

Model risk and Bayesian decisions for financial processes

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Outline

- 1 Introduction
- 2 Density, Hellinger integrals and I_α -divergences
- 3 Bayesian decision procedures and Bayes risk
- 4 Some results of convergence

Asset price dynamics in "path space model":

- observation time $T = n$
- time set $I = \{0, 1, 2, \dots, n\}$ (in units)
- capital growth factor $Y_i = u_i$ or d_i , with $0 < d_i < e^r < u_i$
- asset price at time $t \in I$:

$$X_0 := x, \quad X_t = x \cdot \prod_{i=1}^t Y_i$$

- random paths $\omega = (X_t : t \in I)$
- canonical space of all possible sample paths Ω

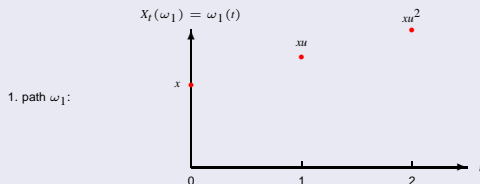
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2-Period-Binomial-Model:



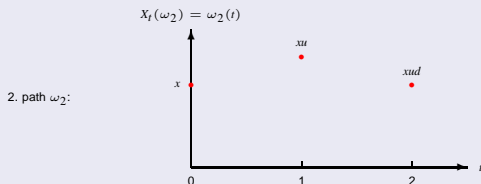
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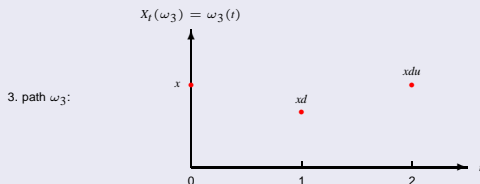
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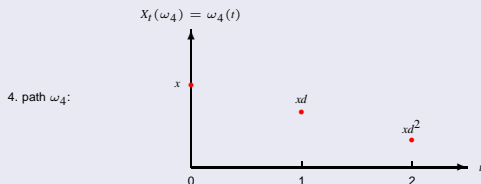
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2-Period-Binomial-Model:



Probabilistic dynamics:

- two possible models:

$$(\mathcal{H}) \quad Q^{\mathcal{H}}[Y_i = u_i] = q_i^{\mathcal{H}} = 1 - Q^{\mathcal{H}}[Y_i = d_i] \quad \text{with} \quad q_i^{\mathcal{H}} \in (0, 1)$$

$$(\mathcal{A}) \quad Q^{\mathcal{A}}[Y_i = u_i] = q_i^{\mathcal{A}} = 1 - Q^{\mathcal{A}}[Y_i = d_i] \quad \text{with} \quad q_i^{\mathcal{A}} \in (0, 1)$$

- assumption: $q_i^{\mathcal{H}} \neq q_i^{\mathcal{A}}$ for at least one $i \in \{1, \dots, n\}$

Probabilistic dynamics:

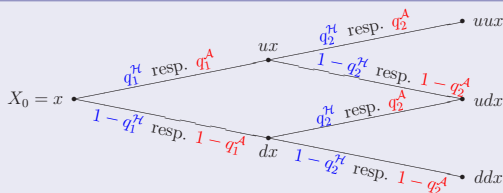
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Bayes decisions



$$\boxed{\begin{matrix} Q^{\mathcal{H}} \\ q_i^{\mathcal{H}} \end{matrix}}$$

vs.

$$\boxed{\begin{matrix} Q^{\mathcal{A}} \\ q_i^{\mathcal{A}} \end{matrix}}$$

$$\Pr[\mathcal{H}] = pr_{\mathcal{H}}^{\text{prior}} = 0.5$$

$$\Pr[\mathcal{A}] = pr_{\mathcal{A}}^{\text{prior}} = 0.5$$

Example 1

- total amount of money $m = 10000$ Euro
- $T = 1$ unit = $\frac{1}{4}$ year
- savings deposit** with annual interest rate $r = 0.19\%$
- stock** with $\sigma = 0.1$ per year, $u = e^{\sigma\sqrt{\frac{T}{1}}} \approx 1.05$, $d = \frac{1}{u} \approx 0.95$

$$\mathcal{H}: q^{\mathcal{H}} = \frac{1}{2} + \frac{1}{2} \frac{1}{\sigma} \left(c_{\mathcal{H}} - \frac{\sigma^2}{2} \right) \sqrt{\frac{T}{1}} \approx \underline{0.52} \text{ with } c_{\mathcal{H}} = 0.013$$

