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**Bayesian Nash Equilibrium for Insider
Trading in Continuous Time**

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Bayesian Nash Equilibrium for Insider Trading in Continuous Time *

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*This paper is greatly in debt with the unpublished paper [8]. Actually, we consider a similar model and use the same methodology. The main difference is that the drift of the dividend process is not given by a stochastic process but it is a constant. In [8], the authors were not able to characterize the equilibrium explicitly. They addressed the market maker's filtering problem, and the insider's optimal consumption-investment problem and they suggested that numerical methods allow to characterize the equilibrium.

Abstract

We consider the Kyle model [12] and we provide a no trade theorem with no transmission of information in a market with an insider trader, noise traders, and a market maker. The main difference between our setting and Kyle's one is that the insider trader is risk averse and that the risky asset pays a continuous time dividend stream, which is described by a Brownian motion with drift. The dividend realization is observed by all agents, but the insider trader knows the dividend growth rate while the market maker does not. We define a Bayesian-Nash equilibrium by addressing the market maker's optimal filtering problem, and the insider trader's optimal consumption-trading strategy. In equilibrium the insider trader does not trade and the market maker does not update his beliefs. Therefore, the equilibrium does not transmit information.

1 Introduction

A well known result on financial markets with privately informed agents establishes that agents do not trade on the basis of private information, because either the others do not want to trade, or are able to identify their private information by observing market prices.

Two pieces of theory make the point. Provided an ex ante Pareto optimal allocation is reached, the no trade Theorem states that no trade occurs after information becomes heterogeneous among agents, because there is no recontracting leading to a Pareto improvement (see [15]). On the other hand, provided that all agents know the features of the economy and trade in order to maximize their utility, under some general assumptions, it can be shown that the rational expectations equilibrium with heterogeneous-private information coincides with the equilibrium of the artificial economy, i.e., the economy where all pieces of private information are of common knowledge (see for example [9]).

To overcome the no trade theorem and to render non fully revealing the rational expectations equilibrium, two main perspectives have been pursued in the literature: the introduction of noise-liquidity traders in the market, i.e., agents who do not trade according to the rationality paradigm, but only for liquidity reasons; the analysis of non perfectly competitive markets. The Kyle model represents an illuminating example (see [12]). In a two-period economy a risky asset is traded at $t = 0$ and there are two classes of agents: a risk neutral insider trader and a set of noise traders. The risky asset demand by noise traders is described by a random variable which is independent of the asset dividend. The latter is a random variable whose realization is observed by all agents at $t = 1$, while the insider trader knows its value since $t = 0$, before trading. The price at $t = 0$ is determined by the market maker as the conditional expected value of the dividend, given the market order flow (the cumulative insider and noise traders' demand). In this setting, it is shown that there exists an equilibrium in which the insider trades. The higher the variance of the noise traders' demand is, the higher the sensitivity of the insider trader's demand to the dividend becomes. The rationale of this result is very simple: if the noise traders' demand can be very large (high variance), then the insider trader can trade in the market profitably, by knowing that his private information is more easily hidden to the eyes of the market maker, and therefore not reflected in the market price. Also the fully revealing property breaks down: after trading, noise traders' precision is the half of the insider trader's one. The main properties of the model are confirmed in a multiperiod setting, assuming that the insider trades

in many intermediate time instants by knowing the dividend delivered at $t = 1$. The insider trader acts gradually in order to strategically exploit his information and market prices do not reflect the insider's private information immediately. When there are multiple insider traders (each insider knows the liquidation value of the asset before starting to trade) we observe that competition among insiders pushes them to trade aggressively and causes most of their private information to be revealed very rapidly (see [10]).

In this paper we consider the Kyle model and we provide a no trade theorem with no transmission of information in a market with an insider, noise traders, and a market maker. The result represents a novelty in the Kyle setting. The main point that differentiates our setting from the Kyle one is that the insider trader is risk averse and the risky asset pays a continuous time dividend stream, which is described by a Brownian motion with drift, an assumption similar to that of [4, 17, 8]. The dividend realization is observed by all agents, but the insider trader knows the dividend growth rate while the market maker does not. The market maker is risk neutral, he updates his beliefs on the dividend growth rate continuously and sets the price of the risky asset as the conditional expectation of the discounted future dividend stream. As in [12, 2], the market maker observes directly only the aggregate order flow, which is a noisy indicator, owing to the presence of noise traders, and he cannot distinguish the insider trader's component. His attempt of detecting the insider trader's demand, and therefore the dividend growth rate, gives rise to a filtering problem: market maker's learning about the insider trader's information from the dividend time series and the aggregate order flow. On the other hand, the insider trader defines his trading strategy by maximizing the expected utility taking into account that the price is set by the market maker.

We define a Bayesian-Nash equilibrium by addressing the market maker's optimal filtering problem and the insider trader's optimal consumption-trading strategy. It is a Bayesian-Nash equilibrium, in a sense that the market maker postulates the insider trader's strategy when he updates his beliefs, and the insider trader conjectures the market maker's updating rule when he addresses his optimal consumption-trading problem. In equilibrium, we impose that conjectures are self-confirming. We restrict our attention to stationary linear equilibria, i.e., both the the market maker's updating rule and the insider trader's trading strategy are linear in the state variables with constant coefficients.

The main result of the paper is that there exists a unique Bayesian-Nash equilibrium which is a No Trade Theorem: in equilibrium the insider trader does not trade and the market maker does not update his beliefs. So the

equilibrium does not transmit information. Noise traders are not enough: if the insider trades in the market, then the market maker immediately recovers the drift of the dividend process and sets the price as the expected value of the discounted dividend stream, thereby rendering insider's trading non profitable.

The paper is organized as follows. In Section 2 we present the model. Section 3 addresses the optimal filtering problem of the market maker. In Section 4 we address the optimal investment-consumption problem of the insider trader. In Section 5 we close the model and we characterize the Bayesian-Nash equilibrium. All proofs of the established propositions are given in the Appendix.

2 The Model

The model considered in this paper is that in [8], the main difference is that the drift of the dividend process is constant and it is not stochastic. There are a risk free and a risky asset, the risk free asset pays an instantaneous interest rate $r > 0$, while the risky asset, with price $P(t)$, yields a dividend stream $D(t)$. We assume that the history of $D(t)$, up to the current instant t , is observed by all agents, that is to say the filtration $(\mathfrak{F}_s^D)_{s \leq t}$ generated by the dividend process is part of the publicly available information. In addition, we assume that the dynamics of $D(t)$ is driven by the stochastic differential equation

$$dD(t) = \pi dt + \sigma_D dw_D(t), \quad (1)$$

where π and σ_D are constant parameters, with $\sigma_D > 0$, and $w_D(t)$ is a Wiener process.

There are three types of agents: a representative noise trader, an insider trader, and a market maker.

The representative noise trader trades smoothly. Namely, the value $\Theta(t)$ of the representative noise trader's order flow is the "derivative" of his inventory. As a consequence $\Theta(t)$ follows an Ornstein-Uhlenbeck process, which is independent of the information supplied by the dividend history. More precisely, we have

$$d\Theta(t) = -\alpha_\Theta \Theta(t) dt + \sigma_\Theta dw_\Theta(t), \quad (2)$$

where α_Θ and σ_Θ are constant positive parameters, and $w_\Theta(t)$ is a Wiener process, which is independent of $w_D(t)$.

Also the insider trader trades smoothly. More precisely, writing $\Psi(t)$ for the insider trader's inventory and $\Phi(t)$ for his order flow, we have that

$$d\Psi(t) = \Phi(t) dt. \quad (3)$$

The insider trader knows the true value of the parameter π in (1) and he can observe the process $\Theta(t)$, because he observes the market order flow. The insider trader is risk averse. He maximizes the expected value of his exponential utility over an infinite time horizon, by controlling the variation of his order flow, $d\Phi(t)$, and his consumption, $c(t)$. Therefore, the insider trader's objective function becomes

$$V(t, m, Y) \stackrel{\text{def}}{=} \sup_{d\Phi(\cdot), c(\cdot)} \{ \mathbf{E}_{t, m, Y} [\int_t^{+\infty} -e^{-(\rho s + \psi c(s))} ds] \}, \quad (4)$$

where $\mathbf{E}_{t, m, Y} [\cdot]$ is the conditional expectation operator given the instant t , the state vector Y of the economy, and the informed trader's wealth m , and where ρ is the discount factor, and ψ is the coefficient of absolute risk-aversion.

The market maker ignores the true value of the parameter π in (1). When making his estimates, he assumes that π is normally distributed. The market maker is risk neutral. He updates his beliefs continuously and sets the price of the risky asset equal to the conditional expected present value of the future dividend stream, given the history of the dividend process and the aggregate order flow information:

$$P(t) \stackrel{\text{def}}{=} \mathbf{E} [\int_t^{+\infty} e^{-r(s-t)} D(s) ds | \mathfrak{F}_t], \quad (5)$$

where \mathfrak{F}_t is the σ -field generated by the dividend process and the total order flow up to t . Note that the market maker observes directly only the aggregate order flow

$$O(t) \stackrel{\text{def}}{=} \Theta(t) + \Phi(t). \quad (6)$$

3 Optimal Filtering by the Market Maker

Our first step is to adopt the market maker's point of view. As above discussed, the market maker only observes the dividend flow $D(t)$ and the aggregate order flow $O(t)$. On the basis of this information, he aims to compute the estimates $\pi_e(t)$ and $\Phi_e(t)$ of the parameter π and of the informed

trader's order flow $\Phi(t)$, respectively. The model is described by the system of stochastic differential equations

$$\begin{cases} dD(t) = \pi dt + \sigma_D dw_D(t), \\ d\Theta(t) = -\alpha_\Theta \Theta(t) dt + \sigma_\Theta dw_\Theta(t), \\ d\Psi(t) = \Phi(t) dt, \end{cases} \quad (7)$$

which can be managed more easily by introducing a suitable matrix notation. Actually, setting $y^\top(t) \equiv (D(t), \pi, \Theta(t), \Phi(t), \Psi(t))$, we can write

$$dy(t) = Ay(t) dt + Q^{1/2} dw(t) + k_\Phi d\Phi(t), \quad (8)$$

where

$$A \equiv \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & -\alpha_\Theta & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix}, \quad Q^{1/2} \equiv \begin{pmatrix} \sigma_D & 0 \\ 0 & 0 \\ 0 & \sigma_\Theta \\ 0 & 0 \\ 0 & 0 \end{pmatrix}, \quad k_\Phi \equiv \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \quad (9)$$

and

$$w(t) \equiv \begin{pmatrix} w_D(t) \\ w_\Theta(t) \end{pmatrix}. \quad (10)$$

In addition, we write $y_o^\top(t) \equiv (D(t), O(t))$ for the observation vector, and $y_e^\top(t) \equiv (D_e(t), \pi_e(t), \Theta_e(t), \Phi_e(t), \Psi_e(t))$ for the market maker's estimate vector of the state vector $y(t)$, given the dividend and the aggregate order flow information. Then we have

$$y_o(t) = M^\top y(t), \quad (11)$$

where

$$M^\top \equiv \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \end{pmatrix},$$

and

$$y_e(t) = \mathbf{E}[y(t) | \mathfrak{F}_t^{y_o}], \quad (12)$$

where $\mathfrak{F}_t^{y_o}$ is the σ -field generated by $y_o(s)$, for $s \leq t$.

