

# **Model Selection in Estimating the Certainty Equivalent Discount Rate**

**Ben Groom (UCL)**

**Phoebe Koundouri (University of Reading)**

**Ekaterini Panopoulou and Theo Pantelides (University of  
Piraeus, Greece)**

**ENVECON Conference  
26<sup>th</sup> March 2004  
Royal Society, London**

## Background (1)

- Emergence of a long-term policy arena: e.g. climate change, biodiversity conservation, long-lived infrastructure.
- Conventional discounting gives CBA a 'defective telescopic faculty' (Pigou 1920). A 'tyranny' of current over the future generations (Chichilnisky 1996)
- Declining Discount Rates (DDRs): a path through the 'dark jungles of the second best' (Baumol 1968)?
- DDRs in the Green Book (2003) for long-term projects  
→ how can practitioners define the schedule of DDRs?

### Rationales for DDRs (see e.g. Pearce et al 2003)

- 1) 'Hyperbolic' Discounting (Loewenstein and Prelec 1992, Laibson 1997, etc)
- 2) Social Choice (e.g. Chichilnisky 1997, Li and Löfgren 2000, 2001): DDRs appeal to certain axioms of social choice and sustainability: non-dictatorship
- 3) Uncertainty:
  - **Growth:** Gollier (2002). DDRs depend upon risk
  - **Discount Rate:** Weitzman (1998). The 'certainty equivalent' rate is a declining function of time.

## Background (2):

### Determining the schedule of DDRs

Some theories are easier to implement than others:

- **Gollier**: detailed knowledge of the utility function (risk and precaution preferences: 5<sup>th</sup> derivatives of utility!!),
- **Social Choice**: requires a number of arbitrary choices concerning social weights
- **Weitzman**: only requires Characterisation of uncertainty of  $r$ .

Weitzman approach has received the most attention, but with alternative interpretations of uncertainty:

- i) **Weitzman (AER 2001)**: 'Gamma Discounting': survey of 2000 Economists (with a 'Blue-Ribbon' group of 50) → current uncertainty/disagreement about the rate to use **now**
- ii) **Newell and Pizer (JEEM 2003)**: We are pretty certain about today's discount rate, but uncertain about the future → Time series analysis: the past is informative about the future. Resulting schedule increases the value of carbon reduction by 84% compared to constant discounting.

# Overview

- 1) **Summary**
- 2) **The Certainty Equivalent Discount Rate**
- 3) **Newell and Pizer's (JEEM 2003) approach**
- 4) **Importance of Model Selection**
- 5) **Method**
- 6) **Data**
- 7) **Results (US) and Simulations**
- 8) **Model Selection**
- 9) **Policy Implications**
- 10) **Conclusion**

# 1. Summary

We focus on the N&P (2003) approach to estimating the schedule of DDRs. We make the following points concerning the estimation of the certainty equivalent discount rate:

- 1) If we believe that the past is informative about the future we would be well advised to characterise the past as realistically as possible. Characterising uncertainty is important for operationalising a theory of DDRs that depends based on uncertainty and which relies upon the forecasting over long-time horizons (400 years +).
- 2) Specification testing in the case of US and UK interest rate data provides a natural progression towards models that account for changes in the data generating process: structural breaks, regime shifts etc.
- 3) We compare N&P's model to the well specified alternatives and compare the outcome of CBA for climate change (US) and nuclear power (UK) in order to illustrate the importance of model selection.
- 4) Model selection can have important policy implications for the long-term policy arena

## 2. The Certainty Equivalent Discount Rate

Under certainty the discount factor at time  $t$ ,  $P_t$ , is:

$$P_t = \exp\left(-\sum_{i=1}^t r_i\right)$$

When  $r$  is stochastic then the Certainty Equivalent Discount factor for a **risk neutral** agent becomes:

$$E[P_t] = E\left[\exp\left(-\sum_{i=1}^t r_i\right)\right]$$

The Certainty Equivalent Discount **Rate**,  $r_w(t)$ , is defined by:

$$-\frac{dE(P_t)/dt}{E(P_t)} = r_w(t)$$

**Weitzman and N&P:** With sufficient persistence this is a declining function of time. With the gamma discounting this is given, with N&P this is an empirical question.

### 3. Newell and Pizer's Approach

AR(3) model of the time series process:

$$r_t = \eta + e_t$$

$$e_t = \rho^3 e_{t-3} + \rho^2 e_{t-2} + \rho e_{t-1} + \xi_t$$

where  $\xi_t \sim N(0, \sigma_\xi^2)$ ,  $\eta \sim N(\bar{\eta}, \sigma_\eta^2)$

Leads to an expression for the certainty equivalent discount rate:

$$r_w(t) = \bar{\eta} - t\sigma_\eta^2 - \sigma_\xi^2 f(\rho, t)$$

- The theoretical model gives DDRs
- Estimated using US interest data. Numerical simulation taking 100000 forecasts over 400 years.
- Cannot distinguish clearly between random walk and mean reverting models
- RW (persistence) leads to a quicker decline in discount rates.

