} - 1 \simeq \log \frac{Y_t}{Z_t} = y_t - z_t$$

$$y_t = \log(Y_t) = z_t + c_t$$

i.e., log-level of GDP consists of a trend  $z_t$  and a cyclical component  $c_t$  that fluctuates around the trend.

### How to govern the degree of time variation in trend growth

Trend growth rate:

$$\frac{Z_t}{Z_{t-1}} - 1 \simeq \log \frac{Z_t}{Z_{t-1}} = z_t - z_{t-1}$$

The change of the trend growth rate:

$$v_t = (z_t - z_{t-1}) - (z_{t-1} - z_{t-2})$$

If we allow for a lot of time-variation in the trend growth rate, we can choose  $v_t$  such that most of the observed changes in  $\log(Y_t)$  is attributable to changes in  $z_t$  and only little to cyclical component  $c_t$  and vice versa.

#### **HP** filter

$$L = \sum_{t} c_{t}^{2} + \lambda \sum_{t} v_{t}^{2}$$

$$= \sum_{t} (\log(Y_{t}) - z_{t})^{2} + \lambda \sum_{t} ((z_{t} - z_{t-1}) - (z_{t-1} - z_{t-2}))^{2}$$

$$\hat{z}_t = \underset{z_t}{\operatorname{arg\,min}} \quad L$$

For  $\lambda \to 0$ :

$$\min_{z_t} L = \min_{z_t} \sum_{t} (\log(Y_t) - z_t)^2$$

- The time trend  $z_t$  that minimizes the loss function is simply  $\log(Y_t) = z_t$ .
- There is no cycle!

For  $\lambda \to \infty$ :

$$\min_{z_t} L = \min_{z_t} \sum_{t} ((z_t - z_{t-1}) - (z_{t-1} - z_{t-2}))^2$$

- The loss function is minimized with  $z_t z_{t-1} = z_{t+j} z_{t+j-1} \quad \forall t, j$ .
- Trend growth is constant, i.e., linear time trend.

Problems with HP filters:

- Instability of the HP trend at the margin.
- HP filter cannot separate demand from supply shocks (i.e., a slowdown in trend growth is identified during a long-lasting recession).
- What is the right  $\lambda$  for HP filter.

#### Production function approach

Cobb-Douglas production function:

$$Y = AK^{\alpha}L^{1-\alpha}$$

- Measure K and L (which is not trivial) and solve for A.
- Filter A, K, L to get their potential levels.

• Plug these back into production function to get potential GDP.

### Another way of writing the HP filter

Assume that the change in the trend is a random variable:

$$\Delta z_t - \Delta z_{t-1} = v_t, v_t \sim \mathcal{N}(0, \sigma^2)$$

$$(z_t - z_{t-1}) - (z_{t-1} - z_{t-2}) = v_t$$

$$\underbrace{\begin{bmatrix} z_t \\ z_{t-1} \end{bmatrix}}_{Z_t} = \underbrace{\begin{bmatrix} 2 & -1 \\ 1 & 0 \end{bmatrix}}_{F} \underbrace{\begin{bmatrix} z_{t-1} \\ z_{t-2} \end{bmatrix}}_{Z_{t-1}} + \underbrace{\begin{bmatrix} \sigma & 0 \\ 0 & 0 \end{bmatrix}}_{Q} \underbrace{\begin{bmatrix} \varepsilon_t^{\Delta z} \\ 0 \end{bmatrix}}_{w_t}$$

$$Z_t = FZ_{t-1} + Qw_t, w_t \sim \mathcal{N}(0, 1)$$

We do not observe the trend  $Z_t$ , but we observe data that is a function of the unobserved trend:

$$\log(Y_t) = z_t + c_t = \underbrace{\begin{bmatrix} 1 & 0 \end{bmatrix}}_{H} \underbrace{\begin{bmatrix} z_t \\ z_{t-1} \end{bmatrix}}_{Z_t} + \underbrace{\begin{bmatrix} \sigma & 0 \end{bmatrix}}_{R} \underbrace{\begin{bmatrix} \lambda \varepsilon_t^c \\ 0 \end{bmatrix}}_{e_t}$$

$$\log(Y_t) = HZ_t + Re_t, e_t \sim \mathcal{N}(0, 1)$$

Note: variance of  $c_t$  is  $\lambda$  times larger than variance of  $v_t$ .