In order to perform his estimates, the market maker postulates that the informed trader adjusts his trading strategy linearly. Then we are able to show that also the market maker can adjust his beliefs linearly. The following Proposition can be established, see the Appendix for the proof.

Proposition 1 *Let us assume that the market maker postulates the informed trader's order flow $\Phi(t)$ satisfies the linear stochastic differential equation*

$$d\Phi(t) = a^\top y(t) dt + b^\top y_e(t) dt + q^\top dw(t), \quad (13)$$

for suitable vectors

$$a^\top \equiv (a_1, a_2, a_3, a_4, a_5), \quad b^\top \equiv (b_1, b_2, b_3, b_4, b_5), \quad q^\top \equiv (q_1, q_2). \quad (14)$$

Then the market maker can write an evolution equation for $y_e(t)$ in the form

$$dy_e(t) = G_e(t)y_e(t) dt + H_e(t)dy_o(t), \quad (15)$$

where, setting

$$\hat{A} \equiv A + k_\Phi a^\top, \quad \hat{B} \equiv k_\Phi b^\top, \quad \hat{Q}^{1/2} \equiv Q^{1/2} + k_\Phi q^\top, \quad (16)$$

the matrices $G_e(t)$ and $H_e(t)$ are given by

$$H_e(t) = (\hat{Q} + \Sigma(t)\hat{A}^\top)MR^{-1} \quad (17)$$

and

$$G_e(t) = (I - H_e(t)M^\top)(\hat{A} + \hat{B}), \quad (18)$$

for

$$\hat{Q} \equiv \hat{Q}^{1/2}(\hat{Q}^{1/2})^\top, \quad R \equiv M^\top \hat{Q} M,$$

and $\Sigma(t)$ is a positive solution of the Riccati equation

$$\begin{aligned} d\Sigma(t) = & (\hat{A}\Sigma(t) + \Sigma(t)\hat{A}^\top + \hat{Q}) dt \\ & - (\Sigma(t)\hat{A}^\top + \hat{Q})MR^{-1}M^\top(\Sigma(t)\hat{A} + \hat{Q})^\top dt. \end{aligned} \quad (19)$$

To establish the existence of a stationary Bayesian-Nash equilibrium, we restrict our attention to $G_e(t)$ and $H_e(t)$ constant over time and to the stationary solutions of (19), which arise as positive solutions of the algebraic Riccati equation

$$\hat{A}\Sigma + \Sigma\hat{A}^\top + \hat{Q} - (\Sigma\hat{A}^\top + \hat{Q})MR^{-1}M^\top(\Sigma\hat{A} + \hat{Q})^\top = 0. \quad (20)$$

These stationary solutions are steady-state solutions of (19), to which the ordinary solutions converge as the initial time t_0 goes to $-\infty$, under rather mild conditions (see, e.g. [1]).

For computational purposes, it is useful to show that (20) can be reduced to an equivalent 3×3 equation. Actually, we have the following Proposition, see the Appendix for the proof.

Proposition 2 Let $\tilde{M} \equiv (\hat{M}, \hat{M}_\perp)$, where

$$\hat{M}^\top \equiv \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/\sqrt{2} & 1/\sqrt{2} & 0 \end{pmatrix}$$

is the matrix whose rows are an orthonormal basis for the linear span of the rows of M^\top in \mathbb{R}^5 , and

$$\hat{M}_\perp^\top \equiv \begin{pmatrix} 0 & 0 & 1/\sqrt{2} & -1/\sqrt{2} & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix},$$

is the matrix whose rows are an orthonormal basis for the subspace of \mathbb{R}^5 which is orthogonal to the above-mentioned linear span. Then (19) is equivalent to

$$\begin{aligned} d(\hat{M}_\perp^\top \Sigma(t) \hat{M}_\perp) &= \hat{M}_\perp^\top (\hat{A} \Sigma(t) + \Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M}_\perp dt \\ &\quad - \hat{M}_\perp^\top (\Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q})^\top \hat{M}_\perp dt, \end{aligned} \quad (21)$$

where

$$\hat{R} \equiv \hat{M}^\top \hat{Q} \hat{M}.$$

Finally, we are left with the task of finding a positive solution of the algebraic Riccati equation

$$\hat{M}_\perp^\top (\hat{A} \Sigma + \Sigma \hat{A}^\top + \hat{Q}) \hat{M}_\perp - \hat{M}_\perp^\top (\Sigma \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \hat{M}^\top (\Sigma \hat{A}^\top + \hat{Q})^\top \hat{M}_\perp = 0. \quad (22)$$

4 The Optimal Investment-Consumption problem for the insider trader

As already discussed, the insider trader aims to maximize the expected value of his exponential intertemporal utility over an infinite time horizon, by controlling the state vector of the economy, actually the stacked vector $Y(t)$, and his wealth $m(t)$, by means of the variation of his order flow $d\Phi(t)$ and his consumption rate $c(t)$.

Under the assumption that the insider trader postulates a linear adjustment of the market makers' estimates $y_e(t)$, as in (15), we can write

$$dy_e(t) = G_e y_e(t) dt + H_e M^\top (A y(t) dt + Q^{1/2} dw(t) + k_\Phi d\Phi(t)), \quad (23)$$

where

$$H_e = (\hat{Q} + \Sigma \hat{A}^\top) M R^{-1} \quad (24)$$

and

$$G_e = (I - H_e M^\top) (\hat{A} + \hat{B}), \quad (25)$$

with Σ solution to the algebraic Riccati equation (22). Hence, it is easy to see that the stacked vector $Y(t)$ satisfies

$$dY(t) = \bar{A}Y(t) dt + \bar{Q}^{1/2} dw(t) + \bar{K}_\Phi d\Phi(t), \quad (26)$$

where

$$\bar{A} \equiv \begin{pmatrix} A & 0 \\ H_e M^\top A & G_e \end{pmatrix}, \quad \bar{Q}^{1/2} \equiv \begin{pmatrix} Q^{1/2} \\ H_e M^\top Q^{1/2} \end{pmatrix}, \quad \bar{K}_\Phi \equiv \begin{pmatrix} k_\Phi \\ H_e M^\top k_\Phi \end{pmatrix}. \quad (27)$$

The informed trader's wealth, $m(t)$, is modelled as a solution of the stochastic differential equation

$$dm(t) = rm(t) dt + \Psi(t) [(D(t) - rP(t)) dt + dP(t)] - c(t) dt, \quad (28)$$

where the risky asset price, $P(t)$, is given by

$$P(t) = \mathbf{E} \left[\int_t^{+\infty} e^{-r(s-t)} D(s) ds \middle| \mathfrak{F}_t^{y_0} \right].$$

On the other hand, since for every $s > t$ we have

$$\begin{aligned} \mathbf{E} [D(s) | \mathfrak{F}_t^{y_0}] &= \mathbf{E} [D(t) | \mathfrak{F}_t^{y_0}] + \mathbf{E} [D(s) - D(t) | \mathfrak{F}_t^{y_0}] \\ &= D(t) + \mathbf{E} [\pi(s-t) + \sigma_D (w_D(s) - w_D(t)) | \mathfrak{F}_t^{y_0}] \\ &= D(t) + \pi_e(t)(s-t), \end{aligned}$$

being $w_D(s) - w_D(t)$ future with respect to \mathfrak{F}_t , we can write

$$P(t) = r^{-1}D(t) + r^{-2}\pi_e(t) \equiv p^\top Y(t), \quad (29)$$

where $p^\top \equiv (r^{-1}, 0, 0, 0, 0, 0, r^{-2}, 0, 0, 0)$. Therefore, combining (28) with (26) through (29), and observing that we can also write

$$\Psi(t) = k_\Psi^\top Y(t) = Y^\top(t) k_\Psi, \quad \text{and} \quad D(t) - rP(t) = -r^{-1}k_{\pi_e}^\top Y(t),$$

where $k_\Psi^\top \equiv (0, 0, 0, 0, 1, 0, 0, 0, 0, 0)$, and $k_{\pi_e}^\top \equiv (0, 0, 0, 0, 0, 0, 1, 0, 0, 0)$, we obtain

$$dm(t) = \left(rm(t) - Y^\top(t) k_\Psi \left(r^{-1}k_{\pi_e}^\top Y(t) - p^\top dY(t) \right) - c(t) \right) dt. \quad (30)$$

In what follows we show that there exists a linear diffusion evolution for $d\Phi(t)$ and a consumption strategy which allow the informed trader to maximize his intertemporal utility. Note that the insider trader knows $Y(t)$ and $m(t)$ at any t before adjusting his trading strategy. In other words, he enjoys a complete observation. Therefore (see [5], [6]), it makes sense that he adjusts the variation of his order flow in a feedback form. Moreover, since the state vector of the economy follows an autonomous system of stochastic differential equations, he can choose a stationary Markov policy

$$d\Phi(t) = a^\top(t) dt + q(t) dw(t), \quad (31)$$

where

$$a(t) \equiv a(Y(t), m(t)), \quad q(t) \equiv q(Y(t), m(t)). \quad (32)$$

The same argument holds for the choice of the insider trader's consumption

$$c(t) \equiv c(Y(t), m(t)).$$

Hence, the insider trader's optimization problem becomes to compute the value function

$$V(t, Y, m) \equiv \max_{a(\cdot), q(\cdot), c(\cdot)} \left\{ \mathbf{E}_{t, Y, m} \left[\int_t^{+\infty} -e^{-(\rho s + \psi c(s))} ds \right] \right\}, \quad (33)$$

where $\mathbf{E}_{t, Y, m}[\cdot]$ is the conditional expectation operator given $Y(t) = Y$ and $m(t) = m$, and the state vector $(Y(t), m(t))$ is subject to

$$dY(t) = \left(\bar{A}Y(t) + \bar{K}_\Phi a^\top(t) \right) dt + \left(\bar{Q}^{1/2} + \bar{K}_\Phi q(t) \right) dw(t), \quad (34)$$

which results from the combination of (26) and (31), and

$$\begin{aligned} dm(t) = & \left(rm(t) - Y^\top(t) k_\Psi \left(r^{-1} k_{\pi_e}^\top Y(t) - p^\top \left(\bar{A}Y(t) + \bar{K}_\Phi a^\top(t) \right) \right) - c(t) \right) dt \\ & + Y^\top(t) k_\Psi p^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q(t) \right) dw(t), \end{aligned} \quad (35)$$

which results from the combination of (30) and (34). In addition, note that the entries of $Y(t)$ must satisfy the relationships

$$y_1(t) \equiv D(t) = D_e(t) \equiv y_6(t), \quad (36)$$

$$y_3(t) + y_4(t) \equiv \Theta(t) + \Phi(t) \equiv O(t) = O_e(t) = \Theta_e(t) + \Phi_e(t) \equiv y_8(t) + y_9(t) \quad (37)$$

for every $t \geq 0$. The following Proposition can be established, see the Appendix for the proof.