## **4. The Importance of Model Selection in inference and forecasting**

- N&P model characterises uncertainty in a very limited way: the data generating process is time invariant, cannot easily distinguish between MR and RW
- Such characteristics determine the nature of the schedule of DDRs and hence are important for CBA

### **Model Selection:**

- The selected model should be able to capture the dynamics of the data generating process in order to achieve an accurate description of the series.
- Complexity and restrictions of the model should correspond to the data generating process. If not, inference will be misleading and forecasting poor.
- Model selection should be based on data observation and statistical testing (e.g. unit roots, serial correlation, heteroscedasticity and parameter instability)
- An out of sample forecasting exercise can be a useful tool for measuring performance.

## 5. Method and Models

- Estimate N&P model: a simple AR(3) model
- Tests for persistence (random walk vs mean reversion) heteroscedasticity (LM test for ARCH), parameter stability.
- Apply appropriate further models as required:
  - i) AR(p) - (I)GARCH(l,m)
  - ii) Regime Switching
  - iii) State Space models
- Compare well specified models on the basis of a number of efficiency and forecasting criteria:
  - i) Coefficients of variation
  - ii) 5%-95% Quantiles of the forecast
  - iii) MSE's of sample forecasts
- Simulations undertaken: random draws of parameters and shocks from empirical distribution, then forecast.

## 6. Data

US nominal interest rates: 10 year bond constant maturity yield

UK nominal Interests Rates: Consol Yield

- Period: 1800-2001
- Adjusted using 10 year moving average inflation measure
- Accounted for the extreme fluctuations in the real rates seen during the 70's
- Continuously compounded equivalents

## 7. Results (U.S.) and Simulations

- N&P AR(3) model is misspecified in both the US and the UK cases:
  - Test for persistence inconclusive in US case (many tests): estimated both MR and RW models (MR in UK case)
  - Residuals are ARCH (LM test)
- The AR(p)- IGARCH(1,m) is employed in order to account for conditional heteroscedasticity.
  - Persistence leads to IGARCH estimates
  - High persistence and volatility indicates structural breaks
- State Space (SS) and Regime Switching (RS) models employed to assess this possibility
- These latter well specified models are compared on the basis of the efficiency criteria outlined above.
- This reveals a preference for the State Space models

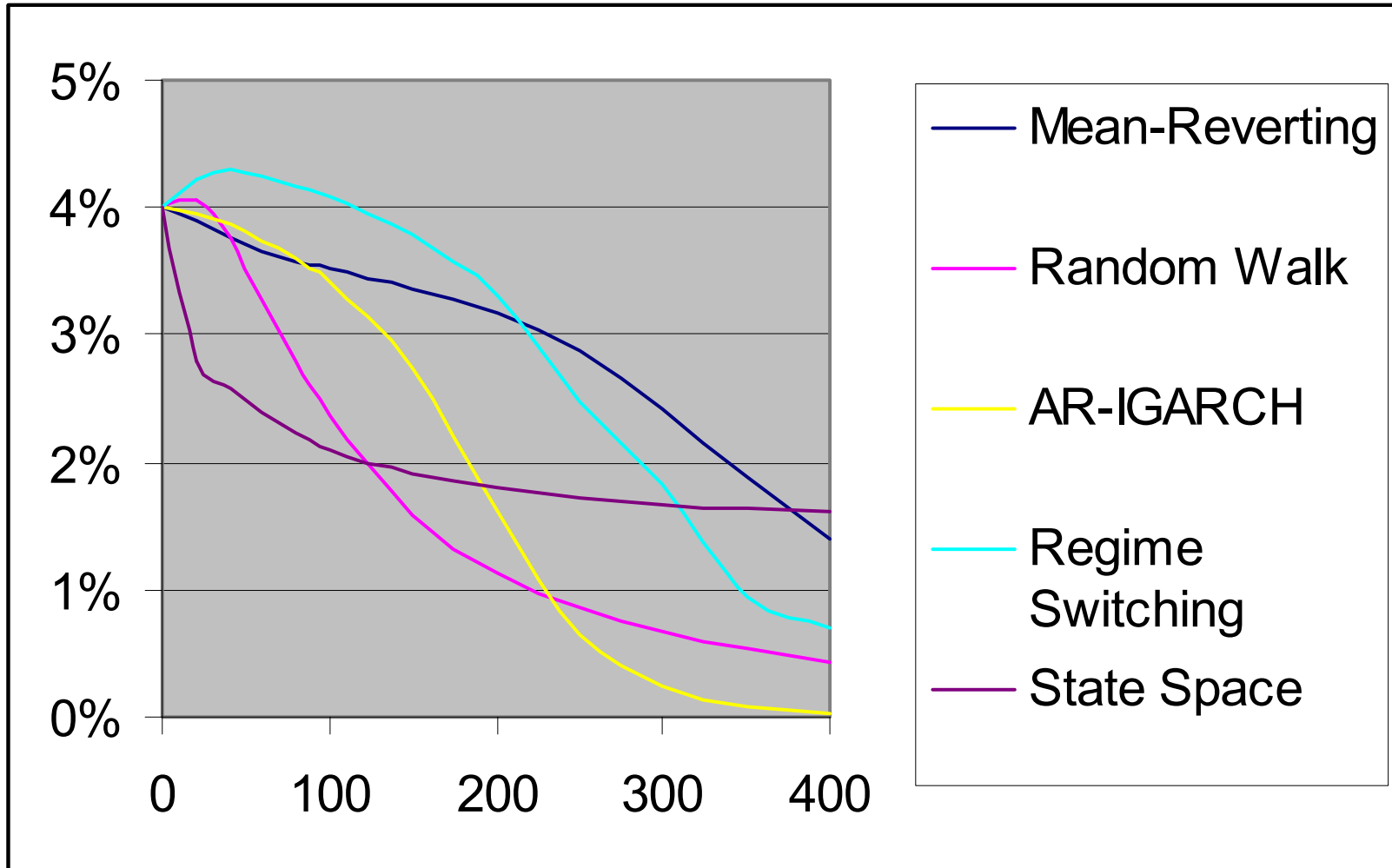
# Numerical Simulations of Certainty Equivalent Discount Factors for the US