### The state-space form

Measurement equation with standard normal i.i.d. shocks  $e_t$ :

$$\log(Y_t) = X_t = HZ_t + Re_t$$

Transition equation with standard normal i.i.d. shocks  $w_t$ :

$$Z_t = FZ_{t-1} + Qw_t$$

Note:

- Whenever, we can write something in state-space form, we can use Kalman filter to estimate the unobserved states  $Z_t$  given the parameter matrices H, R, F, Q and the data  $X_t$ .
- Kalman filter can easily handle missing data.

# 3 Recessions v.s. Expansions

#### Recessions

- Periods with contracting economic activity.
- Popular rule of thumb: two consecutive quarters of negative real GDP growth.

### Regime switching model (Hamilton filter)

- Assume that there are two regimes  $S_t = 1, 2$
- Assume that the economy stochastically switches between the two regimes with the following probabilities

Regimes differ in average growth:

$$\Delta y_t = \begin{cases} c_1 + \varepsilon_t & S_t = 1\\ c_2 + \varepsilon_t & S_t = 2 \end{cases}, \varepsilon_t \sim \mathcal{N}(0, \sigma^2)$$

### Comparison

Linear view:

- Business cycles are fluctuations around a trend.
- Most modern macro models (DSGEs) implicitly have this view.

Non-linear view:

- Business cycles are transitions between two (or more) distinct regimes.
- While there are fluctuations within a regime, what really matters is the transition between the regimes.
- The public and high-level policy makers often take this view.

### One of the best forecasting indicators for recessions: the yield curve

- Yield curve slope is historically one of the best indicators for upcoming recessions.
- All recessions were preceded by an inverted (negatively sloped) yield curves, yield curve was rarely inverted without a recession following.

### Why is an inverted yield curve a good recession predictor?

Assume you have 100 USD to invest for two years

There are two options:

- Option 1: buy a 2-year treasury bond with a yield of  $i_t^{2y}$ , get the principal payment  $2(100(1+i_t^{2y})^2)$  after 2 years.
- Option 2: buy a 1-year treasury bond with a yield of  $i_t^{1y}$ , get the principal payment  $100(1+i_t^{1y})$  after 1 year and reinvest it in a new 1-year treasury bond with an expected yield  $\mathbf{E}\left[(1+i_{t+1}^{1y})\right]$ . You expect to have  $100(1+i_t^{1y})\mathbf{E}\left[(1+i_{t+1}^{1y})\right]$  after 2 years.

Except for the uncertainty about  $i_{t+1}^{1y}$ , the two options are very similar, therefore the return has to be similar

$$100(1+i_t^{2y})^2\approx 100(1+i_t^{1y})\mathbf{E}\left[(1+i_{t+1}^{1y})\right]$$

Take logs of both sides:

$$2\log(1+i_t^{2y}) \approx \log(1+i_t^{1y}) + \log\left(1+\mathbf{E}\left[i_{t+1}^{1y}\right]\right)$$

Use the approximation equation  $\log(1+x) = x$ :

$$i_t^{2y} \approx \frac{1}{2} \left( i_t^{1y} + \mathbf{E} \left[ i_{t+1}^{1y} \right] \right)$$

Similarly, for 10-year yield:

$$i_t^{10y} \approx \frac{1}{10} \sum_{i=0}^{9} \mathbf{E}_t \left[ i_{t+i}^{1y} \right]$$

- Short-term yields are mostly driven by monetary policy  $(i_t^{1y} \approx \text{policy rate})$ .
- An inverted yield curve implies that the policy rate is currently higher than what is expected in the future.
- In other words, market participants expect the central bank to lower the policy rate.
- Central banks typically only lower policy rates in response to economic weakness.
- Word of caution: the risk-premium may lead to an inverted yield curve, without market participants expecting falling policy rates

$$i_t^{10y} = \frac{1}{10} \sum_{i=0}^{9} \mathbf{E}_t i_{t+i}^{1y} + \text{risk premium}_t$$

Recession nowcasting is feasible!

