# **Optimal Monitoring in Dynamic Financial Contracts**

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The Three Ms of Malfeasance

- mismanagement
- misallocation
- misappropriation

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- Applies generally:
  - VC financing today
  - Publicly Traded Corporations

# The Three Ms of Malfeasance

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# **Stylised Facts**

 pay-performance sensitivity and monitoring are *substitutes* [Bernstein, Giroud, Townsend, Bengtsson and Ravid]

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- Components of contract
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  - Performance sensitivity
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#### **Milgrom-Roberts**

output  $y = a + \sigma \varepsilon$ wage  $w = s_0 + \beta y$ 

Optimal  $(a, \sigma, \beta)$  jointly determined

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Limited Liability implies firm "risk averse"

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- Fully determines monitoring intensity
- Comparative statics of risk aversion

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- Empirical literature on Governance ...

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- Agent has (i) limited liability and (ii) no wealth
- Principal covers operating losses

### • Firm produces cash flows

 $dY_t = \mu dt + \sigma_t dB_t$ 

# In DS: $\Sigma$ is singleton

- Volatility  $\sigma_t$  chosen by Principal at Cost  $\rho(\sigma_t)$
- $\sigma_t \in \Sigma = \{\sigma_{(0)}, \dots, \sigma_{(n)}\}, \quad \sigma_{(i)} > \sigma_{(i+1)}$ •  $\rho(\sigma_{(i)}) < \rho(\sigma_{(i+1)})$ : More accuracy is costlier

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# • $D_t \ge 0$ is cash-flow diversion

# In DS: $\Sigma$ is singleton

- Benefit of diversion  $D_t$  is  $\lambda D_t$ , where  $\lambda \in (0, 1]$
- Always optimal to implement truth-telling:
  - $D_t = 0$  for all  $t \ge 0$

# Model (contd)

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#### Recap

Cash Flow

 $dY_t = \mu dt + \sigma_t dB_t$ 

Agent Reports  $\hat{Y}_t$ 

 $d\hat{Y}_t = (\mu - D_t)dt + \sigma_t dB_t$ 

Agent Benefit =  $\lambda D_t$ ,  $\lambda \in (0, 1]$ Principal flow cost =  $\rho(\sigma_t)$ 

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# Alternative Model

Cash Flow

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Agent Benefit =  $\lambda_t D_t, \lambda_t \in (0, 1]$ Principal flow cost =  $\rho(\lambda_t)$ 

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  - $\sigma = (\sigma_t)$ : Monitoring levels

Profit = 
$$F(w = w_0; \Phi) := \mathbf{E}^{D=0,\sigma} \left[ \int_0^\tau e^{-rt} \left[ (\mu - \rho(\sigma_t)) dt - dC_t \right] \right]$$

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• Promise keeping

$$w_0 = \mathbf{E}^{D=0,\sigma} \left[ \int_0^{\tau} e^{-\gamma t} dC_t \right]$$

• Incentive Compatibility

$$\mathbf{E}^{D=0,\sigma}\left[\int_{0}^{\tau} e^{-\gamma t} dC_{t}\right] \geq \mathbf{E}^{D,\sigma}\left[\int_{0}^{\tau} e^{-\gamma t} \left(dC_{t} + \lambda D_{t} dt\right)\right]$$

# **Continuation Utility**

# • $W = (W_t)$ is agent's continuation utility process

$$W_t = \mathbf{E}_t^{\hat{Y},\sigma} \left[ \int_t^\tau e^{-\gamma(s-t)} \left[ dC_s + \lambda (dY_s - d\hat{Y}_s) \right] \right]$$

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  - works because output is BM, iid increments

$$dW_t = \gamma W_t dt - [dC_t + \lambda (dY_t - d\hat{Y}_t)] + Z_t \sigma_t^{-1} \cdot \underbrace{(d\hat{Y}_t - \mu dt)}_{= \sigma_t dB_t - d(Y_t - \hat{Y}_t)}$$