Proposition 3 *Let us assume that (33) is of the form*

$$V(t, Y, m) = -e^{-(\rho t + \frac{1}{2} Y^\top L Y + \psi r m + \lambda)}, \quad (38)$$

where L is a symmetric matrix and λ is a real parameter. Then, L must satisfy the equations

$$\bar{K}_\Phi^\top \left(L + \psi r p k_\Psi^\top \right) Y = 0, \quad (39)$$

and

$$\begin{aligned} & Y^\top \left(rL + \left(L + \psi r p k_\Psi^\top \right)^\top \bar{Q} \left(L + \psi r p k_\Psi^\top \right) \right. \\ & \left. - \left(L + \psi r p k_\Psi^\top \right)^\top \bar{A} - \bar{A}^\top \left(L + \psi r p k_\Psi^\top \right) - \psi \left(k_\Psi k_{\pi_e}^\top + k_{\pi_e} k_\Psi^\top \right) \right) Y = 0, \end{aligned} \quad (40)$$

for every $Y \in \mathbb{R}^{10}$ whose entries are characterized by (36) and (37), and λ satisfies

$$r(1 + \lambda - \ln(r)) - \rho - \frac{1}{2} \text{tr} \left(\Upsilon^\top L \Upsilon \right) = 0, \quad (41)$$

where $\Upsilon \equiv \bar{Q}^{1/2} - (\bar{K}_\Phi^\top L \bar{K}_\Phi)^{-1} \bar{K}_\Phi \bar{K}_\Phi^\top L \bar{Q}^{1/2}$. Moreover, the optimal trading strategy is given by

$$\dot{a}(t) = -(\bar{K}_\Phi^\top L \bar{K}_\Phi)^{-1} (\bar{K}_\Phi^\top L \bar{A} + \psi r \bar{K}_\Phi^\top p k_\Phi^\top) \dot{Y}(t) \quad (42)$$

and

$$\dot{q}(t) = -(\bar{K}_\Phi^\top L \bar{K}_\Phi)^{-1} \bar{K}_\Phi^\top L \bar{Q}^{1/2}, \quad (43)$$

and the optimal consumption is

$$\dot{c}(t) = \frac{\frac{1}{2} \dot{Y}^\top(t) L \dot{Y}(t) + \psi r \dot{m}(t) + \lambda - \ln(r)}{\psi}, \quad (44)$$

where $(\dot{Y}(t), \dot{m}(t))$ is the solution of (34) and (35) corresponding to the choice of the control $(\dot{a}(t), \dot{q}(t), \dot{c}(t))$.

Remark 4 *Note that, by applying a well known verification theorem (see e.g., Fleming & Rishel [5, Thm 4.1, p. 159] (1975), Fleming & Soner [6, Thm 5.1, p. 172] (1993)), it can be shown that (38) is actually the value function (33), for the choice of the symmetric matrix L and the real number λ satisfying (39), (40), and (41).*

5 Bayesian-Nash Equilibrium

To get a Bayesian Nash equilibrium we have to find a positive symmetric solution Σ of the algebraic Riccati Equation

$$\hat{M}_\perp^\top (\hat{A}\Sigma + \Sigma\hat{A}^\top + \hat{Q})\hat{M}_\perp - \hat{M}_\perp^\top (\Sigma\hat{A}^\top + \hat{Q})\hat{M}\hat{R}^{-1}\hat{M}^\top (\Sigma\hat{A}^\top + \hat{Q})^\top \hat{M}_\perp = 0, \quad (45)$$

a symmetric matrix L which fulfills

$$\bar{K}_\Phi^\top \left(L + \psi r p k_\Psi^\top \right) Y = 0, \quad (46)$$

and

$$\begin{aligned} & Y^\top \left(rL + \left(L + \psi r p k_\Psi^\top \right)^\top \bar{Q} \left(L + \psi r p k_\Psi^\top \right) \right. \\ & \left. - \left(L + \psi r p k_\Psi^\top \right)^\top \bar{A} - \bar{A}^\top \left(L + \psi r p k_\Psi^\top \right) - \psi \left(k_\Psi k_{\pi_e}^\top + k_{\pi_e} k_\Psi^\top \right) \right) Y = 0, \end{aligned} \quad (47)$$

for every $Y \in \mathbb{R}^{10}$ satisfying (36) and (37), and a set of parameters

$$a^\top \equiv (a_1, a_2, a_3, a_4, a_5), \quad b^\top \equiv (b_1, b_2, b_3, b_4, b_5), \quad q^\top \equiv (q_1, q_2),$$

such that

$$\begin{aligned} & (a^\top, b^\top) Y(t) dt + q dw(t) \\ & = -(\bar{K}_\Phi^\top L \bar{K}_\Phi)^{-1} \left(\left(\bar{K}_\Phi^\top L \bar{A} + \psi r \bar{K}_\Phi^\top p k_\Phi^\top \right) Y(t) dt + \bar{K}_\Phi^\top L \bar{Q}^{1/2} dw(t) \right). \end{aligned} \quad (48)$$

Equation (45) comes from the optimal filtering problem of the market maker; Equation (46) and (47) come from the optimal consumption-investment problem of the insider trader. The first one is needed to eliminate the linear term in the Bellmann equation related to (33) and the second one represents the first-order conditions on the quadratic term of the Bellmann equation. Equation (48) comes from a fixed point argument requiring that the optimal strategy of the insider, conjectured by the market maker, coincides with the optimal strategy of the insider trader and vice versa.

The computations needed to this task are rather heavy. However, we are able to prove that there exists a unique linear Bayesian-Nash equilibrium such that no insider's trade and no market maker's learning occur. Also in this case, see the Appendix for the proof.

Proposition 5 *In the linear Bayesian-Nash equilibrium we have*

$$q_1 = q_2 = a_1 = a_2 = a_3 = a_4 = a_5 = b_1 = b_2 = b_3 = b_4 = b_5 = 0,$$

$$\Sigma = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \sigma_{5,5} \end{pmatrix}, \quad (49)$$

for an unconstrained coefficient $\sigma_{5,5} > 0$, and

$$L = \begin{pmatrix} \ell_{1,1} & -\ell_{2,6} & -\ell_{3,6} & -\ell_{4,6} & -\ell_{5,6} & \ell_{1,6} & -\ell_{6,7} & -\ell_{6,8} & -\ell_{6,9} & -\ell_{6,10} \\ -\ell_{2,6} & 0 & 0 & 0 & 0 & \ell_{2,6} & 0 & 0 & 0 & 0 \\ -\ell_{3,6} & 0 & \ell_{9,9} & \ell_{9,9} & 0 & \ell_{3,6} & 0 & -\ell_{9,9} & -\ell_{9,9} & 0 \\ -\ell_{4,6} & 0 & \ell_{9,9} & \ell_{9,9} & 0 & \ell_{4,6} & 0 & -\ell_{9,9} & -\ell_{9,9} & 0 \\ -\ell_{5,6} & 0 & 0 & 0 & -\frac{\sigma_D^2 \psi^2}{r} & \ell_{5,6} & \frac{\psi}{r} & 0 & 0 & 0 \\ \ell_{1,6} & \ell_{2,6} & \ell_{3,6} & \ell_{4,6} & \ell_{5,6} & \ell_{6,6} & \ell_{6,7} & \ell_{6,8} & \ell_{6,9} & \ell_{6,10} \\ -\ell_{6,7} & 0 & 0 & 0 & \frac{\psi}{r} & \ell_{6,7} & 0 & 0 & 0 & 0 \\ -\ell_{6,8} & 0 & -\ell_{9,9} & -\ell_{9,9} & 0 & \ell_{6,8} & 0 & \ell_{9,9} & \ell_{9,9} & 0 \\ -\ell_{6,9} & 0 & -\ell_{9,9} & -\ell_{9,9} & 0 & \ell_{6,9} & 0 & \ell_{9,9} & \ell_{9,9} & 0 \\ -\ell_{6,10} & 0 & 0 & 0 & 0 & \ell_{6,10} & 0 & 0 & 0 & 0 \end{pmatrix}, \quad (50)$$

where

$$\ell_{1,1} = -(2\ell_{1,6} + \ell_{6,6}), \quad \ell_{9,9} \neq 0,$$

and the other entries $\ell_{j,k}$ can be chosen freely.

6 Conclusions

This paper provides a striking result. In the Kyle setting with noise traders, when the insider trader knows the drift of the dividend process, no trade occurs. Noise traders are not enough to hide the trading of the insider: as the insider trades, the market maker identifies exactly his private information, sets the market price according to it and this renders the trade of the insider not profitable. As a consequence, in equilibrium the insider trader does not trade, there is no learning by the market maker and no diffusion of information in the market.