Good rule of thumb: a recession has started if 3-month change of the 3-month moving average of seasonally adjusted unemployment rate rises above 0.3 pp.

### Important references: baseline recession probabilities

NBER recession

- Time span: from Q1 1950 until Q2 2021 (286 quarters in total)
- # of expansion quarters = 247
- # of recession quarters = 39
- Unconditional recession probability =  $\frac{39}{286} = 14\%$
- $\bullet$  # of expansion quarters that are followed by a recession = 11
- Conditional of being in an expansion quarter:
  - probability of a recession starting next quarter =  $\frac{11}{247} = 4.5\%$
  - probability of staying in an expansion = 100 4.5 = 95.5%
- Conditional on being in an expansion in t:
  - probability of next recession starting sometimes during next 2 quarters  $(t+1 \text{ and } t+2) = 1-0.955 \times 0.955$
  - probability of next recession starting sometimes during next n quarters (between t+1 and t+n) =  $1-0.955^n$

# 4 Welfare Costs of Business Cycles

HP decomposition:  $\log C_t = z_t + x_t$  (or equally:  $C_t = Z_t X_t$ ), where  $z_t$  is the trend level of consumption and  $x_t$  fluctuations around this trend.

The Lucas calculation - approximating the utility function

$$U(C) \simeq U(Z) + U'(Z)(ZX - Z) + \frac{1}{2}U''(Z)(ZX - Z)^2$$

Calculating expected utility:

$$\mathbf{E}[U(C)] \simeq U(Z) + U'(Z)\mathbf{E}[(ZX - Z)] + \frac{1}{2}U''(Z)\mathbf{E}\left[(ZX - Z)^2\right]$$

Rewriting:

$$\mathbf{E}[U(C)] \simeq U(Z) + \frac{1}{2}U''(Z)\mathbf{E}[(X-1)^2]Z^2$$

Utility function:

$$U(C) = \frac{C_{1-\sigma}}{1-\sigma}$$

Hence,

$$\mathbf{E}[U(C)] \simeq \frac{Z^{1-\sigma}}{1-\sigma} - \frac{\sigma}{2} Z^{-\sigma-1} \mathbf{E} \left[ (\underbrace{X-1}_{\simeq \log X = x})^2 \right] Z^2$$
$$\mathbf{E}[U(C)] \simeq \frac{Z_{1-\sigma}}{1-\sigma} - \frac{\sigma}{2} Z^{1-\sigma} \mathbf{E}[x^2]$$

- $\mathbf{E}[x^2]$  is the variance of log-difference ( $\simeq$  percentage deviation) of consumption from its trend level
- The higher the variance, the lower expected utility
- The higher risk aversion, sigma, the more costly are fluctuations

### Certainty equivalence

Let us scale trend level of consumption by  $\delta$  and offer  $\delta Z$  with certainty to the representative consumer

$$\frac{(\delta Z)^{1-\sigma}}{1-\sigma} = \frac{Z_{1-\sigma}}{1-\sigma} - \frac{\sigma}{2} Z^{1-\sigma} \mathbf{E}[x^2]$$

Rearrange:

$$\delta = \left(1 - (1 - \sigma)\frac{\sigma}{2}\mathbf{E}[x^2]\right)^{\frac{1}{1 - \sigma}}$$

The resulting estimate for the costs of economic fluctuations is tiny, but why should we care about business cycles?