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 for all  $t \ge 0$ 

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### • Value Function

$$F(w) = \sup_{(C,\tau,\sigma)} \mathbf{E}^{D=0,\sigma} \left[ \int_0^\tau e^{-rt} \left[ (\mu - \rho(\sigma_t)) dt - dC_t \right] \right]$$

subject to (i) IC and (ii) PK

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  - mix between w and w' to concavify

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subject to (i) IC and (ii) PK

$$-w \leq F(w) \leq \mu/r - w$$

- Lower bound is immediate termination:  $w \mapsto -w$
- Upper bound is first best:  $w \mapsto \mu/r w$
- *F* is **concave** 
  - mix between w and w' to concavify
  - concave even without mixing ...

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• *F* is C<sup>2</sup> solution [Schauder theory]

#### Dynamics of Monitoring

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$$F'(w^*) = -1 \implies F'(w) = -1 \text{ for } w \ge w^*$$
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• Backloaded payments: Ct satisfies

$$C_t = \int_0^t \mathbb{1}(W_s \ge w^*) dC_s$$

# **Optimal Contract**



# **Optimal Contract: Monitoring**

$$rF(w) = \mu + \gamma w F'(w) + \max_{\substack{C \\ = 0}} \left[ -\left(F'(w) + 1\right) dC \right] + \max_{\substack{z \ge \lambda \sigma, \sigma \in \Sigma}} \left[ \frac{1}{2} z^2 F''(w) - \rho(\sigma) \right]$$

- F concave implies
  - $z = \lambda \sigma$
- Optimal  $\sigma$  depends on F''
- $-\lambda^2 F'' = risk aversion$

## **Optimal Contract: Monitoring**



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# Shape of Risk Aversion

Risk Aversion = 
$$-\lambda^2 F''(w) \propto r \left[\frac{\mu}{r} - F(w) - w\right] + \frac{\gamma w (F'(w) + r/\gamma)}{\frac{\varphi}{\varphi}}$$

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**Proposition:** There exists  $\mu^{\dagger}$  such that

- efficiency loss due to agency ≥ 0: decreases in w
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 $\mu \ge \mu^{\dagger} \implies \begin{cases} F'(0;\mu) \ge 0\\ \mathsf{RA} \uparrow \mathsf{then} \downarrow \end{cases}$  $\mu < \mu^{\dagger} \implies \begin{cases} F'(0;\mu) < 0\\ \mathsf{RA} \mathsf{decreases} \end{cases}$ 

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Risk Aversion increases in  $\mu$  (ii) *w* and  $\mu$  are *complements* 

# **Optimal Contract** — Implementation

#### Securities and Assets

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- Stock pays *dividend*  $\lambda^{-1} dC$  when  $M_t = w^* / \lambda$  [agent controls dividends]
**Stock Prices** 

#### **Stock Price**

$$S_t = \mathbf{E}_t \left[ \int_t^\tau e^{-r(s-t)} \lambda^{-1} dC_s \right]$$

but  $S_t = \mathcal{S}(M_t)$ , so

 $dS_t = rS_t dt + V_t dB_t - \lambda^{-1} dC_t$ 

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BVP for stock prices  $rS(m) = \gamma mS'(m) + \frac{1}{2}\sigma^2(\lambda m)S''(m)$ • S(0) = 0•  $S'(w^*/\lambda) = 1$ 

Then, 
$$S_t = S(M_t)$$
 and  $S(\cdot)$  is

- strictly increasing
- strictly concave  $\implies$  continuous
- C<sup>2</sup> except at finitely many points even though σ(·) discontinuous!