Appendix

Proof. of Proposition 1

Substituting (13) into (8), we obtain

$$dy(t) = \hat{A}y(t) dt + \hat{B}y_e(t) dt + \hat{Q}^{1/2} dw(t). \quad (51)$$

Then, combining (51) and (11), we have

$$dy_o(t) = M^\top \hat{A}y(t) dt + M^\top \hat{B}y_e(t) dt + M^\top \hat{Q}^{1/2} dw(t), \quad (52)$$

and, substituting the latter into (15) we can write

$$dy_e(t) = \hat{A}_e(t)y(t) dt + \hat{B}_e(t)y_e(t) dt + \hat{Q}_e^{1/2}(t) dw(t), \quad (53)$$

where

$$\hat{A}_e(t) \equiv H_e(t)M^\top \hat{A}, \quad \hat{B}_e(t) \equiv G_e(t) + H_e(t)M^\top \hat{B}, \quad \hat{Q}_e^{1/2}(t) \equiv H_e(t)M^\top \hat{Q}^{1/2}.$$

Hence, introducing the stacked vector $Y(t) \equiv (Y_j(t))_{j=1}^{10}$, given by

$$Y(t) \equiv \begin{pmatrix} y(t) \\ y_e(t) \end{pmatrix},$$

on account of (51) and (53), it is easy to see that $Y(t)$ satisfies the stochastic differential equation

$$dY(t) = \tilde{A}(t)Y(t) dt + \tilde{Q}^{1/2}(t) dw(t),$$

where

$$\tilde{A}(t) = \begin{pmatrix} \hat{A} & \hat{B} \\ \hat{A}_e(t) & \hat{B}_e(t) \end{pmatrix}, \quad \tilde{Q}^{1/2}(t) = \begin{pmatrix} \hat{Q}^{1/2} \\ \hat{Q}_e^{1/2}(t) \end{pmatrix}. \quad (54)$$

On the other hand, thanks to (52), we can formulate the evolution of $y_o(t)$ in terms of $Y(t)$ by writing

$$dy_o(t) = CY(t) dt + R^{1/2} dw(t),$$

where

$$C \equiv (M^\top \hat{A}, M^\top \hat{B}), \quad R^{1/2} \equiv M^\top \hat{Q}^{1/2}. \quad (55)$$

Therefore, the filtering problem becomes

$$\begin{cases} dY(t) = \tilde{A}(t)Y(t) dt + \tilde{Q}^{1/2}(t) dw(t), \\ dy_o(t) = CY(t) dt + R^{1/2} dw(t). \end{cases} \quad (56)$$

The latter is a generalization of the standard Kalman-Bucy linear filtering problem, because the matrix

$$R \equiv R^{1/2}(R^{1/2})^\top = \begin{pmatrix} \sigma_D^2 & \sigma_{Dq_1} \\ \sigma_{Dq_1} & (\sigma_\Theta + q_2)^2 + q_1^2 \end{pmatrix}$$

has positive eigenvalues, and we assume that for some initial time t_0 the prior $Y(t_0) \sim \mathcal{N}(Y_0, \tilde{\Sigma}_0)$ is independent of the Wiener process $w(t)$. Hence, we are in a position to apply [14, Thm 10.3, p. 392] to obtain the equation for the optimal estimate $Y_e(t) \equiv \mathbf{E}[Y(t)|\mathfrak{F}_t^{y_0}]$ for every $t \geq t_0$. This is given by

$$\begin{aligned} dY_e(t) &= \tilde{A}(t)Y_e(t) dt \\ &+ (\tilde{Q}^{1/2}(t)(R^{1/2})^\top + \tilde{\Sigma}(t)C)R^{-1}(C(Y(t) - Y_e(t)) dt + R^{1/2}dw(t)), \end{aligned} \quad (57)$$

where, setting $\tilde{Q}(t) \equiv \tilde{Q}^{1/2}(t)(\tilde{Q}^{1/2}(t))^\top$, the matrix

$$\tilde{\Sigma}(t) \equiv \mathbf{E}[(Y(t) - Y_e(t))(Y(t) - Y_e(t))^\top]$$

satisfies the Riccati equation

$$\begin{aligned} d\tilde{\Sigma}(t) &= (\tilde{A}(t)\tilde{\Sigma}(t) + \tilde{\Sigma}(t)\tilde{A}^\top(t) + \tilde{Q}(t)) dt \\ &- (\tilde{Q}^{1/2}(t)(R^{1/2})^\top + \tilde{\Sigma}(t)C^\top)R^{-1}(\tilde{Q}^{1/2}(t)(R^{1/2})^\top + \tilde{\Sigma}(t)C^\top)^\top dt. \end{aligned} \quad (58)$$

On the other hand, $y_e(t)$ is the market maker's estimate of $y(t)$, therefore we must clearly have

$$y_e(t) = \mathbf{E}[y_e(t)|\mathfrak{F}_t^{y_0}].$$

Hence,

$$Y_e(t) = \begin{pmatrix} y_e(t) \\ y_e(t) \end{pmatrix},$$

and it is easily seen that $\tilde{\Sigma}(t)$ can be decomposed into four 5×5 blocks as follows

$$\tilde{\Sigma}(t) = \begin{pmatrix} \Sigma(t) & 0 \\ 0 & 0 \end{pmatrix}, \quad (59)$$

where $\Sigma(t) \equiv \mathbf{E}[(y(t) - y_e(t))(y(t) - y_e(t))^\top]$. Then we have

$$\tilde{\Sigma}(t)C^\top = \begin{pmatrix} \Sigma(t) & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \hat{A}M \\ \hat{B}M \end{pmatrix} = \begin{pmatrix} \Sigma(t)\hat{A}M \\ 0 \end{pmatrix},$$

and

$$\tilde{Q}^{1/2}(R^{1/2})^\top + \tilde{\Sigma}(t)C^\top = \begin{pmatrix} (\hat{Q} + \Sigma(t)\hat{A}^\top)M \\ H_e(t)R \end{pmatrix}.$$

where $\hat{Q} \equiv \tilde{Q}^{1/2}(\tilde{Q}^{1/2})^\top$. Hence, (57) becomes

$$\begin{aligned} dY_e(t) &= \begin{pmatrix} \hat{A} & \hat{B} \\ H_e(t)M^\top \hat{A} & G_e(t) + H_e(t)M^\top \hat{B} \end{pmatrix} Y_e(t) dt \\ &+ \begin{pmatrix} (\hat{Q} + \Sigma(t)\hat{A}^\top)M \\ H_e(t)R \end{pmatrix} R^{-1}M^\top (\hat{A}(y(t) - y_e(t)) dt + R^{1/2} dw(t)). \end{aligned} \quad (60)$$

On the other and, on account of (51),

$$\hat{A}(y(t) - y_e(t)) dt + R^{1/2} dw(t) = -(\hat{A} + \hat{B})y_e(t) + dy(t)$$

and thanks to (11), we can rewrite (60) as

$$\begin{aligned} \begin{pmatrix} dy_e(t) \\ dy_e(t) \end{pmatrix} &= \begin{pmatrix} (\hat{A} + \hat{B} - (\hat{Q} + \Sigma(t)\hat{A}^\top)MR^{-1}M^\top(\hat{A} + \hat{B}))y_e(t) dt \\ G_e(t)y_e(t) dt \end{pmatrix} \\ &+ \begin{pmatrix} (\hat{Q} + \Sigma(t)\hat{A}^\top)MR^{-1} dy_o(t) \\ H_e(t) dy_o(t) \end{pmatrix}, \end{aligned}$$

and the latter clearly implies that (17) and (18) are fulfilled. Now, a straightforward computation gives

$$\tilde{A}\tilde{\Sigma}(t) + \tilde{\Sigma}(t)\tilde{A}^\top + \tilde{Q} = \begin{pmatrix} \hat{A}\Sigma(t) + \Sigma(t)\hat{A}^\top + \hat{Q} & (H_e(t)M^\top(\hat{A}\Sigma(t) + \hat{Q}))^\top \\ H_e(t)M^\top(\hat{A}\Sigma(t) + \hat{Q}) & H_e(t)RH_e^\top(t) \end{pmatrix},$$

and

$$\begin{aligned} &(\tilde{Q}^{1/2}(R^{1/2})^\top + \tilde{\Sigma}(t)C^\top)R^{-1}(\tilde{Q}^{1/2}(R^{1/2})^\top + \tilde{\Sigma}(t)C^\top)^\top \\ &= \begin{pmatrix} (\hat{Q} + \Sigma(t)\hat{A}^\top)MR^{-1}M^\top(\hat{Q} + \Sigma(t)\hat{A}^\top)^\top & (H_e(t)M^\top(\hat{A}\Sigma(t) + \hat{Q}))^\top \\ H_e(t)M^\top(\hat{A}\Sigma(t) + \hat{Q}) & H_e(t)RH_e^\top(t) \end{pmatrix}. \end{aligned}$$

Therefore, on account of (58) and (59), (19) immediately follows. \square

Proof. of Proposition 2

It is evident that with respect to the vector $\hat{y}_o(t) \equiv \hat{M}^\top y(t)$, which carries the same information of the observation vector $y_o(t)$, (19) becomes

$$\begin{aligned} d\Sigma(t) &= (\hat{A}\Sigma(t) + \Sigma(t)\hat{A}^\top + \hat{Q}) dt \\ &- (\Sigma(t)\hat{A}^\top + \hat{Q})\hat{M}\hat{R}^{-1}\hat{M}^\top(\Sigma(t)\hat{A}^\top + \hat{Q})^\top dt. \end{aligned} \quad (61)$$

Now, since the columns of the matrix \tilde{M} make up an orthonormal basis of \mathbb{R}^5 , (61) is equivalent to

$$\begin{aligned} d(\tilde{M}^\top \Sigma(t) \tilde{M}) &= \tilde{M}^\top (\hat{A} \Sigma(t) + \Sigma(t) \hat{A}^\top + \hat{Q}) \tilde{M}^\top dt \\ &\quad - \tilde{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \tilde{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q})^\top \tilde{M}^\top dt, \end{aligned} \quad (62)$$

which splits in (21) plus the three following

$$\begin{aligned} d(\hat{M}^\top \Sigma(t) \hat{M}) &= \hat{M}^\top (\hat{A} \Sigma(t) + \Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} dt \\ &\quad - \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q})^\top \hat{M} dt, \end{aligned} \quad (63)$$

$$\begin{aligned} d(\hat{M}^\top \Sigma(t) \hat{M}_\perp) &= \hat{M}^\top (\hat{A} \Sigma(t) + \Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M}_\perp dt \\ &\quad - \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q})^\top \hat{M}_\perp dt, \end{aligned} \quad (64)$$

$$\begin{aligned} d(\hat{M}_\perp^\top \Sigma(t) \hat{M}) &= \hat{M}_\perp^\top (\hat{A}^\top \Sigma(t) + \Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} dt \\ &\quad - \hat{M}_\perp^\top (\Sigma(t) \hat{A}^\top + \hat{Q}) \hat{M} \hat{R}^{-1} \hat{M}^\top (\Sigma(t) \hat{A}^\top + \hat{Q})^\top \hat{M} dt, \end{aligned} \quad (65)$$

On the other hand, the first and the second column of the matrix $\Sigma(t) \hat{M}$ are given by

$$\begin{pmatrix} \mathbf{E}[(D(t) - D_e(t))(D(t) - D_e(t))] \\ \mathbf{E}[(D(t) - D_e(t))(\pi(t) - \pi_e(t))] \\ \mathbf{E}[(D(t) - D_e(t))(\Theta(t) - \Theta_e(t))] \\ \mathbf{E}[(D(t) - D_e(t))(\Phi(t) - \Phi_e(t))] \\ \mathbf{E}[(D(t) - D_e(t))(\Psi(t) - \Psi_e(t))] \end{pmatrix},$$

and

$$1/\sqrt{2} \begin{pmatrix} \mathbf{E}[(O(t) - O_e(t))(D(t) - D_e(t))] \\ \mathbf{E}[(O(t) - O_e(t))(\pi(t) - \pi_e(t))] \\ \mathbf{E}[(O(t) - O_e(t))(\Theta(t) - \Theta_e(t))] \\ \mathbf{E}[(O(t) - O_e(t))(\Phi(t) - \Phi_e(t))] \\ \mathbf{E}[(O(t) - O_e(t))(\Psi(t) - \Psi_e(t))] \end{pmatrix},$$

respectively. Therefore, since both the processes $D(t)$ and $O(t)$ are observed, the matrix $\Sigma(t) \hat{M}$ vanishes identically. Hence, a straightforward computation shows that also (63)-(65) vanish identically, and the desired result follows. \square