- Observed fluctuations are the ones resulting despite macroeconomic stabilization policies (might be much larger without stabilization)
- Cost of business cycles are not equally distributed over population (some become unemployed, some not; not everyone becomes 4% unemployed, i.e. can work 4% fewer hours)
- Direct costs of unemployment
- Persistent effects of economic fluctuations

### 5 Comovement

### Principal Component Analysis

$$y_{it} = \lambda_i f_t + \varepsilon_{it}$$

- $y_{it}$ : the standardized value of the indicator i at time t
- $f_t$ : the underlying force that drives the common variation in the dataset
- $\lambda_i$ : the "loading" of indicator i on  $f_t$

Assume that we know the loadings  $\lambda_i$ :

$$y_t = \begin{bmatrix} y_{1t} \\ \vdots \\ y_{Nt} \end{bmatrix} = \begin{bmatrix} \lambda_1 \\ \vdots \\ \lambda_N \end{bmatrix} f_t + \begin{bmatrix} \varepsilon_{1t} \\ \vdots \\ \varepsilon_{Nt} \end{bmatrix}$$

For given loadings  $\lambda$ , we can obtain  $f_t$  by simply running a regression in each time period t

$$\hat{f}_t = (\lambda'\lambda)^{-1}\lambda' y_t$$

$$\hat{\varepsilon}_t = y_t - \lambda (\lambda' \lambda)^{-1} \lambda' y_t$$

The sum of squared errors:

$$\sum_{t} \hat{\varepsilon}'_{t} \hat{\varepsilon}_{t} = \sum_{t} (y_{t} - \lambda(\lambda'\lambda)^{-1}\lambda'y_{t})' (y_{t} - \lambda(\lambda'\lambda)^{-1}\lambda'y_{t})$$

# 6 Swiss GDP

### How to estimate Swiss GDP

- Annual GDP adds up the value added according to firm survey
- Value added is estimated from the production side
- Demand side: inventories as residual
- Income side: profits as residual
- First estimate published 8 months after the end of reference year
- i.e. first estimate for 2021 will be published at the end of August 2022

### Quarterly GDP is an intra- and extra-polation of annual GDP

• Regress annual value added of a sector  $y_t^A$  on indicators

$$y_t^A = \beta_0 + \beta_1 x_t^A + \varepsilon_t$$

# 7 Introduction to Theory

IS curve: goods & services market equilibrium

Aggregate demand equals investment, private consumption and government spending (closed economy):

$$Y^D = I + C + G$$

Aggregate demand depends on aggregate income  $Y^{I}$ , interest rate, r, and other factors,  $v_{D}$ 

$$Y^D = f(Y_+^I, \underline{r}) + v_D$$

In equilibrium:

$$\underbrace{\operatorname{aggregate \ demand}}_{Y^D} = \underbrace{\operatorname{aggregate \ income}}_{Y^I} = \underbrace{\operatorname{aggregate \ production}}_{Y^P}$$

LM curve: money market Monetary policy rule (MP-rule)

$$r = \rho + \theta_{\pi}\pi + v_{MP}$$

- $\pi$ : inflation
- $v_{MP}$ : monetary policy shock

The IS-LM model provides the equilibrium combination of interest rate and inflation for a given demand and monetary policy shock.

- The IS-LM model describes goods and money market equilibrium in the short-term
- Inflation is assumed to be constant

#### AS-AD model:

- The AS-AD model is an extension of the IS-LM framework
- It describes the relationship between aggregate output and prices

AD curve: aggregate demand

- The AD curve is defined by the IS-LM equilibrium
- For given inflation rate,  $\pi$ , the MP rule describes, what interest rate, r, should be chosen by the central bank
- From IS curve, we can get the resulting aggregate demand, Y
- The AD curve describes the resulting aggregate demand for each inflation rate
- The monetary policy shock and the demand shock shift the AD curve

