

Insider trading with penalties in continuous time

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Research Seminar @ Institute of Mathematics and Statistics
WU Vienna
15 May 2024

Outline of talk

- 1 The Kyle model without frictions
- 2 Kyle model with penalties
- 3 Emergence of Schrödinger potentials
- 4 Equilibrium for the Kyle-Back model with penalties
- 5 Regulating insider trading
- 6 Conclusion

The Kyle model without frictions

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The Kyle model without frictions

Kyle-Back model of informed trading

Inspired by Kyle (1985), Back (1992) studies a market for a bond and a risky asset with three types of participants:

- 1** *Noise traders*: The noise traders have no private information and are not strategic. Their cumulative demand is given by σB , where B is a Brownian motion and σ is constant.
- 2** *Informed trader, a.k.a. insider*: The insider knows the value of the risky asset at time 1, which is given by a random variable, V , independent of B . Being risk-neutral, her objective is to maximize her expected profit.
- 3** *Market makers*: The market makers observe the total order and set the price of the risky asset to clear the market via a Bertrand competition.

The pricing mechanism of the market

- Market makers decide the price by looking at the total order

$$Y_t = \sigma B_t + \theta_t,$$

where θ_t is the position of the insider in the risky asset at time t .

- Thus, the filtration of the market maker is the one generated by Y . Note that θ is not necessarily adapted to \mathcal{F}^Y .
- The market makers have a *pricing rule*, $H : [0, 1] \times \mathbb{R} \mapsto \mathbb{R}$, to assign the price in the following form:

$$S_t = H(t, Y_t),$$

where S_t is the market price of the risky asset at time t . H is strictly increasing in Y .

- The market makers choose a *rational* pricing rule, i.e. a pricing rule so that

$$H(t, Y_t) = \mathbb{E}[V | \mathcal{F}_t^Y],$$

for every $t \in [0, 1]$.

- The insider aims to maximize her expected profit out of trading.
- Equilibrium is a pair (H^*, θ^*) such that the following conditions are satisfied:
 - 1 *Market efficiency*: Given θ^* , H^* is a rational pricing rule.
 - 2 *Insider optimality*: Given H^* , θ^* maximizes the expected profit of the insider.

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Insider trading with legal penalties

- Recently Carré, Collin-Dufresne and Gabriel (2022, JET) and Kacperczyk and Pagnotta (2023, forthcoming in JoF) study in a one-period model the Kyle equilibrium when the insider is subject to additional transaction costs (or legal risk).

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- These additional costs could be a result of frictions in executing large portfolios, or
- The penalties arise in case the informed trader is an illegal insider and pays a penalty (in addition to losing all her profits) after investigation.
- Carré et al. consider general convex penalties and uniformly distributed noise trades while Kacperczyk and Pagnotta have quadratic penalties and normally distributed noise demand.

Penalties in continuous time

- Let's associate the following quadratic transaction cost by time t to the strategy $d\theta_t = \alpha_t dt$:

$$C_t := \frac{c}{2} \int_0^t \alpha_s^2 ds.$$

for some $c > 0$.

- Her objective is still to maximize the expected final wealth, W_1 , given by

$$W_1 = \int_0^1 (V - S_s) \alpha_s ds - \frac{c}{2} \int_0^1 \alpha_s^2 ds.$$

Legal penalties in continuous time

- The above cost structure can also arise as a legal penalty.
- Indeed, suppose an investigation identifies illegal inside trading with probability p , after which the insider pays a legal penalty of $k \int_0^1 \alpha_t^2 dt$.

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- Indeed, suppose an investigation identifies illegal inside trading with probability p , after which the insider pays a legal penalty of $k \int_0^1 \alpha_t^2 dt$.
- The expected profit of the insider under this scenario is

$$\begin{aligned} E^v \left[(1-p) \int_0^1 (V - S_s) \alpha_s ds - pk \int_0^1 \alpha_s^2 ds \right] \\ = (1-p) E^v \left[\int_0^1 (V - S_s) \alpha_s ds - \frac{pk}{1-p} \int_0^1 \alpha_s^2 ds \right]. \end{aligned}$$

Thus the coefficient c from the previous slide can be associated with $\frac{pk}{1-p}$, which gets large as the probability of a successful investigation gets bigger.