# **Comparative Statics**

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### Dynamics: Monitoring increases after drop in stock price (Vafeas 1999)

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## Proposition

$$(1-\lambda)S_t + D_t > F(\lambda M_t) + M_t$$

market value

true value

Difference = 
$$\mathbf{E}_t \left[ \int_t^\tau e^{-r(s-t)} \rho(\sigma_s) ds \right]$$
 = monitoring costs

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## **Comparative Statics — Extinction Time**

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#### **Follow Sannikov**

Output  $dX_t = a_t dt + \sigma_t dB_t$ Effort cost  $h(a) = \frac{1}{2}a^2$ IC  $\beta_t = a_t$ 

$$rF(w) = \max_{\sigma,c,\beta} \left[ r(\underset{=\beta}{a} - c) + rF'(w)(w - u(c) + \frac{1}{2} \underset{=\beta^2}{a^2}) + \frac{1}{2}F''(w)r^2\beta^2\sigma_t^2 \right]$$
(HJB)

### **Milgrom-Roberts**

when pay-sensitivity  $\beta$  is higher, monitoring is also higher (because  $\sigma$  is correspondingly lower)

- monitoring and pay-sensitivity are complements
- sensitivity β increasing in optimal action

### **Follow Sannikov**

Output  $dX_t = a_t dt + \sigma_t dB_t$ Effort cost  $h(a) = \frac{1}{2}a^2$ IC  $\beta_t = a_t$ 

# **Monitoring Intensity**

$$\boldsymbol{\beta} = -\left[r\sigma^{2}\boldsymbol{F}''(\boldsymbol{w}) + \boldsymbol{F}'(\boldsymbol{w})\right]^{-1}$$

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### **Comparative Statics via Comparison Theorem**

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# Doesn't require F, G to be C<sup>2</sup>; $\Psi$ can be nonlinear

Let  $F, G : [0, m^{\dagger}] \rightarrow \mathbb{R}$  and  $\Psi(m, F, F', F'') \ge 0$   $\Psi(m, G, G', G'') \le 0$ If  $G(m^{\dagger}) \le F(m^{\dagger})$ , then  $G(m) \le F(m) \quad \forall m \in [0, m^{\dagger}]$ 

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# **Example: Stock Prices and increase in** $\mu$ BVP for stock prices

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# Conclusion

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# Thank you!

# **SOX and Public Policy**

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$$\Delta_{\text{sox}} \mathcal{H}(m) := \underbrace{\Delta_{\text{gov}} \mathcal{H}(m)}_{\text{governance}} + \underbrace{\Delta_{\mu} \mathcal{H}(m)}_{\text{profitability}}$$

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### Effect of SOX

$$\Delta_{\text{sox}} \hat{\mathcal{F}}(m) < 0$$

$$\Delta_{\text{sox}} m^{\star} = \underline{\Delta}_{\text{gov}} m^{\star} + \underline{\Delta}_{\mu} m^{\star}_{<0}$$

$$\Delta_{\text{sox}} \mathcal{S}(m) = \underline{\Delta}_{\text{gov}} \mathcal{S}(m) + \underline{\Delta}_{\mu} \mathcal{S}(m)_{<0}$$

$$\Delta_{\text{sox}} \mathcal{Z}(m) = \underline{\Delta}_{\text{gov}} \mathcal{Z}(m) + \underline{\Delta}_{\mu} \mathcal{Z}(m)_{<0}$$

# **SDE for** *W*

Continuation Utility

$$\underbrace{V_t \cdot \Delta_t \cdot S_t}_{\text{local governance}} = \lambda \sigma_t$$

$$\bigvee_{t} \cdot \Delta_{t} \cdot S_{t} = \lambda \sigma_{t}$$
local governance

•  $V_t = local volatility$ 

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Measurement

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• V<sub>t</sub> via **Dupire's formula** 

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#### Measurement

- V<sub>t</sub> via **Dupire's formula** 
  - $\mathcal{V}_t = \mathcal{S}' \big( \mathcal{S}^{-1}(\mathcal{S}_t) \big) \sigma(\lambda \mathcal{S}^{-1}(\mathcal{S}_t)) / \mathcal{S}_t$

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- Induces Governance Smile ...