#### AS curve

- The inflation rate follows from the pricing decisions of producers
- The AS curve describes the change in prices (i.e., inflation)
- $\bullet$  Prices  $\to$  aggregate output of Y produced by firms
- If firms produce a lot relative to some benchmark level of output  $(\bar{Y})$ , they tend to raise prices

$$\pi = \bar{\pi} + \kappa (Y - \bar{Y})$$

# Assessing the Economic Effects of the Coronavirus with the AS-AD Model Split the economy into 2 groups:

- directly affected sectors: businesses that are directly affected by containment measures (e.g., restaurants)
- non-affected sectors: all other businesses (e.g., universities can switch to online education, insurances to home office)

# 8 New Keynesian IS Curve

### Household utility maximization problem:

$$\max_{\{C_{t+j}, B_{t+j}, N_{t+j}\}_{j=0}^{\infty}} \mathbb{E}_t \left[ \sum_{j=0}^{\infty} \beta^j Z_{t+j} \left( \frac{C_{t+j}^{1-\sigma}}{1-\sigma} - \frac{N_{t+j}^{1+\phi}}{1+\phi} \right) \right]$$
s.t. 
$$P_{t+j} C_{t+j} + Q_{t+j} B_{t+j} \le B_{t-1+j} + W_{t+j} N_{t+j} + D_{t+j}$$

- $Z_t$ : preference shifter
- $Q_t$ : Bond price in t, pays 1 in t+1
- $B_t$ : quantity of bonds purchased in t
- $N_t$ : labor supply
- $W_t$ : hourly wage payment
- $P_t$ : the price of a consumption bundle  $C_t$
- $D_t$ : dividends paid by firms to households

### Optimal consumption:

$$C_t^{-\sigma} = \mathbb{E}_t \left[ \beta (1+i_t) \frac{Z_{t+1}}{Z_t} \frac{1}{1+\pi_{t+1}} C_{t+1}^{-\sigma} \right]$$
 Euler equation

- $\beta$  low: impatient, higher consumption today.
- $\frac{Z_{t+1}}{Z_t}$  high: assign a lot to today's utility relative to tomorrow.
- $i_t$  high: less consumption and more saving today.
- $\pi_{t+1}$  high: tomorrow's prices high, more consumption today.
- $\sigma$  low: households substitute consumption today more strongly with consumption tomorrow when the interest rate is raised ( $\sigma$  equals the inverse elasticity of substitution).

### Optimal labor:

$$\frac{N_t^{\phi}}{C_t^{-\sigma}} = \frac{W_t}{P_t}$$

Log-linear form of Euler equation (using first-order Taylor approximation and defining  $\log \beta \equiv \rho$ ):

$$c_t \simeq \mathbb{E}_t c_{t+1} - \frac{1}{\sigma} (i_t - \mathbb{E}_t \pi_{t+1} - \rho) + \frac{1 - \rho_z}{\sigma} z_t$$
 NKIS

# 9 New Keynesian Phillips Curve

Optimal allocation of spending across varieties:

$$\max_{C_t(i)} C_t = \left( \int_{i=0}^{1} C_t(i)^{\frac{\varepsilon-1}{\varepsilon}} di \right)^{\frac{\varepsilon}{\varepsilon-1}}$$
s.t. 
$$\int_{i=0}^{1} P_t(i) C_t(i) di \le S_t$$

Optimal consumption of good i:

$$C_t(i) = C_t \left(\frac{P_t(i)}{P_t}\right)^{-\varepsilon}$$

where

$$P_t \equiv \left( \int_{i=0}^1 P_t(i)^{1-\varepsilon} \mathrm{d}i \right)^{\frac{1}{1-\varepsilon}}$$

Good market clearing:  $Y_t(i) = C_t(i), Y_t = C_t$ 

$$Y_t(i) = Y_t \left(\frac{P_t(i)}{P_t}\right)^{-\varepsilon}$$
 Demand function

$$Y_t(i) = A_t L_t(i)$$
 Production function

Profit maximization under flexible prices:

$$\max_{P_t(i)} \quad \Pi_t(i) = P_t(i)Y_t(i) - W_tL_t(i)$$

$$\max_{P_t(i)} P_t(i) Y_t \left(\frac{P_t(i)}{P_t}\right)^{-\varepsilon} - \frac{W_t}{A_t} Y_t \left(\frac{P_t(i)}{P_t}\right)^{-\varepsilon}$$