Proof of Proposition 3

It is well known that (33) must satisfy both the Bellman equation

$$\partial_t V(t, Y, m) + \max_{a, q, c} \{ \mathcal{L}V(t, Y, m) - e^{-(\rho t + \psi c)} \} = 0, \quad (66)$$

where \mathcal{L} is the infinitesimal generator of the diffusion process (34), (35), and the transversality condition

$$\lim_{T \rightarrow +\infty} \mathbf{E}_{t, m, Y} \left[V(t+T, \dot{Y}(t+T), \dot{m}(t+T)) \right] = 0, \quad (67)$$

where $(\dot{Y}(t), \dot{m}(t))$ is a solution of (34) and (35) corresponding to the choice of an optimal control $(\dot{a}(t), \dot{q}(t), \dot{c}(t))$. In this case, a feedback optimal control is given by

$$(\dot{a}(t), \dot{q}(t), \dot{c}(t)) = \arg \max_{a(t), q(t), c(t)} \{ \mathcal{L}V(t, Y(t), m(t)) - e^{-(\rho t + \psi c(t))} \}, \quad (68)$$

where $(Y(t), m(t))$ is a solution of (34) and (35) corresponding to the choice of the control $(a(t), q(t), c(t))$.

To prove that the function $V(t, Y, m) = -e^{-(\rho t + \frac{1}{2} Y^\top L Y + \psi r m + \lambda)}$ is a solution to (66), let's start by computing the operator \mathcal{L} . It is easily seen that

$$\begin{aligned} \mathcal{L} \equiv & \frac{1}{2} \sum_{i, j=1}^{10} \left((\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top \right)_{i, j} \partial_{Y_i, Y_j}^2 \\ & + Y^\top k_\Psi \sum_{j=1}^{10} \left(p^\top (\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top \right)_j \partial_{m, Y_j}^2 \\ & + \frac{1}{2} Y^\top k_\Psi p^\top (\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top p k_\Psi^\top Y \partial_{m, m}^2 \\ & + \sum_{j=1}^{10} (\bar{A} Y + \bar{K}_\Phi a^\top)_j \partial_{Y_j} \\ & + \left(r m - Y^\top k_\Psi \left(r^{-1} k_{\pi_e}^\top Y - p^\top (\bar{A} Y + \bar{K}_\Phi a^\top) \right) - c \right) \partial_m. \end{aligned} \quad (69)$$

Hence, on account of

$$\begin{aligned}
\partial_t V &= -\rho V, \\
\partial_{Y_j} V &= -(Y^\top L)_j V, \\
\partial_m V &= -\psi r V, \\
\partial_{Y_i, Y_j}^2 V &= (LYY^\top L - L)_{i,j} V, \\
\partial_{Y_j, m}^2 V &= \psi r (Y^\top L)_j V, \\
\partial_{m, m}^2 V &= \psi^2 r^2 V,
\end{aligned}$$

where we are using V as a shorthand for $V(t, Y, m)$, we obtain

$$\begin{aligned}
\mathcal{L}V &= \frac{1}{2} \sum_{i,j=1}^{10} \left((\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top \right)_{i,j} (LYY^\top L - L)_{i,j} V \quad (70) \\
&+ \psi r Y^\top k_\Psi \sum_{j=1}^{10} p^\top (\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)_j^\top (Y^\top L)_j V \\
&+ \frac{1}{2} \psi^2 r^2 Y^\top k_\Psi p^\top (\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top p k_\Psi^\top Y V \\
&- \sum_{j=1}^{10} (\bar{A}Y + \bar{K}_\Phi a^\top)_j (Y^\top L)_j \dot{V} \\
&- \psi r \left(rm - Y^\top k_\Psi \left(r^{-1} k_{\pi_e}^\top Y - p^\top (\bar{A}Y + \bar{K}_\Phi a^\top) \right) - c \right) V.
\end{aligned}$$

On the other hand, thanks to the properties of the trace functional, we can write

$$\begin{aligned}
&\sum_{i,j=1}^{10} \left((\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top \right)_{i,j} (LYY^\top L - L)_{i,j} V \quad (71) \\
&= \text{tr} \left((\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top (LYY^\top L - L) (\bar{Q}^{1/2} + \bar{K}_\Phi q) \right) \\
&= Y^\top L (\bar{Q}^{1/2} + \bar{K}_\Phi q) (\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top LY \\
&\quad - \text{tr} \left((\bar{Q}^{1/2} + \bar{K}_\Phi q)^\top L (\bar{Q}^{1/2} + \bar{K}_\Phi q) \right).
\end{aligned}$$

Moreover,

$$\begin{aligned} \sum_{j=1}^{10} \left(p^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top \right)_j (Y^\top L)_j & \quad (72) \\ & = p^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top LY, \end{aligned}$$

and

$$\sum_{j=1}^{10} \left(\bar{A}Y + \bar{K}_\Phi a^\top \right)_j (Y^\top L)_j = Y^\top L \left(\bar{A}Y + \bar{K}_\Phi a^\top \right). \quad (73)$$

Therefore, combining (70) with (71)-(73), and observing that

$$\begin{aligned} & \frac{1}{2} Y^\top L \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top LY \\ & + \psi r Y^\top k_\Psi p^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top LY \\ & + \frac{1}{2} \psi^2 r^2 Y^\top k_\Psi p^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top p k_\Psi^\top Y \\ & = \frac{1}{2} Y^\top \left(L + \psi r p k_\Psi^\top \right)^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top \left(L + \psi r p k_\Psi^\top \right) Y, \end{aligned}$$

and

$$\begin{aligned} & Y^\top L \left(\bar{A}Y + \bar{K}_\Phi a^\top \right) + \psi r Y^\top k_\Psi p^\top \left(\bar{A}Y + \bar{K}_\Phi a^\top \right) \\ & = Y^\top \left(L + \psi r p k_\Psi^\top \right)^\top \left(\bar{A}Y + \bar{K}_\Phi a^\top \right) \end{aligned}$$

we can rewrite

$$\begin{aligned} \mathcal{L}V & = -\frac{1}{2} \text{tr} \left(\left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top L \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \right) V \\ & + \frac{1}{2} Y^\top \left(L + \psi r p k_\Psi^\top \right)^\top \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right) \left(\bar{Q}^{1/2} + \bar{K}_\Phi q \right)^\top \left(L + \psi r p k_\Psi^\top \right) Y V \\ & - Y^\top \left(L + \psi r p k_\Psi^\top \right)^\top \left(\bar{A}Y + \bar{K}_\Phi a^\top \right) V \\ & - \psi \left(r^2 m + Y^\top k_\Psi p^\top Y \right) V + \psi r c V. \end{aligned} \quad (74)$$

From the above computations, it follows that to achieve the maximum in (66) we must get rid of the linear term

$$-Y^\top \left(L + \psi r p k_\Psi^\top \right)^\top \bar{K}_\Phi a^\top V.$$

To this goal, we need to establish (39). As a consequence, (74) becomes

$$\begin{aligned}
\mathcal{L}V &= -\frac{1}{2}\text{tr}\left(\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)^\top L\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)\right)V \quad (75) \\
&+ \frac{1}{2}Y^\top\left(L + \psi r p k_\Psi^\top\right)^\top \bar{Q}\left(L + \psi r p k_\Psi^\top\right)YV \\
&- Y^\top\left(L + \psi r p k_\Psi^\top\right)^\top \bar{A}YV \\
&- \psi\left(r^2 m + Y^\top k_\Psi k_{\pi_e}^\top Y\right)V + \psi r cV,
\end{aligned}$$

where $\bar{Q} \equiv \bar{Q}^{1/2}(\bar{Q}^{1/2})^\top$. Hence, setting

$$\begin{aligned}
I(t, Y, m, a, q) &\stackrel{\text{def}}{=} -\frac{1}{2}\text{tr}\left(\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)^\top L\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)\right)V \quad (76) \\
&+ \frac{1}{2}Y^\top\left(L + \psi r p k_\Psi^\top\right)^\top \bar{Q}\left(L + \psi r p k_\Psi^\top\right)YV \\
&- Y^\top\left(L + \psi r p k_\Psi^\top\right)^\top \bar{A}YV \\
&- \rho V - \psi\left(r^2 m + Y^\top k_\Psi k_{\pi_e}^\top Y\right)V,
\end{aligned}$$

and

$$J(t, Y, m, c) \stackrel{\text{def}}{=} \psi r cV - e^{-(\rho t + \psi c)}, \quad (77)$$

(66) becomes

$$\max_{a, q} \{I(t, Y, m, a, q)\} + \max_c \{J(t, Y, m, c)\} = 0. \quad (78)$$

Maximizing $I(t, Y, m, a, q)$ with respect to q , the first order conditions yield

$$\partial_{q_j} \text{tr}\left(\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)^\top L\left(\bar{Q}^{1/2} + \bar{K}_\Phi q\right)\right) = 0, \quad (79)$$

for $j = 1, 2, 3$. On the other hand,

$$\begin{aligned}
& \partial_{q_j} \text{tr} \left(\left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right)^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) \right) \\
&= \text{tr} \left(\partial_{q_j} \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right)^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) \right) \\
&= \text{tr} \left(e_j^{\top} \bar{K}_{\Phi}^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) + \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right)^{\top} L \bar{K}_{\Phi} e_j \right) \\
&= \text{tr} \left(e_j^{\top} \bar{K}_{\Phi}^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) \right) + \text{tr} \left(\left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right)^{\top} L \bar{K}_{\Phi} e_j \right) \\
&= 2 \text{tr} \left(e_j^{\top} \bar{K}_{\Phi}^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) \right) \\
&= 2 \bar{K}_{\Phi}^{\top} L \left(\bar{Q}^{1/2} + \bar{K}_{\Phi} q \right) e_j.
\end{aligned}$$

Therefore, from (79), we obtain

$$q = -(\bar{K}_{\Phi}^{\top} L \bar{K}_{\Phi})^{-1} \bar{K}_{\Phi}^{\top} L \bar{Q}^{1/2}. \quad (80)$$

and

$$\begin{aligned}
\max_{a, q} \{I(t, Y, m, a, q)\} &= -\frac{1}{2} \text{tr} \left(Y^{\top} L Y \right) \\
&+ \frac{1}{2} Y^{\top} \left(L + \psi r p k_{\Psi}^{\top} \right)^{\top} \bar{Q} \left(L + \psi r p k_{\Psi}^{\top} \right) Y V \\
&- Y^{\top} \left(L + \psi r p k_{\Psi}^{\top} \right)^{\top} \bar{A} Y V - \rho V - \psi \left(r^2 m + Y^{\top} k_{\Psi} k_{\pi_e}^{\top} Y \right) V.
\end{aligned} \quad (81)$$