| <b>Year</b> | <b>4% Constant</b> | <b>Mean-Reverting</b> | <b>Random Walk</b> | <b>AR-IGARCH</b> | <b>Regime-Switching</b> | <b>State Space</b> |
|-------------|--------------------|-----------------------|--------------------|------------------|-------------------------|--------------------|
| <b>1</b>    | 0.96154            | 0.96154               | 0.96154            | 0.96154          | 0.96154                 | 0.96154            |
| <b>20</b>   | 0.45639            | 0.45906               | 0.45177            | 0.45876          | 0.45390                 | 0.56424            |
| <b>40</b>   | 0.20829            | 0.21661               | 0.20917            | 0.21250          | 0.19576                 | 0.33136            |
| <b>60</b>   | 0.09506            | 0.10471               | 0.10480            | 0.10062          | 0.08458                 | 0.20296            |
| <b>80</b>   | 0.04338            | 0.05150               | 0.05777            | 0.04894          | 0.03700                 | 0.12889            |
| <b>100</b>  | 0.01980            | 0.02567               | 0.03482            | 0.02455          | 0.01647                 | 0.08408            |
| <b>150</b>  | 0.00279            | 0.00476               | 0.01333            | 0.00529          | 0.00238                 | 0.03132            |
| <b>200</b>  | 0.00039            | 0.00095               | 0.00683            | 0.00178          | 0.00041                 | 0.01255            |
| <b>250</b>  | 0.00006            | 0.00022               | 0.00419            | 0.00104          | 0.00010                 | 0.00526            |
| <b>300</b>  | 0.00001            | 0.00006               | 0.00289            | 0.00086          | 0.00003                 | 0.00227            |
| <b>350</b>  | 0.00000            | 0.00002               | 0.00215            | 0.00080          | 0.00002                 | 0.00100            |
| <b>400</b>  | 0.00000            | 0.00001               | 0.00169            | 0.00078          | 0.00001                 | 0.00044            |

## Simulated Certainty Equivalent Rates for the US

| <b>Year</b> | <b>Mean-Reverting</b> | <b>Random Walk</b> | <b>AR-IGARCH</b> | <b>Regime Switching</b> | <b>State Space</b> |
|-------------|-----------------------|--------------------|------------------|-------------------------|--------------------|
| <b>1</b>    | 4.00                  | 4.00               | 4.00             | 4.00                    | 4.00               |
| <b>20</b>   | 3.91                  | 4.05               | 3.96             | 4.22                    | 2.79               |
| <b>40</b>   | 3.76                  | 3.76               | 3.88             | 4.31                    | 2.59               |
| <b>60</b>   | 3.65                  | 3.28               | 3.74             | 4.26                    | 2.38               |
| <b>80</b>   | 3.58                  | 2.80               | 3.60             | 4.18                    | 2.23               |
| <b>100</b>  | 3.51                  | 2.37               | 3.42             | 4.09                    | 2.10               |
| <b>150</b>  | 3.36                  | 1.59               | 2.75             | 3.79                    | 1.91               |
| <b>200</b>  | 3.16                  | 1.14               | 1.62             | 3.31                    | 1.79               |
| <b>250</b>  | 2.87                  | 0.85               | 0.65             | 2.46                    | 1.72               |
| <b>300</b>  | 2.43                  | 0.66               | 0.23             | 1.83                    | 1.67               |
| <b>350</b>  | 1.87                  | 0.53               | 0.09             | 0.95                    | 1.64               |
| <b>400</b>  | 1.41                  | 0.44               | 0.04             | 0.70                    | 1.61               |

