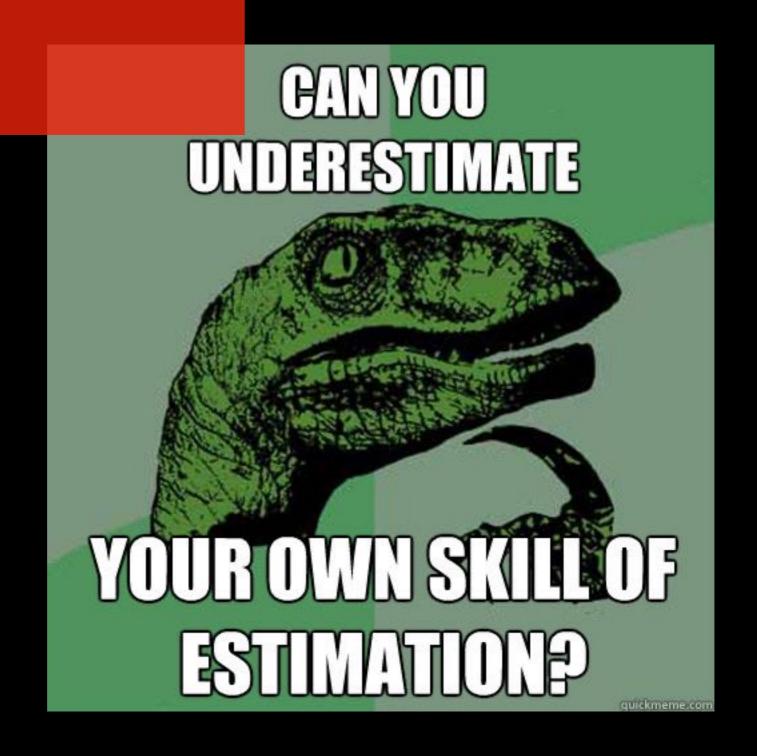
#### **Estimation**



## Learning outcomes

How to choose the **right data** for estimation Implications of **ergodicity** on estimation Use the **method of simulated moments** 

## Calibration vs estimation

Calibration: minimise the difference between R and M Estimation: find the true values of the variables

## Calibration vs estimation

Real world output

Calibration: minimise the difference between R and M Estimation: find the true values of the variables

**Model output** 

## Comparing apples with apples

How do we compare the **modelproduced data** with the real-world**produced data**?

#### Comparing apples with apples

How do we compare the modelproduced data with the real-worldproduced data?

Look for equilibria!

## The importance of ergodicity

Our model generates  $y_t$ , which becomes stationary with mean  $\mu^*=r\theta$  Say the model is non-ergodic, so  $\mu^*$  changes at every run, for the same  $\theta$ 

## The importance of ergodicity

Our model generates  $y_t$ , which

We need to understand the properties

of the model and of the data

before the estimation phase

changes at every run, for the same  $\theta$ 

# Preliminary tests

**Stationarity** is easy to test but things can be tricky if the data is not stationary

**Ergodicity** cannot be tested unless we observe many realisations of the same process

## Example

JVC - vhs

Sony - betamax





#### Conclusion

If we, somehow, get to be fairly positive about the ergodicity of the system, then we can use that for estimation.

Otherwise there are other methods to deal with estimation (just wait a few slides)

#### Simulated minimum distance

We want to minimise the distance between the summary statistics of the real-world system and those of our model