Similarly, maximizing $J(t, Y, m, c)$ with respect to c , the first order condition yields

$$e^{-(\rho t + \psi c)} + r V = 0, \quad (82)$$

and this clearly implies

$$c = \frac{\frac{1}{2} Y^{\top} L Y + \psi r m(t) + \lambda - \ln(r)}{\psi},$$

and

$$\max_c \{J(t, Y, m, c)\} = r \left(\frac{1}{2} Y^{\top} L Y + \psi r m + \lambda - \ln(r) + 1 \right) V. \quad (83)$$

Finally, combining (81) with (83), we obtain

$$\begin{aligned}
& \max_{a,q} \{I(t, Y, m, c)\} + \max_c \{J(t, Y, m, c)\} \\
&= (r(\lambda - \ln(r) + 1) - \rho)V - \frac{1}{2} \text{tr} \left(\Upsilon^\top L \Upsilon \right) V \\
&+ Y^\top \left(\frac{1}{2} r L + \frac{1}{2} \left(L + \psi r p k_\Psi^\top \right)^\top \bar{Q} \left(L + \psi r p k_\Psi^\top \right) \right. \\
&\quad \left. - \left(L + \psi r p k_\Psi^\top \right)^\top \bar{A} - \psi k_\Psi k_{\pi_e}^\top \right) Y V.
\end{aligned}$$

The latter clearly implies that $V(t, Y, m)$ is a solution of the Bellman equation (66) provided that (40) and (41) are satisfied.

Now, to show that the transversality condition (67) holds true we apply the Itô formula to the identity

$$\begin{aligned}
& V(t + \Delta t, \dot{Y}(t + \Delta t), \dot{m}(t + \Delta t)) - V(t, \dot{Y}(t), \dot{m}(t)) \\
&= \int_t^{t+\Delta t} dV(s, \dot{Y}(s), \dot{m}(s)).
\end{aligned}$$

Therefore, we can write

$$\begin{aligned}
& V(t + \Delta t, \dot{Y}(t + \Delta t), \dot{m}(t + \Delta t)) - V(t, \dot{Y}(t), \dot{m}(t)) \tag{84} \\
&= \int_t^{t+\Delta t} (\partial_s V(s, \dot{Y}(s), \dot{m}(s)) + \mathcal{L}V(s, \dot{Y}(s), \dot{m}(s))) ds \\
&\quad + \int_t^{t+\Delta t} \sigma(\dot{Y}(s), \dot{m}(s)) \nabla_{Y,m} V(s, \dot{Y}(s), \dot{m}(s)) dw(s),
\end{aligned}$$

where $\sigma(\dot{Y}(t), \dot{m}(t))$ denotes the diffusion matrix of the process $(\dot{Y}(t), \dot{m}(t))$ and $\nabla_{Y,m}$ denotes the gradient operator in the state space of $(\dot{Y}(t), \dot{m}(t))$. On the other hand, since $V(t, Y, m)$ is a solution of the Bellman equation (66) and $(\dot{Y}(t), \dot{m}(t))$ corresponds to an optimal control, we have

$$\begin{aligned}
& \int_t^{t+\Delta t} (\partial_s V(s, \dot{Y}(s), \dot{m}(s)) + \mathcal{L}V(s, \dot{Y}(s), \dot{m}(s))) ds \\
&= \int_t^{t+\Delta t} e^{-(\rho s + \psi \dot{z}(s))} ds.
\end{aligned}$$

On account of the latter, by applying the expectation operator on both the

sides of (84), we obtain

$$\begin{aligned} & \frac{\mathbf{E}_{t,Y,m} \left[V(t + \Delta t, \dot{Y}(t + \Delta t), \dot{m}(t + \Delta t)) \right] - \mathbf{E}_{t,Y,m} \left[\dot{V}(t, \dot{Y}(t), \dot{m}(t)) \right]}{\Delta t} \\ &= \frac{1}{\Delta t} \mathbf{E}_{t,Y,m} \left[\int_t^{t+\Delta t} e^{-(\rho s + \psi \dot{c}(s))} ds \right], \end{aligned}$$

and, passing to the limit as $\Delta t \rightarrow 0$, it then follows

$$\frac{d\mathbf{E}_{t,Y,m} \left[\dot{V}(t, \dot{Y}(t), \dot{m}(t)) \right]}{dt} = \mathbf{E}_{t,Y,m} \left[e^{-(\rho t + \psi \dot{c}(t))} \right],$$

where, by virtue of (82),

$$e^{-(\rho t + \psi \dot{c}^*(t))} = -r \dot{V}(t, \dot{Y}(t), \dot{m}(t)).$$

Therefore $\mathbf{E}_{t,Y,m} \left[\dot{V}(t, \dot{Y}(t), \dot{m}(t)) \right]$ satisfies the differential equation

$$\frac{d\mathbf{E}_{t,Y,m} \left[\dot{V}(t, \dot{Y}(t), \dot{m}(t)) \right]}{dt} = -r \mathbf{E}_{t,Y,m} \left[\dot{V}(t, \dot{Y}(t), \dot{m}(t)) \right],$$

and the desired transversality condition clearly follows.

Finally, we are only left with the task of proving that $(\dot{q}(t), \dot{q}(t), \dot{c}(t))$ given by (42)-(44) satisfies (68). To this, let's begin by observing that, on account of (39), for an optimal sample path $Y(t)$ of the state vector of the economy, we must have

$$\bar{K}_\Phi^\top \left(\dot{L} + \psi r p k_\Psi^\top \right) Y(t) = 0$$

for every $t \geq 0$. Differentiating the latter and substituting $dY(t)$ with its expression (26) we obtain

$$\bar{K}_\Phi^\top \left(L + \psi r p k_\Psi^\top \right) (\bar{A}Y(t) dt + \bar{Q}^{1/2} dw(t) + \bar{K}_\Phi d\Phi(t)) = 0,$$

which implies

$$\begin{aligned} -\bar{K}_\Phi^\top L \bar{K}_\Phi d\Phi(t) &= \bar{K}_\Phi^\top L \bar{A}Y(t) dt + \psi r \bar{K}_\Phi^\top p k_\Psi^\top (\bar{A}Y(t) dt + \bar{Q}^{1/2} dw(t) \\ &\quad + \bar{K}_\Phi d\Phi(t)) + \bar{K}_\Phi^\top L \bar{Q}^{1/2} dw(t) \\ &= \bar{K}_\Phi^\top L \bar{A}Y(t) dt + \psi r \bar{K}_\Phi^\top p k_\Psi^\top dY(t) + \bar{K}_\Phi^\top L \bar{Q}^{1/2} dw(t). \end{aligned}$$

Hence, taking into account that

$$k_{\Psi}^{\top} dY(t) = d\Psi(t) = \Phi(t) dt = k_{\Phi}^{\top} Y(t) dt$$

we end up with

$$\begin{aligned} d\Phi(t) &= -(\bar{K}_{\Phi}^{\top} L \bar{K}_{\Phi})^{-1} \left(\left(\bar{K}_{\Phi}^{\top} L \bar{A} + \psi r \bar{K}_{\Phi}^{\top} p k_{\Phi}^{\top} \right) Y(t) dt + \bar{K}_{\Phi}^{\top} L \bar{Q}^{1/2} dw(t) \right). \end{aligned}$$

Therefore, from the previous computations, it clearly follows that we have

$$\max_{a(t), q(t), c(t)} \{ \mathcal{L}V(t, Y(t), m(t)) - e^{-(\rho t + \psi c(t))} \} = \mathcal{L}V(t, \dot{Y}(t), \dot{m}(t)) - e^{-(\rho t + \psi \dot{c}(t))},$$

where $(\dot{Y}(t), \dot{m}(t))$ is the solution of (34) and (35) corresponding to the choice of $(\dot{a}(t), \dot{q}(t), \dot{c}(t))$, which in turn satisfies (42)-(44). \square

Proof of Proposition 5

The algebraic computations exploited in the proof are rather heavy. We give here only a sketch, informing the reader that a more detailed version is available from the authors upon request.

We divide our Proof in several steps.

As a first step, setting

$$\Sigma \equiv \begin{pmatrix} \sigma_{1,1} & \sigma_{1,2} & \sigma_{1,3} & \sigma_{1,4} & \sigma_{1,5} \\ \sigma_{1,2} & \sigma_{2,2} & \sigma_{2,3} & \sigma_{2,4} & \sigma_{2,5} \\ \sigma_{1,3} & \sigma_{2,3} & \sigma_{3,3} & \sigma_{3,4} & \sigma_{3,5} \\ \sigma_{1,4} & \sigma_{2,4} & \sigma_{3,4} & \sigma_{4,4} & \sigma_{4,5} \\ \sigma_{1,5} & \sigma_{2,5} & \sigma_{3,5} & \sigma_{4,5} & \sigma_{5,5} \end{pmatrix},$$

we note that, since $\Sigma \hat{M} = 0$ (see Section 3), we must have

$$\sigma_{1,1} = 0, \quad \sigma_{1,2} = 0, \quad \sigma_{1,3} = 0, \quad \sigma_{1,4} = 0, \quad \sigma_{1,5} = 0, \quad (85)$$

and

$$\sigma_{2,4} = -\sigma_{2,3}, \quad \sigma_{3,4} = -\sigma_{3,3}, \quad \sigma_{4,4} = -\sigma_{3,3}, \quad \sigma_{4,5} = -\sigma_{3,5}. \quad (86)$$

In addition, it can be shown that (45) implies

$$\sigma_{2,2} = 0.$$

From the latter, taking into account that Σ is a variance-covariance matrix, it follows

$$\sigma_{2,3} = 0, \quad \sigma_{2,4} = 0, \quad \sigma_{2,5} = 0.$$

Hence, (45) gives rise to the following equations

$$\begin{aligned} & ((\alpha_\Theta + a_4 - a_3)\sigma_{3,3} - a_5\sigma_{3,5})^2 \\ & - 2\sigma_\Theta(\sigma_\Theta + q_2)((\alpha_\Theta + a_4 - a_3)\sigma_{3,3} - a_5\sigma_{3,5}) \\ & + 2\alpha_\Theta(\sigma_\Theta + q_2)^2\sigma_{3,3} = 0, \end{aligned} \quad (87)$$

$$\begin{aligned} & ((\alpha_\Theta + a_4 - a_3)\sigma_{3,3} - a_5\sigma_{3,5})((\alpha_\Theta + a_4 - a_3)\sigma_{3,5} - a_5\sigma_{5,5}) \\ & - \sigma_\Theta(\sigma_\Theta + q_2)((\alpha_\Theta + a_4 - a_3)\sigma_{3,5} - a_5\sigma_{5,5}) \\ & + (\sigma_\Theta + q_2)^2(\sigma_{3,3} + \alpha_\Theta\sigma_{3,5}) = 0, \end{aligned} \quad (88)$$

$$((\alpha_\Theta + a_4 - a_3)\sigma_{3,5} - a_5\sigma_{5,5})^2 + 2(\sigma_\Theta + q_2)^2\sigma_{3,5} = 0. \quad (89)$$