$c_{\mathcal{H}} \hat{=}$ "average growth rate" (per year) ... **optimistic**: 1.3%

$$\mathcal{A}: q^{\mathcal{A}} = \frac{1}{2} + \frac{1}{2} \frac{1}{\sigma} \left(c_{\mathcal{A}} - \frac{\sigma^2}{2} \right) \sqrt{\frac{T}{1}} \approx \underline{0.49} \text{ with } c_{\mathcal{A}} = 0.001$$

$c_{\mathcal{A}} \hat{=}$ "average growth rate" (per year) ... **pessimistic**: 0.1%

We would either like to invest **all** the money in the stock (take the decision $d_{\mathcal{H}}$) or **all** the money in the savings deposit (take the decision $d_{\mathcal{A}}$) for the next period $(\frac{1}{4}, \frac{1}{2}]$.

Example 1

decision losses (rely on **expected wealth for $t = 2$ unit = $\frac{1}{2}$ year**):

- savings deposit: $m \cdot e^{0.0019 \cdot (\frac{1}{2} - \frac{1}{4})} = 10004.75$ Euro
- stock under model \mathcal{H} :

$$\begin{aligned} EQ^{\mathcal{H}} \left[\frac{m}{X_1} \cdot X_2 \right] &= m \left\{ q^{\mathcal{H}} \cdot u + (1 - q^{\mathcal{H}}) \cdot d \right\} \\ &= 10000 \cdot \{ \underline{0.52} \cdot 1.05 + (1 - \underline{0.52}) \cdot 0.95 \} = 10020 \text{ Euro} \end{aligned}$$

- stock under model \mathcal{A} :

$$\begin{aligned} EQ^{\mathcal{A}} \left[\frac{m}{X_1} \cdot X_2 \right] &= m \left\{ q^{\mathcal{A}} \cdot u + (1 - q^{\mathcal{A}}) \cdot d \right\} \\ &= 10000 \cdot \{ \underline{0.49} \cdot 1.05 + (1 - \underline{0.49}) \cdot 0.95 \} = 9990 \text{ Euro} \end{aligned}$$

- $d_{\mathcal{H}} \hat{=}$ invest all the money in the stock

$$\tilde{L}_{\mathcal{H}} := L(d_{\mathcal{H}}, \mathcal{H}) = 0 \text{ (relatively seen)}$$

$$L_{\mathcal{A}} := L(d_{\mathcal{H}}, \mathcal{A}) = 10004.75 - 9990 = 14.75 \text{ Euro}$$

- $d_{\mathcal{A}} \hat{=}$ invest all the money in the savings deposit

$$\tilde{L}_{\mathcal{A}} := L(d_{\mathcal{A}}, \mathcal{A}) = 0 \text{ (relatively seen)}$$

$$L_{\mathcal{H}} := L(d_{\mathcal{A}}, \mathcal{H}) = 10020 - 10004.75 = 15.25 \text{ Euro}$$

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density of Q^A with respect to $Q^{\mathcal{H}}$