Profit-maximizing price:

$$P_t^*(i) = \frac{\varepsilon}{\varepsilon - 1} \frac{W_t}{A_t}$$

Aggregate production under flexible prices:

$$Y_t = A_t L_t$$

Labor market equilibrium:  $L_t = N_t, Y_t = C_t$ 

$$\begin{cases} \frac{N_t^{\varphi}}{C_t^{-\sigma}} = \frac{W_t}{P_t} \\ P_t^*(i) = \frac{\varepsilon}{\varepsilon - 1} \frac{W_t}{A_t} \\ Y_t = A_t L_t \end{cases} \implies \begin{cases} L_t = \left( A_t^{1-\sigma} \frac{\varepsilon - 1}{\varepsilon} \right)^{\frac{1}{\varphi + \sigma}} \\ Y_t = A_t^{\frac{1+\varphi}{\varphi + \sigma}} \left( \frac{\varepsilon - 1}{\varepsilon} \right)^{\frac{1}{\varphi + \sigma}} \end{cases}$$

**Key insight:** Equilibrium production does not depend on prices.

Profit maximization under sticky prices:

$$\max_{P_t(i)} PV = \mathbb{E}_t \left[ \sum_{j=0}^{\infty} \frac{\theta^j}{R_{t+j}} \left( P_t(i) Y_{t+j} \left( \frac{P_t(i)}{P_{t+j}} \right)^{-\varepsilon} - \frac{W_{t+j}}{A_{t+j}} Y_{t+j} \left( \frac{P_t(i)}{P_{t+j}} \right)^{-\varepsilon} \right) \right]$$

where

$$R_{t+j} = \prod_{\tau=1}^{j} (1 + i_{t+\tau} - \mathbb{E}_t \pi_{t+\tau})$$

Optimal price under sticky prices:

$$O_t = \mathbb{E}_t \left[ \sum_{j=0}^{\infty} \omega_j P_{t+j}^* \right]$$

where

$$\omega_j = \frac{\frac{\theta^j}{R_{t+j}} Y_{t+j} P_{t+j}^{\varepsilon}}{\sum_{j=0}^{\infty} \frac{\theta^j}{R_{t+j}} Y_{t+j} P_{t+j}^{\varepsilon}}$$

- $\theta^{j}$ : probability that price is still in place in t+j
- $\frac{1}{R_{t+j}}$ : the present value of the profit that is generated in t+j
- $Y_{t+j}P_{t+j}^{\varepsilon}$ : how many units a firm can sell for a given price in t+j

Log-linear approximation of the optimal price:

$$o_t \simeq \sum_{j=0}^{\infty} \frac{(\beta \theta)^j}{\sum_{k=0}^{\infty} (\beta \theta)^k} \mathbb{E}_t p_{t+j}^*$$

$$\frac{\frac{\theta^{j}}{R_{t+j}}Y_{t+j}P_{t+j}^{\varepsilon}}{\sum_{j=0}^{\infty}\frac{\theta^{j}}{R_{t+j}}Y_{t+j}P_{t+j}^{\varepsilon}} \implies \frac{(\beta\theta)^{j}}{\sum_{k=0}^{\infty}(\beta\theta)^{k}}$$

- In steady state  $Y_{t+j}P_{t+j}^{\varepsilon}$  is constant and therefore cancels out
- In steady state we have  $i = -\log \beta = \log \frac{1}{\beta} \simeq \frac{1}{\beta} 1$  and  $\pi = 0$ , hence  $1 + i = \frac{1}{\beta}$

$$R_{t+j} = \prod_{\tau=1}^{j} (1+i-\pi) = \prod_{\tau=1}^{j} \frac{1}{\beta} = \left(\frac{1}{\beta}\right)^{j} \Longleftrightarrow \frac{1}{R_{t+j}} = \beta^{j}$$