As a second step, by virtue of (46), we can prove that

$$\bar{K}_\Phi^\top L\bar{Q}^{1/2} = 0.$$

This implies

$$q_1 = q_2 = 0. \quad (90)$$

Moreover, it can be shown that

$$\bar{K}_\Phi^\top p = 0,$$

which gives

$$(a^\top, b^\top) = -(\bar{K}_\Phi^\top L\bar{K}_\Phi)^{-1}\bar{K}_\Phi^\top L\bar{A}.$$

On the other hand, computing $\bar{K}_\Phi^\top L\bar{K}_\Phi$ and $\bar{K}_\Phi^\top L\bar{A}$ on account of (46), we obtain

$$a_1 = a_2 = a_3 = a_4 = a_5 = b_1 = b_2 = b_3 = b_5 = 0, \quad (91)$$

and

$$b_4 = -(\ell_{4,9} - \ell_{8,9} + \ell_{9,9})^{-1}(\ell_{4,9} + \ell_{8,9}). \quad (92)$$

As a third step, we note that, thanks to (90) and (91), Equation (87) yields

$$\sigma_{3,3} = 0$$

and, on account of the nature of Σ ,

$$\sigma_{3,4} = 0, \quad \sigma_{3,5} = 0.$$

As a consequence, (88) and (89) are satisfied for any positive value of $\sigma_{5,5}$. Therefore, we obtain (49).

As a fourth step, it remains to prove that $b_4 = 0$ and (50) holds true. To this task, by a straightforward computation, it follows that (47) gives rise to the following system of quadratic equations

$$\begin{aligned} r(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6}) + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})^2 \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})^2 - 4(\ell_{1,2} + \ell_{2,6}) = 0, \end{aligned} \quad (93)$$

$$\begin{aligned} r(\ell_{1,2} + \ell_{2,6}) + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,2} + \ell_{2,6}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{2,3} - \ell_{2,4}) - 2\ell_{2,2} = 0, \end{aligned} \quad (94)$$

$$\begin{aligned} r(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9}) \\ + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9}) \\ - 2(\ell_{2,3} + \ell_{2,9}) - (b_4 + \alpha_\Theta)(\ell_{1,8} + \ell_{6,8}) + \alpha_\Theta(\ell_{1,3} + \ell_{3,6}) = 0, \end{aligned} \quad (95)$$

$$\begin{aligned} r(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9}) \\ + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,4} - \ell_{4,4} + \ell_{3,9} - \ell_{4,9}) \\ - (b_4 + \alpha_\Theta)(\ell_{1,8} + \ell_{6,8}) - 2(\ell_{2,4} + \ell_{2,9}) = 0, \end{aligned} \quad (96)$$

$$\begin{aligned} r(\ell_{1,5} + \ell_{5,6}) + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\psi + \ell_{1,5} + \ell_{5,6}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,5} - \ell_{4,5}) \\ - (\ell_{1,4} + \ell_{4,6}) - 2\ell_{2,5} = 0, \end{aligned} \quad (97)$$

$$\begin{aligned} r(\ell_{1,7} + \ell_{6,7}) + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,7} + \ell_{6,7}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,7} - \ell_{4,7}) - 2\ell_{2,7} = 0, \end{aligned} \quad (98)$$

$$\begin{aligned} r(\ell_{1,8} - \ell_{1,9} + \ell_{6,8} - \ell_{6,9}) \\ + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,8} - \ell_{1,9} + \ell_{6,8} - \ell_{6,9}) \\ + \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,8} - \ell_{3,9} - \ell_{4,8} + \ell_{4,9}) \\ + 2(b_4 + \alpha_\Theta)(\ell_{1,8} + \ell_{6,8}) + \alpha_\Theta(\ell_{1,3} + \ell_{3,6}) - 2(\ell_{2,8} - \ell_{2,9}) = 0, \end{aligned} \quad (99)$$

$$\begin{aligned}
r(\ell_{1,10} + \ell_{6,10}) + \sigma_D^2(\ell_{1,1} + 2\ell_{1,6} + \ell_{6,6})(\ell_{1,10} + \ell_{6,10}) & (100) \\
+ \sigma_\Theta^2(\ell_{1,3} - \ell_{1,4} + \ell_{3,6} - \ell_{4,6})(\ell_{3,10} - \ell_{4,10}) \\
- (\ell_{1,9} + \ell_{6,9}) - 2\ell_{2,10} = 0,
\end{aligned}$$

$$r\ell_{2,2} + \sigma_D^2(\ell_{1,2} + \ell_{2,6})^2 + \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})^2 = 0, \quad (101)$$

$$\begin{aligned}
r(\ell_{2,3} + \ell_{2,9}) + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\ell_{1,3} + \ell_{3,6} + \ell_{1,9} + \ell_{6,9}) & (102) \\
+ \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9}) \\
- (b_4 + \alpha_\Theta)\ell_{2,8} + \alpha_\Theta\ell_{2,3} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{2,4} + \ell_{2,9}) + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9}) & (103) \\
+ \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9}) - (b_4 + \alpha_\Theta)\ell_{2,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r\ell_{2,5} + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\psi + \ell_{1,5} + \ell_{5,6}) & (104) \\
+ \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,5} - \ell_{4,5}) - \ell_{2,4} = 0,
\end{aligned}$$

$$r\ell_{2,7} + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\ell_{1,7} + \ell_{6,7}) + \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,7} - \ell_{4,7}) = 0, \quad (105)$$

$$\begin{aligned}
r(\ell_{2,8} - \ell_{2,9}) + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\ell_{1,8} - \ell_{1,9} + \ell_{6,8} - \ell_{6,9}) & (106) \\
+ \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,8} - \ell_{3,9} - \ell_{4,8} + \ell_{4,9}) \\
+ \alpha_\Theta\ell_{2,3} + 2(b_4 + \alpha_\Theta)\ell_{2,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r\ell_{2,10} + \sigma_D^2(\ell_{1,2} + \ell_{2,6})(\ell_{1,10} + \ell_{6,10}) & (107) \\
+ \sigma_\Theta^2(\ell_{2,3} - \ell_{2,4})(\ell_{3,10} - \ell_{4,10}) - \ell_{2,9} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,3} + 2\ell_{3,9} + \ell_{9,9}) + \sigma_D^2(\ell_{1,3} + \ell_{3,6} + \ell_{1,9} + \ell_{6,9})^2 & (108) \\
+ \sigma_\Theta^2(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9})^2 \\
- 2(b_4 + \alpha_\Theta)(\ell_{3,8} + \ell_{3,9}) + 2\alpha_\Theta(\ell_{3,3} + \ell_{3,9}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,4} + \ell_{3,9} + \ell_{4,9} + \ell_{9,9}) & (109) \\
+ \sigma_D^2(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9})(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9}) \\
+ \sigma_\Theta^2(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9})(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9}) \\
+ \alpha_\Theta(\ell_{3,4} + \ell_{3,9}) - (b_4 + \alpha_\Theta)(\ell_{3,8} + 2\ell_{8,9} + \ell_{4,8}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,5} + \ell_{5,9}) + \sigma_D^2(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9})(\psi + \ell_{1,5} + \ell_{5,6}) & \quad (110) \\
+ \sigma_\Theta^2(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9})(\ell_{3,5} - \ell_{4,5}) \\
+ \alpha_\Theta \ell_{3,5} - (\ell_{3,4} + \ell_{4,9}) - (b_4 + \alpha_\Theta) \ell_{5,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,7} + \ell_{7,9}) + \sigma_D^2(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9})(\ell_{1,7} + \ell_{6,7}) & \quad (111) \\
+ \sigma_\Theta^2(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9})(\ell_{3,7} - \ell_{4,7}) \\
+ \alpha_\Theta \ell_{3,7} - (b_4 + \alpha_\Theta) \ell_{7,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,8} - \ell_{3,9} + \ell_{8,9} - \ell_{9,9}) & \quad (112) \\
+ \sigma_D^2(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9})(\ell_{1,8} - \ell_{1,9} + \ell_{6,8} - \ell_{6,9}) \\
+ \sigma_\Theta^2(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9})(\ell_{3,8} - \ell_{3,9} - \ell_{4,8} + \ell_{4,9}) \\
+ \alpha_\Theta(\ell_{3,3} + \ell_{3,8}) + (b_4 + \alpha_\Theta)(2\ell_{3,8} + 3\ell_{8,9} - \ell_{8,8}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{3,10} + \ell_{9,10}) + \sigma_D^2(\ell_{1,3} + \ell_{1,9} + \ell_{3,6} + \ell_{6,9})(\ell_{1,10} + \ell_{6,10}) & \quad (113) \\
+ \sigma_\Theta^2(\ell_{3,10} - \ell_{4,10})(\ell_{3,3} - \ell_{3,4} + \ell_{3,9} - \ell_{4,9}) \\
- (\ell_{3,9} + \ell_{9,9}) + \alpha_\Theta \ell_{3,10} - (b_4 + \alpha_\Theta) \ell_{8,10} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{4,4} + 2\ell_{4,9} + \ell_{9,9}) + \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})^2 & \quad (114) \\
+ \sigma_\Theta^2(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9})^2 \\
- 2(b_4 + \alpha_\Theta)(\ell_{4,8} + \ell_{8,9}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{4,5} + \ell_{5,9}) + \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})(\psi + \ell_{1,5} + \ell_{5,6}) & \quad (115) \\
+ \sigma_\Theta^2(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9})(\ell_{3,5} - \ell_{4,5}) \\
- (\ell_{4,4} + \ell_{4,9}) - (b_4 + \alpha_\Theta) \ell_{5,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{4,7} + \ell_{7,9}) + \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})(\ell_{1,7} + \ell_{6,7}) & \quad (116) \\
+ \sigma_\Theta^2(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9})(\ell_{3,7} - \ell_{4,7}) \\
- (b_4 + \alpha_\Theta) \ell_{7,8} = 0,
\end{aligned}$$