Insider's optimality

Recall $dY_t = \sigma dB_t + \alpha_t dt$, and let

$$J(t, y) = \sup_{\alpha \in \mathcal{A}(H)} E^v \left[\int_t^1 (v - H(u, Y_u)) \alpha_u du - \frac{c}{2} \int_t^1 \alpha_t^2 dt \mid Y_t = y \right].$$

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Direct calculations lead to

$$J_t + \frac{\sigma^2}{2} J_{yy} + \sup_{\alpha} \left\{ \alpha (J_y + v - H) - \frac{c\alpha^2}{2} \right\} = 0.$$

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Therefore,

$$\alpha^* = \frac{J_y(t, y) + v - H(t, y)}{c}. \quad (1)$$

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and

$$J_t + \frac{\sigma^2}{2} J_{yy} + \frac{(J_y + v - H)^2}{2c} = 0. \quad (2)$$

A quadratic BSDE

Next suppose there exists a smooth function J^0 such that

$$J_t^0 + \frac{\sigma^2}{2} J_{yy}^0 = 0, \quad J_y^0 = H - v. \quad (3)$$

Thus, if one defines $u = J - J^0$ and conjectures that $J(1, \cdot) \equiv 0$, one obtains

$$u_t + \frac{1}{2} \sigma^2 u_{yy} + \frac{1}{2c} u_y^2 = 0, \quad u(1, y) = -j^0(y, v) := -J^0(1, y). \quad (4)$$

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This has a simple (BSDE) formulation

$$dU_t = \sigma Z_t dB_t - \frac{1}{2c} Z_t^2 dt, \quad U_1 = u(1, \sigma B_1), \quad (5)$$

whose solution is given by $u(t, x) = c\sigma^2 \log \rho(t, x, v)$, where

$$\rho(t, x, v) := E^v \left[\exp \left(- \frac{j^0(\sigma B_1, v)}{c\sigma^2} \right) \middle| \sigma B_t = x \right]. \quad (6)$$

Towards a recipe for equilibrium

- The above implies the optimal control of the insider is $\alpha^*(t, Y_t, V)$, where

$$\alpha^*(t, y, v) := \sigma^2 \frac{\rho_y(t, y, v)}{\rho(t, y, v)}, \quad (7)$$

implying an h -transformation in making!

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$$\begin{aligned} \mathbb{E} \left[\frac{\rho_y(t, Y_t, V)}{\rho(t, Y_t, V)} \middle| \mathcal{F}_t^Y \right] &= \int \rho_y(t, Y_t, v) \Pi(dv) \\ &= \frac{d}{dy} \int \rho(t, y, v) \Pi(dv) \Big|_{y=Y_t} = 0, \end{aligned}$$

Proposition 1

Suppose that there exists a continuous function j^0 such that $\rho(0, 0, \cdot) \equiv 1$, where ρ is defined via (6). Assume further that there exists a unique strong solution on $(\Omega, \mathcal{G}, (\mathcal{G}_t)_{t \in [0, 1]}, \mathbb{Q})$ to

$$Y_t = \sigma B_t + \int_0^t \sigma^2 \frac{\rho_y(s, Y_s, V)}{\rho(s, Y_s, V)} ds$$

such that

$$\mathbb{E}^{\mathbb{Q}} \left[\int_0^t \left(\frac{\rho_y(s, Y_s, V)}{\rho(s, Y_s, V)} \right)^2 ds \right] < \infty, \quad t \in [0, 1]. \quad (8)$$

Then,

$$\rho(t, Y_t, v) \Pi(dv) = \mathbb{P}(V \in dv | \mathcal{F}_t^Y), \quad t \in [0, 1].$$