•  $\theta^{j}$ : probability that price is still in place after j periods

$$o_{t} \simeq \sum_{j=0}^{\infty} \frac{(\beta \theta)^{j}}{\sum_{k=0}^{\infty} (\beta \theta)^{k}} \mathbb{E}_{t} p_{t+j}^{*}$$

$$= (1 - \beta \theta) \sum_{k=0}^{\infty} (\beta \theta)^{k} \mathbb{E}_{t} p_{t+j}^{*}$$

$$= (1 - \beta \theta) p_{t}^{*} + (1 - \beta \theta) \sum_{k=1}^{\infty} (\beta \theta)^{k} \mathbb{E}_{t} p_{t+j}^{*}$$

$$= (1 - \beta \theta) p_{t}^{*} + \beta \theta \mathbb{E}_{t} o_{t+1}$$

- The optimal price today is a weighted average of today's flexible price and tomorrow's optimal price
- Higher discounting  $(\beta \downarrow)$  and more often firms adjust  $(\theta \downarrow)$ , the larger weight on today's flexible price

#### Aggregate price dynamics:

The price index in t is a weighted average between the price chosen by the adjusting firms  $(o_t)$  and last period's price index  $(p_{t-1})$ 

$$p_t \simeq (1 - \theta)o_t + \theta p_{t-1}$$

$$\underbrace{p_t - p_{t-1}}_{\pi_t} \simeq (1 - \theta)(o_t - p_{t-1})$$

$$\pi_t \simeq (1 - \theta)(o_t - p_{t-1})$$

Intuition: inflation is driven by

- $1 \theta$ : how many adjusting firms reset prices
- $o_t p_{t-1}$ : how much adjusting firms change prices

Iterate above equation one period forward:

$$\mathbb{E}_t \pi_{t+1} = (1 - \theta)(\mathbb{E}_t o_{t+1} - p_t)$$

$$\begin{cases} o_t = (1 - \beta \theta) p_t^* + \beta \theta \mathbb{E}_t o_{t+1} \\ \pi_t = (1 - \theta) (o_t - p_{t-1}) \\ \mathbb{E}_t \pi_{t+1} = (1 - \theta) (\mathbb{E}_t o_{t+1} - p_t) \end{cases} \implies \pi_t = \frac{1 - \theta}{\theta} (1 - \beta \theta) (p_t^* - p_t) + \beta \mathbb{E}_t \pi_{t+1}$$

Inflation is high whenever

- expected inflation is high (comes from adjusting firms choosing high  $o_t$  because they expect high  $\mathbb{E}_t o_{t+1}$ )
- current flexible prices are high relative to the price level  $(p_t^* p_t)$  (comes from adjusting firms choosing high  $o_t$  because  $p_t^*$  is high)

Labor market equilibrium:  $L_t = N_t, Y_t = C_t$ 

$$\begin{cases} \frac{N_t^{\varphi}}{C_t^{-\sigma}} = \frac{W_t}{P_t} \\ P_t^*(i) = \frac{\varepsilon}{\varepsilon - 1} \frac{W_t}{A_t} \end{cases} \implies \begin{cases} \frac{P_t^*}{P_t} = \frac{\varepsilon}{\varepsilon - 1} A_t^{-\varphi - 1} Y_t^{\varphi + \sigma} \\ 1 = \frac{\varepsilon}{\varepsilon - 1} A_t^{-\varphi - 1} (Y_t^n)^{\varphi + \sigma} \end{cases} \implies \frac{P_t^*}{P_t} = \left(\frac{Y_t}{Y_t^n}\right)^{\varphi + \sigma}$$

Log-linearize above equation:

$$p^* - p_t = (\varphi + \sigma) \underbrace{(y_t - y_t^n)}_{\widetilde{y}_t}$$

$$\pi_t = \kappa \widetilde{y}_t + \beta \mathbb{E}_t \pi_{t+1}$$
 NKPK

where

$$\kappa = \frac{1 - \theta}{\theta} (1 - \beta \theta)(\varphi + \sigma)$$