$$Z_T(\omega) = \begin{cases} \frac{q_1^A \dots q_n^A}{q_1^{\mathcal{H}} \dots q_n^{\mathcal{H}}}, & \text{if } X_T(\omega) = x u_1 \dots u_n, \\ \frac{q_1^A \dots q_{n-1}^A (1 - q_n^A)}{q_1^{\mathcal{H}} \dots q_{n-1}^{\mathcal{H}} (1 - q_n^{\mathcal{H}})}, & \text{if } X_T(\omega) = x u_1 \dots u_{n-1} d_n, \\ \frac{q_1^A \dots q_{n-2}^A (1 - q_{n-1}^A) q_n^A}{q_1^{\mathcal{H}} \dots q_{n-2}^{\mathcal{H}} (1 - q_{n-1}^{\mathcal{H}}) q_n^{\mathcal{H}}}, & \text{if } X_T(\omega) = x u_1 \dots u_{n-2} d_{n-1} u_n, \\ \frac{q_1^A \dots q_{n-2}^A (1 - q_{n-1}^A) (1 - q_n^A)}{q_1^{\mathcal{H}} \dots q_{n-2}^{\mathcal{H}} (1 - q_{n-1}^{\mathcal{H}}) (1 - q_n^{\mathcal{H}})}, & \text{if } X_T(\omega) = x u_1 \dots u_{n-2} d_{n-1} d_n, \\ \vdots & \vdots \\ \vdots & \vdots \\ \frac{(1 - q_1^A) \dots (1 - q_n^A)}{(1 - q_1^{\mathcal{H}}) \dots (1 - q_n^{\mathcal{H}})}, & \text{if } X_T(\omega) = x d_1 \dots d_n \end{cases}$$

Remark

If $q_i^{\mathcal{H}} \equiv q^{\mathcal{H}}$ and $q_i^A \equiv q^A$, then Z_T depends only on X_T (and not on the path to X_T).

Hellinger integral

For $\alpha \in \mathbb{R}$:

$$\begin{aligned}
 H_\alpha(Q^A \| Q^{\mathcal{H}}) &= \int \{g^A\}^\alpha \{g^{\mathcal{H}}\}^{1-\alpha} d\mu = EQ^{\mathcal{H}}[(Z_T)^\alpha] \\
 &= (q_1^A \dots q_n^A)^\alpha \cdot (q_1^{\mathcal{H}} \dots q_n^{\mathcal{H}})^{1-\alpha} \\
 &\quad + (q_1^A \dots q_{n-1}^A (1 - q_n^A))^\alpha \cdot (q_1^{\mathcal{H}} \dots q_{n-1}^{\mathcal{H}} (1 - q_n^{\mathcal{H}}))^{1-\alpha} \\
 &\quad + (q_1^A \dots q_{n-2}^A (1 - q_{n-1}^A) q_n^A)^\alpha \cdot (q_1^{\mathcal{H}} \dots q_{n-2}^{\mathcal{H}} (1 - q_{n-1}^{\mathcal{H}}) q_n^{\mathcal{H}})^{1-\alpha} \\
 &\quad + (q_1^A \dots q_{n-2}^A (1 - q_{n-1}^A) (1 - q_n^A))^\alpha \cdot (q_1^{\mathcal{H}} \dots q_{n-2}^{\mathcal{H}} (1 - q_{n-1}^{\mathcal{H}}) (1 - q_n^{\mathcal{H}}))^{1-\alpha} \\
 &\quad \quad \quad \vdots \quad \quad \quad \vdots \\
 &\quad + ((1 - q_1^A) \dots (1 - q_n^A))^\alpha \cdot ((1 - q_1^{\mathcal{H}}) \dots (1 - q_n^{\mathcal{H}}))^{1-\alpha},
 \end{aligned}$$

where $g^A = \frac{dQ^A}{d\mu} \Big|_I$ and $g^{\mathcal{H}} = \frac{dQ^{\mathcal{H}}}{d\mu} \Big|_I$ are the densities with respect to the specially chosen reference law $\mu = Q^{\mathcal{H}}$ (n time points).