A recipe for equilibrium

- 1** Find a continuous function j^0 such that i) $\rho(0, 0, \cdot) \equiv 1$, where ρ is given by (6), and ii) it is differentiable in its first parameter with $j_y^0(y, v) = h(y) - v$. Note that the first condition entails

$$\int_{\mathbb{R}} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{y^2}{2\sigma^2}\right) \exp\left(-\frac{j^0(y, v)}{\sigma^2 c}\right) dy = 1, \quad \forall v. \quad (9)$$

- 2** Set

$$H_t + \frac{1}{2}\sigma^2 H_{yy} = 0, \quad H(1, y) = h(y).$$

- 3** Show that (H, θ^*) with $d\theta_t^* = \sigma^2 \frac{\rho_y(t, Y_t, V)}{\rho(t, Y_t, V)} dt$ is equilibrium provided they are admissible by using the candidate value function $J = J^0 + u$ that satisfies (2).

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Towards a fixed-point algorithm

- **Notation:** $\hat{c} := c\sigma^2$. That $j_y^0 = h(y) - v$ for some h to be determined implies we are searching for a j^0 such that

$$j^0(y, v) = \Psi(v) + \phi(y) - yv. \quad (10)$$

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- Next, since $\exp(-j^0(y, v)/\hat{c})\Pi(dv)$ is the conditional distribution of V given $Y_1 = y$, it must integrate to 1:

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- Moreover, (9) is equivalent to

$$\int_{\mathbb{R}} \exp\left(\frac{yv - \phi(y)}{\hat{c}}\right) p(\sigma, y) dy = \exp\left(\frac{\Psi(v)}{\hat{c}}\right),$$

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where $p(\sigma, \cdot)$ is the density of $N(0, \sigma^2)$.

- Thus, if they exist, ϕ/\hat{c} and Ψ/\hat{c} are the *unique* potentials of the entropic optimal transport problem from $N(0, \sigma^2)$ to Π with the quadratic “cost function” $\frac{1}{2\hat{c}}(x - y)^2$.

Existence of solutions

- Given the positivity of the cost function, the associated entropic optimal transport problem has a solution (see lecture notes by M. Nutz) with the potentials ϕ and Ψ solving the *Schrödinger equations*

$$\begin{aligned}\phi(y) &= \hat{c} \log \int_{f(\mathbb{R})} \exp\left(\frac{yv - \Psi(v)}{\hat{c}}\right) \Pi(dv), \\ \Psi(v) &= \hat{c} \log \int_{\mathbb{R}} \exp\left(\frac{yv - \phi(y)}{\hat{c}}\right) p(\sigma, y) dy.\end{aligned}\tag{11}$$

- Moreover, $\mathbb{E}[|\phi(\sigma B_1)| + |\Psi(V)|] < \infty$.

Differentiability of solutions

- Assume that V has all exponential moments. Then it is a simple exercise to show that ϕ and Ψ are infinitely differentiable.
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- In particular,

$$\phi'(y) = \int v \nu_{\Psi}(dv), \quad \phi''(y) = \frac{1}{\hat{c}} \left(\int v^2 \nu_{\Psi}(dv) - \left(\int v \nu_{\Psi}(dv) \right)^2 \right), \quad (12)$$

where

$$\nu_{\Psi}(dv) = \frac{\exp\left(\frac{yv - \Psi(v)}{\hat{c}}\right) \Pi(dv)}{\int \exp\left(\frac{yv - \Psi(v)}{\hat{c}}\right) \Pi(dv)}.$$

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- Note: ν_{Ψ} will be the conditional distribution of V given Y_1 in equilibrium!