## Certainty Equivalent Discount Rates U.S.



## 8. Model Selection

- On the basis of the Coefficient of Variation and the proximity of 5% and 95% quantiles of the empirical distribution we cannot separate the SS and RS models.
- We evaluate the forecasting performance of models over a 30-year horizon using available real data (the annual nominal forward rates suggested by the term structure of the US government bonds).
- We calculate Mean Square Error (MSE) and compare four kernels, which weigh observations by their relevant proximity to the present.

### Average MSE's for the US

| <b>Criterion</b> | <b>Mean-Reverting</b> | <b>Random Walk</b> | <b>AR-IGARCH</b> | <b>Regime-Switching</b> | <b>State Space</b> |
|------------------|-----------------------|--------------------|------------------|-------------------------|--------------------|
| <b>AMSE</b>      | 2.058                 | 2.171              | 2.102            | 2.323                   | 1.832              |
| <b>AMSE (B)</b>  | 1.692                 | 1.724              | 1.692            | 1.687                   | 1.499              |
| <b>AMSE (P)</b>  | 1.725                 | 1.746              | 1.720            | 1.683                   | 1.426              |
| <b>AMSE (QS)</b> | 0.842                 | 0.870              | 0.848            | 0.879                   | 0.760              |
| <b>AMSE (TH)</b> | 1.769                 | 1.797              | 1.765            | 1.738                   | 1.550              |

## 9. Policy Implications of Model Selection

Value of Carbon emissions reduction: profile of benefits from Nordhaus DICE model

Value of Carbon (1989\$/tC, Base Year 1995)

| <b>Model</b>           | <b>Carbon Values (\$/tc 400years)</b> | <b>Relative to Constant Rate</b> | <b>Relative to Mean Reverting</b> | <b>Relative to Random Walk</b> |
|------------------------|---------------------------------------|----------------------------------|-----------------------------------|--------------------------------|
| <b>Regime Switch</b>   | 5.22                                  | -9.0%                            | -18.7%                            | -31.7%                         |
| <b>Constant (4.0%)</b> | 5.74                                  |                                  | -10.7%                            | -25.0%                         |
| <b>IGARCH</b>          | 6.37                                  | +10.9%                           | -1.0%                             | -16.8%                         |
| <b>N&amp;P (MR)</b>    | 6.43                                  | +12.0%                           |                                   | -16.0%                         |
| <b>N&amp;P (RW)</b>    | 7.65                                  | +33.3%                           | +19.0%                            |                                |
| <b>State Space</b>     | 14.44                                 | +151.7%                          | +124.7%                           | +88.8%                         |

## Implications for Nuclear Build in the U.K.

### NPV of Nuclear Build with Carbon Credits using CERs

| (£/kW)          | CAPEX  | OPEX   | Decom  | Rev's  | Carbon Credits 400 | NPV (Carbon Credits) |
|-----------------|--------|--------|--------|--------|--------------------|----------------------|
| Constant (3.5%) | 2172.8 | 2336.0 | 427.4  | 4062.4 | 227.9              | -645.9               |
| AR(4)           | 2167.0 | 2244.7 | 396.0  | 3903.7 | 215.4              | -688.6               |
| Regime Switch   | 2177.5 | 2401.0 | 479.4  | 4175.5 | 249.1              | -633.4               |
| State Space     | 2196.1 | 2973.1 | 1118.5 | 5170.4 | 547.3              | -570.0               |

## 10. Conclusion

- Misspecification testing in both the US and UK real interest rate series lead us away from simple AR(p) towards time varying models, i.e. the State Space and Regime-Switching models.
  
- As expected, the selection of the econometric model is of considerable moment in operationalising a theory of DDR's that depends upon uncertainty: econometric models contain different assumptions concerning the probability distribution and time series process of the object of interest.
  - i) SS gives lower CER over the first 200 years, RS give the lower rate for period longer than 200 years.
  
  - ii) RS and SS models give completely different measurements of the value of carbon reduction
  
- The out-of-sample forecast exercise reveals that the State Space model produces the more accurate forecasts and is preferred.



# Certainty Equivalent Discount Factor $E[e^{-\tilde{r}t}]$ and Rate: $r_w(t)$