$$\begin{aligned}
r(\ell_{4,8} - \ell_{4,9} + \ell_{8,9} - \ell_{9,9}) & \quad (117) \\
+ \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})(\ell_{1,8} - \ell_{1,9} + \ell_{6,8} - \ell_{6,9}) \\
+ \sigma_\Theta^2(\ell_{3,4} + \ell_{3,9} - \ell_{4,4} - \ell_{4,9})(\ell_{3,8} - \ell_{3,9} - \ell_{4,8} + \ell_{4,9}) \\
+ \alpha_\Theta(\ell_{3,4} + \ell_{3,9}) + (b_4 + \alpha_\Theta)(2\ell_{4,8} + 3\ell_{8,9} - \ell_{8,8}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(l_{4,10} + l_{9,10}) + \sigma_D^2(l_{1,4} + l_{1,9} + l_{4,6} + l_{6,9})(l_{1,10} + l_{6,10}) & (118) \\
+ \sigma_\Theta^2(l_{3,4} + l_{3,9} - l_{4,4} - l_{4,9})(l_{3,10} - l_{4,10}) \\
- (l_{4,9} + l_{9,9}) - (b_4 + \alpha_\Theta)l_{8,10} = 0,
\end{aligned}$$

$$rl_{5,5} + \sigma_D^2(\psi + l_{1,5} + l_{5,6})^2 + \sigma_\Theta^2(l_{3,5} - l_{4,5})^2 - 2l_{4,5} = 0, \quad (119)$$

$$\begin{aligned}
rl_{5,7} + \sigma_D^2(\psi + l_{1,5} + l_{5,6})(l_{1,7} + l_{6,7}) & (120) \\
+ \sigma_\Theta^2(l_{3,5} - l_{4,5})(l_{3,7} - l_{4,7}) - l_{4,7} - \psi = 0,
\end{aligned}$$

$$\begin{aligned}
r(l_{5,8} - l_{5,9}) + \sigma_D^2(\psi + l_{1,5} + l_{5,6})(l_{1,8} - l_{1,9} + l_{6,8} - l_{6,9}) & (121) \\
+ \sigma_\Theta^2(l_{3,5} - l_{4,5})(l_{3,8} - l_{3,9} - l_{4,8} + l_{4,9}) \\
+ \alpha_\Theta l_{3,5} - (l_{4,8} - l_{4,9}) + 2(b_4 + \alpha_\Theta)l_{5,8} = 0,
\end{aligned}$$

$$\begin{aligned}
rl_{5,10} + \sigma_D^2(\psi + l_{1,5} + l_{5,6})(l_{1,10} + l_{6,10}) & (122) \\
+ \sigma_\Theta^2(l_{3,5} - l_{4,5})(l_{3,10} - l_{4,10}) - (l_{5,9} + l_{4,10}) = 0,
\end{aligned}$$

$$rl_{7,7} + \sigma_D^2(l_{1,7} + l_{6,7})^2 + \sigma_\Theta^2(l_{3,7} - l_{4,7})^2 = 0, \quad (123)$$

$$\begin{aligned}
r(l_{7,8} - l_{7,9}) + \sigma_D^2(l_{1,7} + l_{6,7})(l_{1,8} - l_{1,9} + l_{6,8} - l_{6,9}) & (124) \\
+ \sigma_\Theta^2(l_{3,7} - l_{4,7})(l_{3,8} - l_{3,9} - l_{4,8} + l_{4,9}) \\
+ \alpha_\Theta l_{3,7} + 2(b_4 + \alpha_\Theta)l_{7,8} = 0,
\end{aligned}$$

$$\begin{aligned}
rl_{7,10} + \sigma_D^2(l_{1,7} + l_{6,7})(l_{1,10} + l_{6,10}) & \\
+ \sigma_\Theta^2(l_{3,7} - l_{4,7})(l_{3,10} - l_{4,10}) - l_{7,9} = 0, & (125)
\end{aligned}$$

$$\begin{aligned}
r(l_{8,8} - 2l_{8,9} + l_{9,9}) + \sigma_D^2(l_{1,8} - l_{1,9} + l_{6,8} - l_{6,9})^2 & (126) \\
+ \sigma_\Theta^2(l_{3,8} - l_{3,9} - l_{4,8} + l_{4,9})^2 \\
+ 2\alpha_\Theta(l_{3,8} - l_{3,9}) + 4(b_4 + \alpha_\Theta)(l_{8,8} - l_{8,9}) = 0,
\end{aligned}$$

$$\begin{aligned}
r(l_{8,10} - l_{9,10}) + \sigma_D^2(l_{1,8} - l_{1,9} + l_{6,8} - l_{6,9})(l_{1,10} + l_{6,10}) & (127) \\
+ \sigma_\Theta^2(l_{3,8} - l_{3,9} - l_{4,8} + l_{4,9})(l_{3,10} - l_{4,10}) \\
+ \alpha_\Theta l_{3,10} - (l_{8,9} - l_{9,9}) + 2(b_4 + \alpha_\Theta)l_{8,10} = 0,
\end{aligned}$$

$$r\ell_{10,10} + \sigma_D^2(\ell_{1,10} + \ell_{6,10})^2 + \sigma_\Theta^2(\ell_{3,10} - \ell_{4,10})^2 - 2\ell_{9,10} = 0. \quad (128)$$

In addition, (46) yields the linear system

$$\ell_{1,4} + \ell_{1,8} + \ell_{4,6} + \ell_{6,8} = 0, \quad (129)$$

$$\ell_{2,4} + \ell_{2,8} = 0, \quad (130)$$

$$\ell_{3,4} + \ell_{3,8} + \ell_{4,9} + \ell_{8,9} = 0, \quad (131)$$

$$\ell_{4,4} + \ell_{4,8} + \ell_{4,9} + \ell_{8,9} = 0, \quad (132)$$

$$\ell_{4,5} + \ell_{5,8} = 0, \quad (133)$$

$$\ell_{4,7} + \ell_{7,8} = 0, \quad (134)$$

$$\ell_{4,8} - \ell_{4,9} - \ell_{8,9} + \ell_{8,8} = 0, \quad (135)$$

$$\ell_{4,10} + \ell_{8,10} = 0, \quad (136)$$

and, from (92), it follows that we must have

$$\ell_{4,9} - \ell_{8,9} + \ell_{9,9} \neq 0 \quad (137)$$

and

$$(1 + b_4)\ell_{4,9} + (1 - b_4)\ell_{8,9} + b_4\ell_{9,9} = 0. \quad (138)$$

Hence, combining (129)-(136) with (138), we obtain

$$\ell_{3,4} = -\ell_{3,8} - \frac{b_4}{1 + b_4}(2\ell_{8,9} - \ell_{9,9}), \quad (139)$$

$$\ell_{4,4} = \ell_{8,8} - \frac{2b_4}{1 + b_4}(2\ell_{8,9} - \ell_{9,9}), \quad (140)$$

$$\ell_{4,8} = -\ell_{8,8} + \frac{b_4}{1 + b_4}(2\ell_{8,9} - \ell_{9,9}), \quad (141)$$

$$\ell_{4,9} = -\ell_{8,9} + \frac{b_4}{1 + b_4}(2\ell_{8,9} - \ell_{9,9}), \quad (142)$$

$$\ell_{2,8} = -\ell_{2,4}, \quad (143)$$

$$\ell_{5,8} = -\ell_{4,5}, \quad (144)$$

$$\ell_{7,8} = -\ell_{4,7}, \quad (145)$$

$$\ell_{8,10} = -\ell_{4,10}, \quad (146)$$

$$\ell_{1,8} = -(\ell_{1,4} + \ell_{4,6} + \ell_{6,8}), \quad (147)$$

and

$$\ell_{4,9} - \ell_{8,9} + \ell_{9,9} = -\frac{1}{1 + b_4}(2\ell_{8,9} - \ell_{9,9}). \quad (148)$$

Now, substituting (139)-(147) in (114), (117), and (126), we obtain

$$\begin{aligned}
& r(\ell_{8,8} - 2\ell_{8,9} + \ell_{9,9}) \quad (149) \\
& + \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})^2 + \sigma_\Theta^2(\ell_{3,8} - \ell_{3,9} + \ell_{8,9} - \ell_{9,9})^2 \\
& - 2(b_4 + \alpha_\Theta)(\ell_{8,9} - \ell_{8,8} + \frac{b_4}{1+b_4}(2\ell_{8,9} - \ell_{9,9})) = 0,
\end{aligned}$$

$$\begin{aligned}
& -r(\ell_{8,8} - 2\ell_{8,9} + \ell_{9,9}) \quad (150) \\
& - \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})^2 - \sigma_\Theta^2(\ell_{3,8} - \ell_{3,9} + \ell_{8,8} - \ell_{8,9})^2 \\
& + \frac{(2b_4 + \alpha_\Theta)b_4}{1+b_4}(2\ell_{8,9} - \ell_{9,9}) \\
& + 3(b_4 + \alpha_\Theta)(\ell_{8,9} - \ell_{8,8}) - \alpha_\Theta(\ell_{3,8} - \ell_{3,9}) = 0,
\end{aligned}$$

$$\begin{aligned}
& r(\ell_{8,8} - 2\ell_{8,9} + \ell_{9,9}) - \sigma_D^2(\ell_{1,4} + \ell_{1,9} + \ell_{4,6} + \ell_{6,9})^2 \quad (151) \\
& + \sigma_\Theta^2(\ell_{3,8} - \ell_{3,9} + \ell_{8,8} - \ell_{8,9})^2 \\
& + 2\alpha_\Theta(\ell_{3,8} - \ell_{3,9}) + 4(b_4 + \alpha_\Theta)(\ell_{8,8} - \ell_{8,9}) = 0,
\end{aligned}$$

Hence, summing term by term (149) and (150), we have

$$\begin{aligned}
& (b_4 + \alpha_\Theta)(\ell_{8,9} - \ell_{8,8}) - \alpha_\Theta(\ell_{3,8} - \ell_{3,9}) \quad (152) \\
& - \frac{\alpha_\Theta b_4}{1+b_4}(2\ell_{8,9} - \ell_{9,9}) = 0.
\end{aligned}$$

and subtracting term by term (149) and (151), it follows

$$-2(b_4 + \alpha_\Theta)(\ell_{8,8} - \ell_{8,9}) - 2\alpha_\Theta(\ell_{3,8} - \ell_{3,9}) \quad (153)$$

$$-2(b_4 + \alpha_\Theta) \frac{b_4}{1+b_4}(2\ell_{8,9} - \ell_{9,9}) = 0. \quad (154)$$

Therefore, combining (152) and (153), we obtain

$$\frac{2b_4^2}{1+b_4}(2\ell_{8,9} - \ell_{9,9}) = 0.$$

This, by virtue of (137) and (148), implies

$$b_4 = 0. \quad (155)$$

Finally, on account of what shown above, it is possible to obtain (50). \square

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Anno: 1997

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Elenco dei report pubblicati

Anno: 2000

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Anno: 2001

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Elenco dei report pubblicati

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Anno: 2003

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