I_α -divergence

For $\alpha \in \mathbb{R}$:

$$I_\alpha(Q^A \| Q^{\mathcal{H}}) = \int f_\alpha \left(\frac{dQ^A}{dQ^{\mathcal{H}}} \Big|_I \right) dQ^{\mathcal{H}},$$

with the nonnegative functions $f_\alpha : [0, \infty) \rightarrow [0, \infty)$ defined by

$$f_\alpha(\rho) = \begin{cases} -\log \rho + \rho - 1, & \text{if } \alpha = 0, \\ \frac{\alpha\rho + 1 - \alpha - \rho^\alpha}{\alpha(1-\alpha)}, & \text{if } \alpha \in \mathbb{R} \setminus \{0, 1\}, \\ \rho \log \rho + 1 - \rho, & \text{if } \alpha = 1. \end{cases}$$

For $\alpha \in \mathbb{R} \setminus \{0, 1\}$ that holds

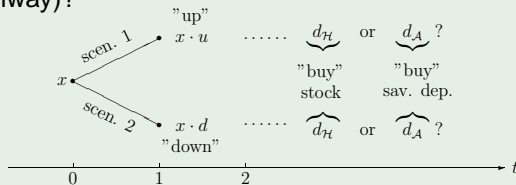
$$I_\alpha(Q^A \| Q^{\mathcal{H}}) = \frac{1 - H_\alpha(Q^A \| Q^{\mathcal{H}})}{\alpha(1 - \alpha)}.$$

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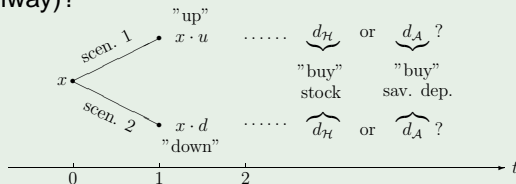
Example 1 (continuing)

How to plan (now) an optimal decision way/decision rule (in an Bayesian way)?



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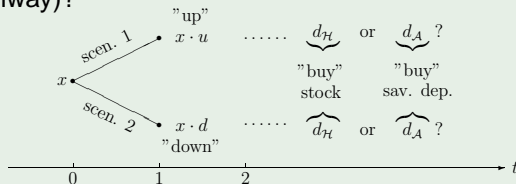
Mean loss

- decision rule $\delta : \Omega \longrightarrow \{d_{\mathcal{H}}, d_{\mathcal{A}}\}$, $\delta(\omega) = \begin{cases} d_{\mathcal{A}}, & \text{if } \omega \in G, \\ d_{\mathcal{H}}, & \text{if } \omega \notin G, \end{cases}$
- mean loss:

$$\begin{aligned}
 \mathcal{L}(\delta) &= \mathcal{L}_x(\delta_G) \\
 &= L_{\mathcal{H}} \Pr[\delta_G(\omega) = d_{\mathcal{A}}, \mathcal{H}] + L_{\mathcal{A}} \Pr[\delta_G(\omega) = d_{\mathcal{H}}, \mathcal{A}] \\
 &= L_{\mathcal{H}} \Pr[\omega \in G | \mathcal{H}] \cdot \Pr[\mathcal{H}] + L_{\mathcal{A}} \Pr[\omega \in \Omega \setminus G | \mathcal{A}] \cdot \Pr[\mathcal{A}] \\
 &= L_{\mathcal{H}} pr_{\mathcal{H}}^{\text{prior}} Q^{\mathcal{H}}[G] + L_{\mathcal{A}} pr_{\mathcal{A}}^{\text{prior}} Q^{\mathcal{A}}[\Omega \setminus G]
 \end{aligned}$$

Example 1 (continuing)

How to plan (now) an optimal decision way/decision rule (in an Bayesian way)?



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 &= L_{\mathcal{H}} pr_{\mathcal{H}}^{\text{prior}} Q^{\mathcal{H}}[G] + L_{\mathcal{A}} pr_{\mathcal{A}}^{\text{prior}} Q^{\mathcal{A}}[\Omega \setminus G] \quad \longrightarrow \text{min!}
 \end{aligned}$$

1st decision method

- optimal decision rule:

$$\delta_{\text{opt}}(\omega) = \delta_{G_{\min}} = \begin{cases} d_{\mathcal{A}}, & \text{if } \omega \in G_{\min}, \\ d_{\mathcal{H}}, & \text{if } \omega \notin G_{\min}, \end{cases}$$

with "optimal decision set" $G_{\min} := \left\{ \text{paths } \omega \in \Omega : Z_T(\omega) \geq \frac{L_{\mathcal{H}} pr_{\mathcal{H}}^{\text{prior}}}{L_{\mathcal{A}} pr_{\mathcal{A}}^{\text{prior}}} \right\}$