We can use the method of simulated moments

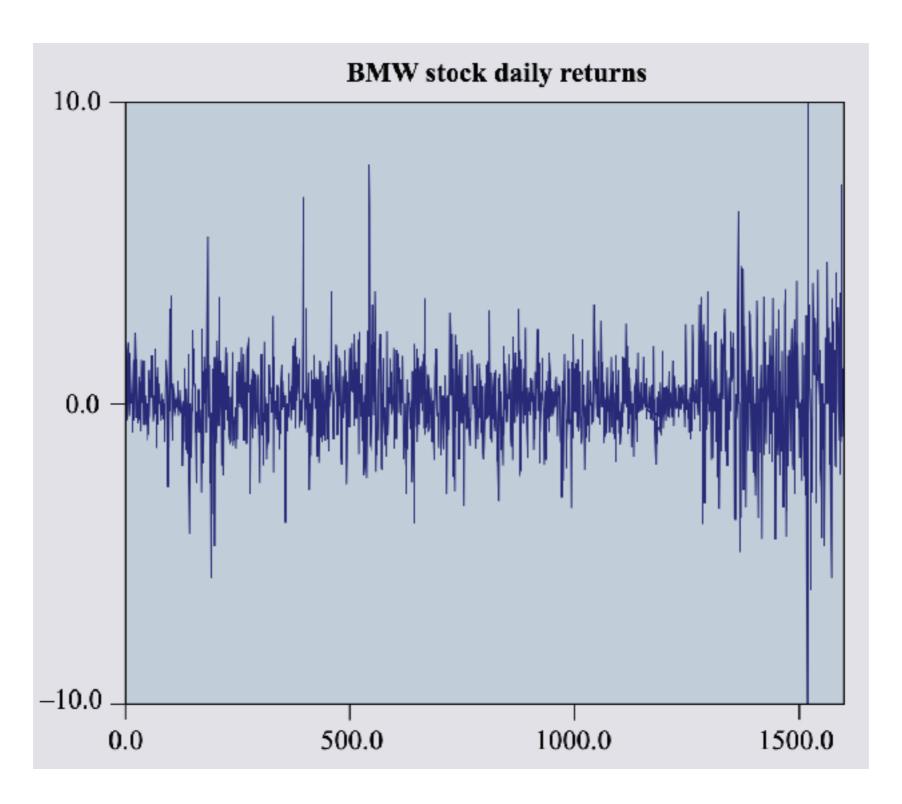
# Method of simulated moments

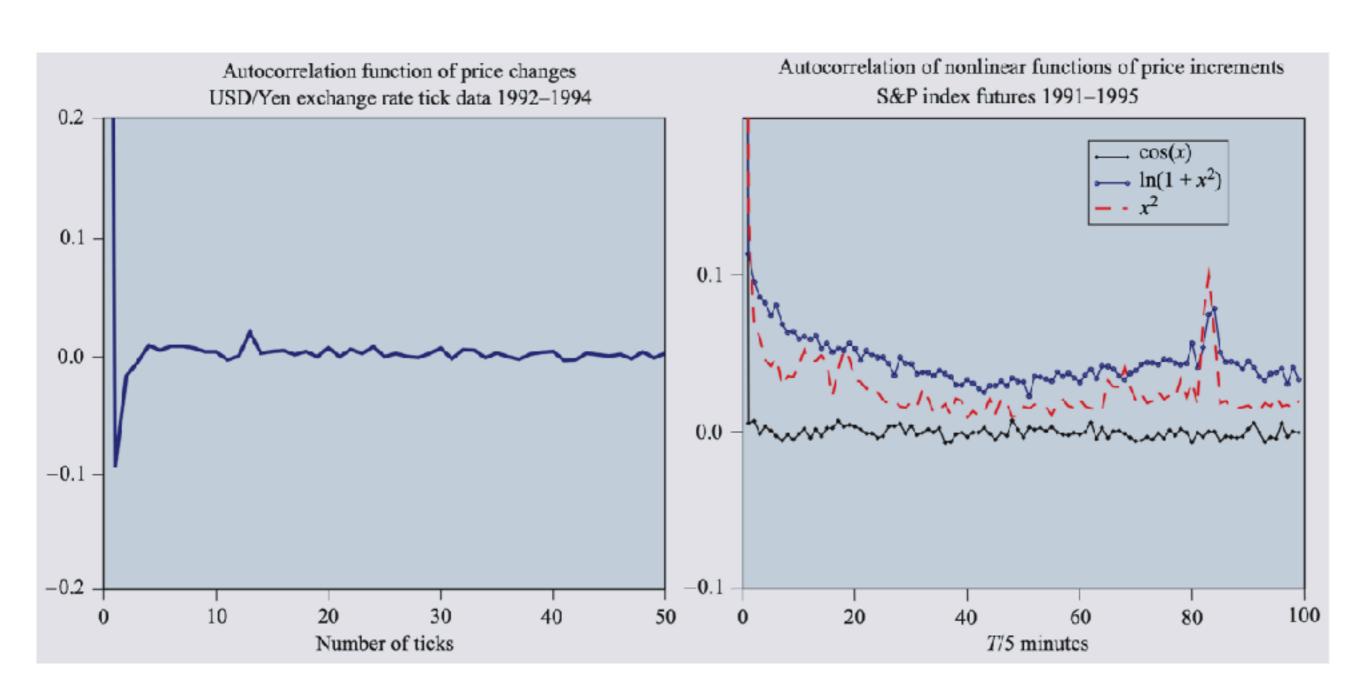
$$\hat{\theta} = argmin_{\theta}[\mu^*(\theta) - \mu_R]'W^{-1}[\mu^*(\theta) - \mu_R]$$