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- $\phi^*(y) = \frac{\lambda^* y^2}{2} + \mu y$ for

$$\lambda^* = \frac{-c + \sqrt{c^2 + 4 \frac{\gamma^2}{\sigma^2}}}{2}. \quad (13)$$

Moreover,

$$\psi^*(v) = \frac{\hat{c}}{2} \log \frac{c}{c + \lambda^*} + \frac{(\mu - v)^2}{2(c + \lambda^*)}. \quad (14)$$

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Equilibrium for the Kyle-Back model with penalties

- Now, denote the solution of (11) by (ϕ^*, Ψ^*) with the normalisation that $\phi^*(0) = 0$.
- Next define

$$H^*(t, y) = \int \rho^*(t, y, z) z \Pi(dz), \quad (15)$$

where

$$\rho^*(t, y, \nu) := E^\nu \left[\exp \left(- \frac{j^*(\sigma B_1, \nu)}{\hat{c}} \right) \middle| \sigma B_t = y \right]. \quad (16)$$

- Observe that

$$H_t^* + \frac{\sigma^2}{2} H_{yy}^* = 0, \quad H^*(1, \cdot) = h^* = \frac{d\phi^*}{dy}. \quad (17)$$

- Then (H^*, θ^*) is an equilibrium where

$$d\theta_t^* = \sigma^2 \frac{\rho_y^*(t, Y_t, V)}{\rho^*(t, Y_t, V)} dt, t \in [0, 1], \text{ and } \theta_0^* = 0. \quad (18)$$

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$$E^{0,v}[W_1^{\theta^*}] = J(0, 0) = \Psi^*(v) + E^{0,v}[\phi^*(\sigma B_1)]. \quad (19)$$

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- The *price inefficiency* of the equilibrium, denoted by δ , is given by

$$\delta := \mathbb{E}[\text{Var}(V | \mathcal{F}_1^{Y^*})] = \hat{c} \mathbb{E}\left[\frac{d^2 \phi^*}{dy^2}(Y_1^*)\right]. \quad (20)$$

Alternative representation of the optimal strategy

- Suppose $H^*(1, \cdot)$ is of at most exponential growth. Then,

$$\alpha_t^* := \frac{d\theta_t^*}{dt} = \frac{1}{c}(v - E^{0,v}[h^*(Y_1^*)|\mathcal{F}_t^I]), \quad (21)$$

with $E^{0,v}[h(Y_1^*)|\mathcal{F}_t^I] = \mathcal{P}(t, Y_t^*; v)$, where

$$\mathcal{P}(t, y; v) = \frac{\int_{\mathbb{R}} h^*(x) \exp\left(\frac{vx - \phi^*(x)}{\hat{c}}\right) p(\sigma\sqrt{1-t}, x-y) dx}{\int_{\mathbb{R}} \exp\left(\frac{vx - \phi^*(x)}{\hat{c}}\right) p(\sigma\sqrt{1-t}, x-y) dx}.$$

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- Thus, insider trades on the differential between her private signal and the expected terminal price, and trades aggressively if the penalty, i.e. c , is small.

Connection with h -transforms

- Consider the SDE associated to the equilibrium demand:

$$Y_t^* = \sigma B_t + \sigma^2 \int_0^t \frac{\rho_y^*(s, Y_s^*, v)}{\rho^*(s, Y_s, v)} ds. \quad (22)$$

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- Let $(\mathcal{B}_t)_{t \in [0, 1]}$ be the right continuous augmentation of the natural filtration of the coordinate process X , and \mathbb{W} be the Wiener measure.
- Then \mathbb{Q}^* is given by the following h -transform of Brownian motion:

$$\mathbb{E}^{\mathbb{Q}^*}[F] = \frac{\mathbb{E}^{\mathbb{W}} \left[F \exp \left(\frac{v\sigma X_1 - \phi^*(\sigma X_1)}{\hat{c}} \right) \right]}{\mathbb{E}^{\mathbb{W}} \left[\exp \left(\frac{v\sigma X_1 - \phi^*(\sigma X_1)}{\hat{c}} \right) \right]}, \quad F \in \mathcal{B}_1. \quad (23)$$