- Bayes risk (minimal mean loss):

$$\mathcal{L}(\delta_{\text{opt}}) = \mathcal{L}(\delta_{G_{\min}}) = \dots = EQ^{\mathcal{H}} \left[\min \{ L_{\mathcal{H}} pr_{\mathcal{H}}^{\text{prior}}, L_{\mathcal{A}} pr_{\mathcal{A}}^{\text{prior}} Z_T \} \right]$$

1st decision method

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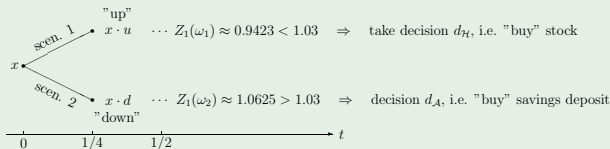
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Example 1 (continuing)

$$G_{\min} = \left\{ \omega \in \Omega : Z_1 \geq \frac{0.5 \cdot 15.25}{0.5 \cdot 14.75} \approx 1.03 \right\}$$



$$\mathcal{L}(\delta_{G_{\min}}) = 0.4850 \text{ Euro} \quad (\text{of total investment of 10000 Euro})$$

2nd decision method (\Leftrightarrow 1st method)

- decision criterion:

If $pr_{\mathcal{H}}^{post} L_{\mathcal{H}} \leq pr_{\mathcal{A}}^{post} L_{\mathcal{A}}$, then decide for $d_{\mathcal{A}}$,

If $pr_{\mathcal{H}}^{post} L_{\mathcal{H}} > pr_{\mathcal{A}}^{post} L_{\mathcal{A}}$, then decide for $d_{\mathcal{H}}$,

$$\text{where } pr_{\mathcal{H}}^{post} = \frac{pr_{\mathcal{H}}^{prior}}{pr_{\mathcal{H}}^{prior} + (1 - pr_{\mathcal{H}}^{prior}) Z_T}, \quad pr_{\mathcal{A}}^{post} = \frac{(1 - pr_{\mathcal{H}}^{prior}) Z_T}{pr_{\mathcal{H}}^{prior} + (1 - pr_{\mathcal{H}}^{prior}) Z_T}$$

are the "overall evidence" probabilities posterior to our observation between the times 0 and T .

- Bayes risk as with 1st decision method
- Bayes factor: $B_T = \frac{\text{posterior odds ratio of } \mathcal{A} \text{ to } \mathcal{H}}{\text{prior odds ratio of } \mathcal{A} \text{ to } \mathcal{H}} = Z_T$

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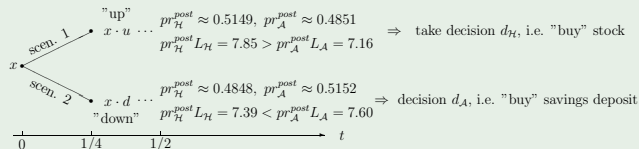
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Example 1 (continuing)



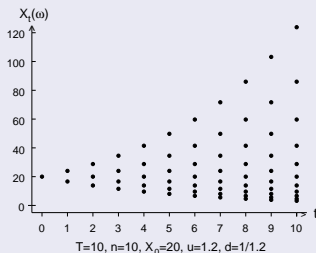
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Special homogeneous case (SPH) with n periods



observations at times
 $0, \frac{T}{n}, 2\frac{T}{n}, \dots, n\frac{T}{n} = T$, with
 fixed **approximation**
 step n

in **each knot** like in the one-period-model, independently of past/history, with

$$u_i = u_{(n)} := e^{\sigma\sqrt{\frac{T}{n}}}, \quad d_i = d_{(n)} := e^{-\sigma\sqrt{\frac{T}{n}}},$$