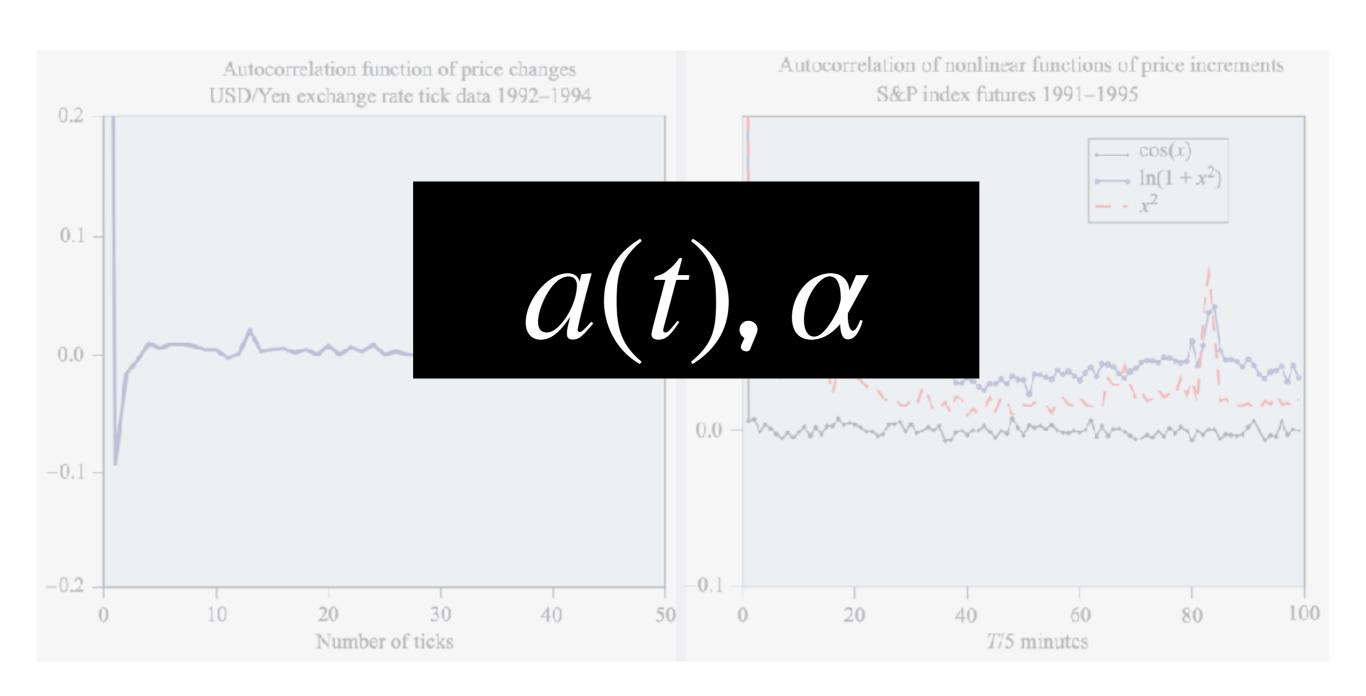
$$\uparrow$$

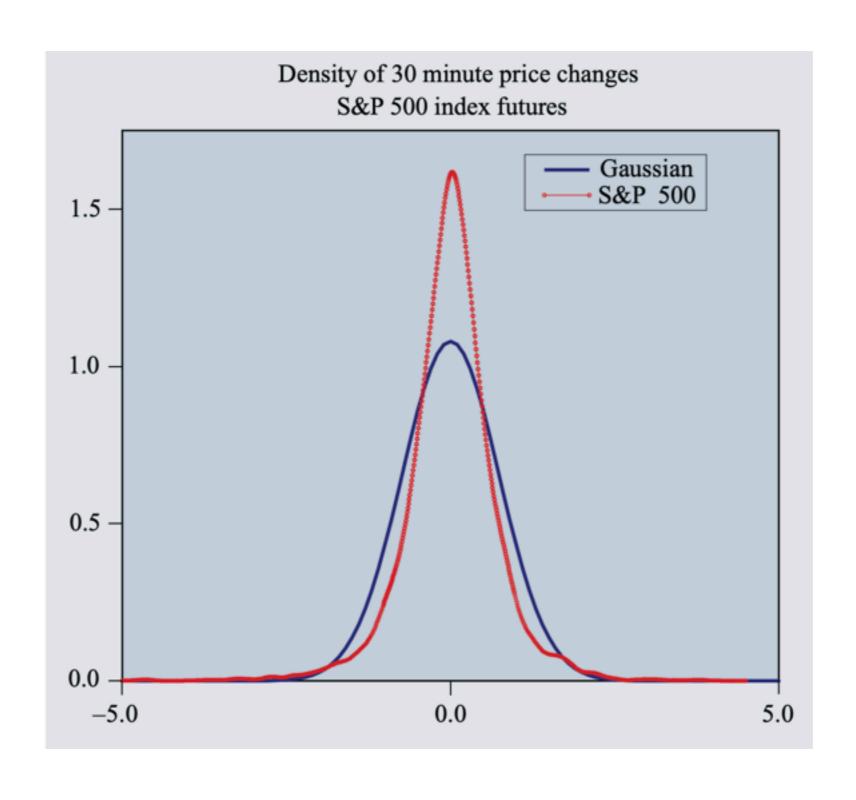
$$\downarrow$$

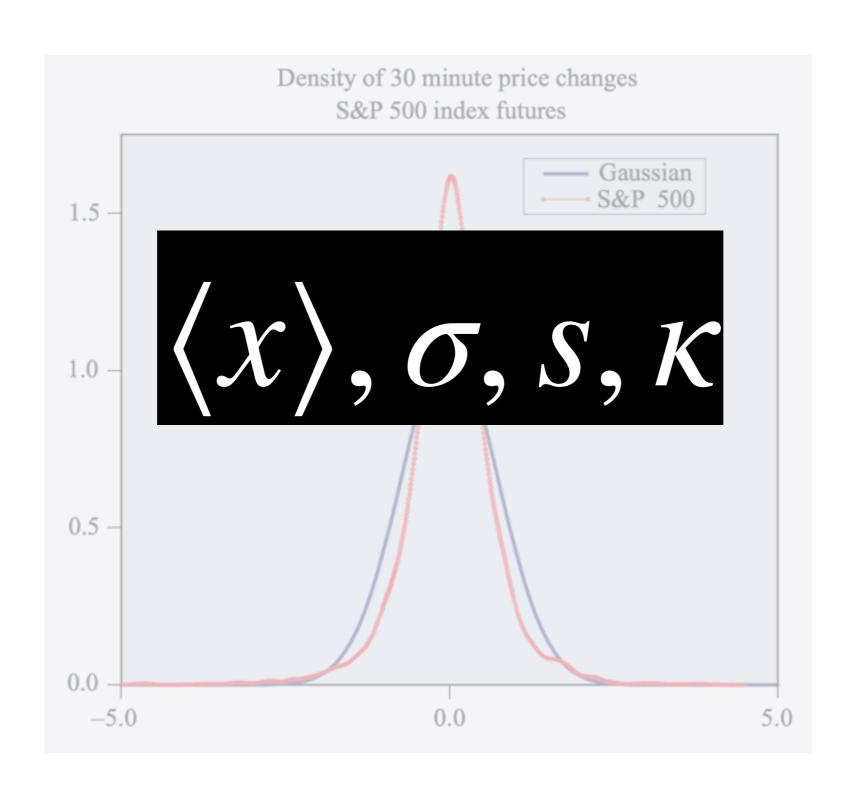
$$Weights$$











# Method of simulated moments

$$\hat{\theta} = argmin_{\theta}[w_1(\hat{\kappa} - \kappa) + w_2(\hat{a}(0) - a(0)) + w_3(\hat{\kappa} - \langle \kappa \rangle)]$$

# Method of simulated moments

$$\hat{\theta} = argmin_{\theta}[w_1(\hat{\kappa} - \kappa) + w_2(\hat{a}(0) - a(0)) + w_3(\hat{\kappa} - \langle \kappa \rangle)]$$

$$\hat{\theta} = argmin_{\theta}[w_1(\hat{\kappa} - \kappa) + w_2(\hat{\alpha} - \alpha) + w_3(\hat{x} - \langle x \rangle)]$$

## Bayesian estimation

#### More appropriate when:

- inference is needed
- there are prior and posterior distributions

#### Tesco example - Recap

Two types of customer agents with heterogeneous variables\*

One type of worker agent

Variables: time spent in store, expenditure, bias on deals, etc.
Actions: buy, recommend (last year people also suggested "steal")
Interactions: recommend products to others, interact with store



What parameters do you want to calibrate/estimate?

#### Tesco example - Recap

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What summary statistics would you use for estimation/calibration?

#### How to build abms cheat sheet

Nature of agents
List of variables describing their state
List of actions the agents can perform
Structure of their interaction with other agents

Agent design

Definition of output variables of interest Appropriate experimental design Analysis of equilibria Sensitivity analysis



#### How to build abms cheat sheet

Select the appropriate data Input validation
Output validation

Select the appropriate data

Test for stationarity and ergodicity

Methods of simulated distances



Estimation

# Congratulations! Now you know how to build an agent-based model

Fin