Relative entropy of the h -transform

- The relative entropy of the above change of measure is not only finite but can be computed explicitly.
- Indeed,

$$H(\mathbb{Q}^* || \mathbb{W}) = \frac{v(\Psi^*)'(v) - \Psi^*(v) - E^{0,v}[\phi^*(Y_1^*)]}{\hat{c}}$$

- Consequently, the insider expects to the following penalty in equilibrium:

$$v(\Psi^*)'(v) - \Psi^*(v) - E^{0,v}[\phi^*(Y_1^*)].$$

Back to the Gaussian case

- For a better parametrisation suppose $c = \kappa \frac{\gamma}{\sigma}$.
- Then

$$dY_t^* = \sigma dB_t + \Lambda(\kappa) \sigma \frac{\frac{V - \mu}{\sigma} - \frac{\Lambda(\kappa) Y_t^*}{\sigma}}{1 - t \Lambda^2(\kappa)} dt,$$

where $\Lambda(\kappa) = \frac{\sqrt{\kappa^2 + 4} - \kappa}{2}$.

- $\lambda^* = \Lambda(\kappa) \frac{\gamma}{\sigma}$.

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- Then

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where $\Lambda(\kappa) = \frac{\sqrt{\kappa^2 + 4} - \kappa}{2}$.

- $\lambda^* = \Lambda(\kappa) \frac{\gamma}{\sigma}$.
- Y^* is a martingale for market makers, but it is an O-U process mean reverting around $\frac{(V-\mu)\sigma}{\gamma\Lambda(\kappa)}$ for the insider.

Back to the Gaussian case

- For a better parametrisation suppose $c = \kappa \frac{\gamma}{\sigma}$.
- Then

$$dY_t^* = \sigma dB_t + \Lambda(\kappa) \sigma \frac{V - \mu - \frac{\Lambda(\kappa) Y_t^*}{\sigma}}{1 - t \Lambda^2(\kappa)} dt,$$

where $\Lambda(\kappa) = \frac{\sqrt{\kappa^2 + 4} - \kappa}{2}$.

- $\lambda^* = \Lambda(\kappa) \frac{\gamma}{\sigma}$.
- Y^* is a martingale for market makers, but it is an O-U process mean reverting around $\frac{(V - \mu)\sigma}{\gamma \Lambda(\kappa)}$ for the insider.
- As $\kappa \rightarrow 0$, Y^* converges to

$$dY_t^* = \sigma dB_t + \sigma \frac{V - \mu - \frac{Y_t^*}{\sigma}}{1 - t} dt;$$

that is, $\frac{Y^*}{\sigma}$ becomes a Brownian bridge from 0 to $\frac{V - \mu}{\gamma}$.

- In the above Gaussian setting

$$\alpha_t^* \sim N\left(\frac{V - \mu}{\gamma} \Lambda(\kappa) \sigma, \frac{t \sigma^2 \Lambda^4(\kappa)}{1 - t \Lambda^2(\kappa)}\right).$$

That is, the rate of trading is constant on average over the trading horizon.

Asymptotics for large penalties

- In the above Gaussian setting

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- As $\kappa \rightarrow \infty$, $\Lambda \rightarrow 0$. Thus, even after normalizing the insider's trades by $\Lambda(\kappa)$, the standard deviation of the trading rate remains small, that is, order of $\Lambda(\kappa)$.
- Thus, the insider buys (sells) at the constant rate $\frac{|V - \mu|}{\gamma} \Lambda(\kappa) \sigma$ if her private value is greater (smaller) than the initial price, μ , of the asset.

Expected penalties are non-monotone

Insider's expected wealth is decreasing in κ . However, the expected penalty is not monotone!