$$\text{for } \mathcal{H}: q_i^{\mathcal{H}} = q_{(n)}^{\mathcal{H}} := \frac{1}{2} + \frac{1}{2} \frac{1}{\sigma} \left(c_{\mathcal{H}} - \frac{\sigma^2}{2} \right) \sqrt{\frac{T}{n}} \dots \dots Q^{\mathcal{H}}$$

$$\text{for } \mathcal{A}: q_i^{\mathcal{A}} = q_{(n)}^{\mathcal{A}} := \frac{1}{2} + \frac{1}{2} \frac{1}{\sigma} \left(c_{\mathcal{A}} - \frac{\sigma^2}{2} \right) \sqrt{\frac{T}{n}} \dots \dots Q^{\mathcal{A}}$$

with volatility $\sigma > 0$, and "average growth rate" $c_{\mathcal{H}} \in \mathbb{R}$ resp. $c_{\mathcal{A}} \in \mathbb{R}$

homogeneous binomial models $\begin{matrix} q_{(n)}^{\mathcal{H}} \\ q_{(n)}^{\mathcal{A}} \end{matrix} \xrightarrow{n \rightarrow \infty} \text{geom. Brown. motions} \begin{matrix} (c_{\mathcal{H}}, \sigma^2) \\ (c_{\mathcal{A}}, \sigma^2) \end{matrix}$

Some Results of Convergence

- Hellinger integral:

$$H_{\alpha}^{(n)}(Q^{\mathcal{A}} \| Q^{\mathcal{H}}) \xrightarrow{n \rightarrow \infty} \exp \left\{ \left(\frac{c_{\mathcal{A}} - c_{\mathcal{H}}}{\sigma} \right)^2 T \frac{\alpha(\alpha - 1)}{2} \right\} =: C_{\alpha}$$

- I_{α} -divergence:

- $\alpha \in \mathbb{R} \setminus \{0, 1\}$: $I_{\alpha}^{(n)}(Q^{\mathcal{A}} \| Q^{\mathcal{H}}) \xrightarrow{n \rightarrow \infty} \frac{1 - C_{\alpha}}{\alpha(1 - \alpha)}$

- $\alpha = 1$: $I_1^{(n)}(Q^{\mathcal{A}} \| Q^{\mathcal{H}}) \xrightarrow{n \rightarrow \infty} \frac{1}{2} \left(\frac{c_{\mathcal{A}} - c_{\mathcal{H}}}{\sigma} \right)^2 T$ (relative entropy)

- $\alpha = 0$: $I_0^{(n)}(Q^{\mathcal{A}} \| Q^{\mathcal{H}}) \xrightarrow{n \rightarrow \infty} \frac{1}{2} \left(\frac{c_{\mathcal{A}} - c_{\mathcal{H}}}{\sigma} \right)^2 T$

- minimal mean loss:

$$\mathcal{L}(\delta_{G_{\min}}^{T,n}) \xrightarrow{n \rightarrow \infty} \lambda_{\mathcal{A}} \cdot (1 - \Phi_{\mathcal{N}(0,1)}(\theta_1)) + \lambda_{\mathcal{H}} \cdot \Phi_{\mathcal{N}(0,1)}(\theta_2)$$

with: $\theta_1 := \frac{\Delta \sqrt{T}}{2} - \frac{\ln(\lambda_{\mathcal{H}}/\lambda_{\mathcal{A}})}{\Delta \sqrt{T}}$, $\theta_2 := -\frac{\Delta \sqrt{T}}{2} - \frac{\ln(\lambda_{\mathcal{H}}/\lambda_{\mathcal{A}})}{\Delta \sqrt{T}}$, and

$$\Delta := \frac{\mu_{\mathcal{H}} - \mu_{\mathcal{A}}}{\sigma}, \lambda_{\mathcal{H}} := pr_{\mathcal{H}}^{\text{prior}} L_{\mathcal{H}}, \lambda_{\mathcal{A}} := pr_{\mathcal{A}}^{\text{prior}} L_{\mathcal{A}}$$

(conform with Stummer/Vajda 2005)

- further investigations (e.g. more general losses.....)

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