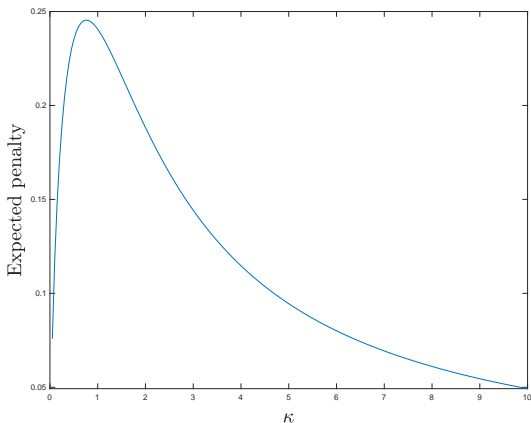


Figure: Expected penalties in equilibrium normalised by $\gamma\sigma$.

The Kyle model without frictions

Kyle model with penalties

Emergence of Schrödinger potentials

Equilibrium for the Kyle-Back model with penalties

Regulating insider trading

Conclusion

Regulating insider trading

Regulator's dilemma

- Suppose V is Gaussian, and the regulator has the following simple objective:

$$\begin{aligned} & \min_{\kappa} \Lambda(\kappa)\gamma\sigma + R\kappa\Lambda(\kappa)\gamma^2, \\ & \text{subject to } -\frac{\gamma\sigma}{2}\kappa \log(\kappa\Lambda(\kappa)) \geq b, \end{aligned} \tag{24}$$

for some $R > 0$ and $b > 0$.

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- Note that the first term is the expected loss of the noise traders and the second is the expected post-trade variance adjusted by the factor R , which measures the sensitivity of the policy toward price efficiency.
- Since the expected penalties are bounded, if

$$b^0 := \frac{b}{\gamma\sigma} < \bar{P} := \sup -\frac{1}{2}\kappa \log(\kappa\Lambda(\kappa)),$$

the budget constraint cannot be satisfied. Then it can be shown that if the amount of noise trading is rather small, the regulator does not run an investigation.

- Now suppose $b^0 \geq \bar{P}$. Define

$$\mathcal{K}(a) := \{\kappa : -\kappa \log(\kappa \Lambda(\kappa)) \geq a\}. \quad (25)$$

- Then the optimal penalty rate is $c^* := \kappa^* \frac{\gamma}{\sigma}$, where

$$\kappa^* = \arg \min_{\kappa \in \mathcal{K}(b_0)} \Lambda(\kappa) + R \frac{\gamma}{\sigma} (1 - \Lambda^2(\kappa)). \quad (26)$$

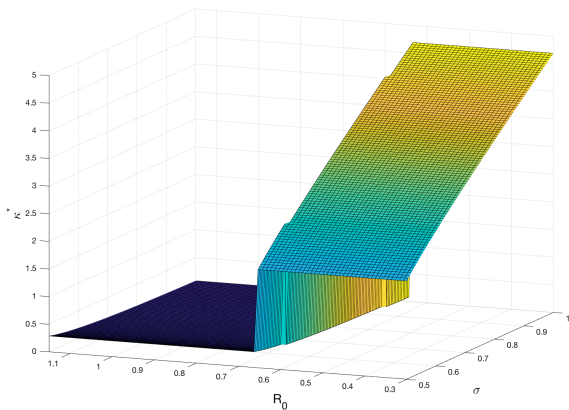


Figure: The optimal penalty rate κ^* as a function of noise volatility σ , and the regulator's sensitivity toward price efficiency, $R_0 = R \frac{\gamma}{\sigma}$. The figure assumes $\gamma = 1$ and $b = 0.1$.

Conclusion

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- Solved in continuous time the extension of the Kyle model with penalties on insider trading.
- The solution reveals an interesting connection between quadratic BSDEs and h-transforms, where the terminal condition of the BSDE is determined in equilibrium.
- One can use this setup to solve the regulators's problem with the objective to minimise uninformed traders losses but also keep the informational efficiency above a certain level. For details, see U. Çetin, *Insider trading with legal risk in continuous time*, SSRN preprint.

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- Extension to general convex cost functions will be of great interest as it is important for the regulator to decide the best cost functional to regulate insider trading.