

UCLA Anderson  
Accounting Theory Study Guide

2020

Athanasse Dimitri Zafirov

This study document spans most of the signaling papers covered in the Winter 2019 *Doctoral Seminar in Financial Accounting Theory* (MGMTPHD 236) class thought by Professor Brett Trueman and the Spring 2020 *Theoretical models in accounting* (MGMTPHD 236) class thought by Professor Judson Caskey.

The theory sections of the UCLA Anderson Accounting area comprehensive examination is different from most final exams. It is best approached by a deep understanding of the base material instead of attempts to “study to the exam” (which is sufficient for many classes). While around 50 theoretical papers, many of which use mathematical theorems you might be encountering for the first time, may seem like a lot, it is good to be able to break these down into categories.

Accounting theory (and this exam) focuses on signaling games, which is a subcategory of sequential games with incomplete information, a subfield of game theory. From there it inherits the main equilibrium concept of Perfect Bayesian Equilibrium (PBE). This should be reviewed beforehand to better appreciate the many papers using them as a base.

We can attempt to (imperfectly) break down most of the main signaling papers covered in these classes into the following categories along with their representative papers:

- Persuasion games
  - with Full Revelation, aka “unraveling” (e.g.: Grossman [1981])
  - with Disclosure Costs (e.g.: Verrecchia [1983])
  - with Information Uncertainty (e.g.: Jung & Kwon [1988])
  - with Uncertainty About Sender Incentives (e.g.: Einhorn [2007])
- Costless Signaling Games, aka “Cheap Talk” (e.g.: Crawford & Sobel [1982])
- Costly Reporting Games, aka “Signal Jamming”
  - with Costly Disclosure and Full Revelation (e.g.: Stein [1989])
  - with Costly Disclosure and Reporting Uncertainty (e.g.: Fischer & Verrecchia [2000])
- Rational Expectations
  - Price Determined by Market Clearing (e.g.: Grossman & Stiglitz [1980])
  - Price Determined by Competitive Market Maker (e.g.: Kyle [1985])

Herding could also be used as a useful category, exemplified in papers such as Scharfstein & Stein (1990) and Trueman (1994).

Solving for equilibrium is usually your main concern and challenge in the questions you may face. There are a few types of equilibria to keep in mind:

- Fully Revealing (also known as Fully Separating)
- Partially Revealing (or Partially Separating)

- Pooling
- Partition Equilibrium (as seen in “Cheap Talk” models)
  - where Babbling Equilibrium is a special case

Together these canonical papers span the main equilibrium solution styles you should master and know when to use, these are:

- Solving for the indifference threshold
  - When there is a cost to disclosure
  - When there is uncertain information endowment
  - When there is uncertainty about Sender’s objective function
- Inverting the Sender’s reporting function
  - Which may include solving for the other agent’s linear conjecture coefficients, as often seen in Rational Expectations models
- Use of the Intuitive Criterion refinement, such as for an “unraveling” argument and commonly seen in achieving certain pooling equilibria

Finally, there are papers that escape the above categories or span a number of them. The rest of this document is heavily based on notes handed out in class, with some extra steps added in for clarity. It tends to focus either on the paper’s main equilibrium result of interest or what was focused on in the class it was presented.

One particularly helpful aspect of the process boxes I made use of is the ability to clearly isolate all the “inputs” to a paper’s model, and have them for quick reference. Between 50 papers, it is easy to confuse variables and assumptions (“was  $r$  risk-aversion or risk-tolerance in this paper? Was  $x$  demand or liquidity?”), this document attempts to help with this by trying to have everything on a single (sometimes two-sided) sheet of paper.

I would like to thank the UCLA Anderson Accounting professors for their support and guidance. All typos and mistakes are my own.

The next section lists the papers and problems covered in this document.

# ***Doctoral Seminar in Financial Accounting Theory***

## **(MGMTPHD 236, Winter 2019)**

Professor Brett Trueman

### **Rational Expectations**

- S. Grossman and J. Stiglitz, "On the Impossibility of Informationally Efficient Markets," *American Economic Review*, June 1980.
- R. Verrecchia, "Information Acquisition in a Noisy Rational Expectations Economy," *Econometrica*, November 1982.
- A. Kyle, "Continuous Auctions and Insider Trading," *Econometrica*, November 1985.
- O. Kim and R. Verrecchia, "Trading Volume and Price Reactions to Public Announcements," *Journal of Accounting Research*, Autumn 1991.

### **Signaling**

- G. Akerlof, "The Market for Lemons," *Quarterly Journal of Economics*, August 1970.
- M. Spence, "Job Market Signaling", *Quarterly Journal of Economics*, August 1973.
- H. Leland and D. Pyle, "Informational Asymmetries, Financial Structure, and Financial Intermediation," *Journal of Finance*, May 1977.
- S. Titman and B. Trueman, "Information Quality and the Valuation of New Issues," *Journal of Accounting and Economics*, May 1986.
- S. Ross, "The Determination of Financial Structure: The Incentive Signaling Approach," *Bell Journal of Economics*, Spring 1977.
- M. Miller and K. Rock, "Dividend Policy Under Asymmetric Information," *Journal of Finance*, September 1985.
- S. Datar, J. Feltham, and J. Hughes, "The Role of Audits and Audit Quality in Valuing New Issues," *Journal of Accounting and Economics*, March 1991.

### **Voluntary Disclosure**

- R. Verrecchia, "Discretionary Disclosure," *Journal of Accounting and Economics*, December 1983.
- R. Dye, "Disclosure of Nonproprietary Information," *Journal of Accounting Research*, Spring 1985.
- W. Jung and Y. Kwon, "Disclosure When the Market is Unsure of Information Endowment of Managers," *Journal of Accounting Research*, Spring 1988.
- J. Hughes and S. Pae, "Voluntary Disclosure of Precision Information," *Journal of Accounting and Economics*, June 2004.
- M. Darrough and N. Stoughton, "Financial Disclosure Policy in an Entry Game," *Journal of Accounting and Economics*, January 1990.
- M. McNichols and B. Trueman, "Public Disclosure, Private Information Collection and Short-term Trading", *Journal of Accounting and Economics*, January 1994.
- P. Fischer, and R. Verrecchia, "Reporting Biases," *The Accounting Review*, April 2000.
- E. Einhorn, "Voluntary Disclosure Under Uncertainty About the Reporting Objective," *Journal of Accounting and Economics*, July 2007.

### **Asymmetric Information**

- S. Myers and N. Majluf, "Stock Issues and Investment Policy When Firms Have Information That Investors Do Not Have," *Journal of Financial Economics*, June 1984.
- B. Trueman, "Analyst Forecasts and Herding Behavior," *Review of Financial Studies*, Spring 1994.
- A. Beyer, "Financial Analysts' Forecast Revisions and Managers' Reporting Behavior," *Journal of Accounting and Economics*, December 2008.
- D. Scharfstein and J. Stein, "Herd Behavior and Investment," *American Economic Review*, June 1990.

### ***Theoretical models in accounting***

#### **(MGMTPHD 236, Spring 2020)**

Professor Judson Caskey

#### **Background**

- Milgrom, P., and N. Stokey. 1982. Information, trade and common knowledge. *Journal of Economic Theory* 26(1): 17-27
- Grossman, J. 1981. The informational role of warranties and private disclosure about product quality. *Journal of Law and Economics* 24(3): 461-484
- Milgrom, P. 1981. Good news and bad news: Representation theorems and applications. *Bell Journal of Economics* 12(2): 380-391

#### **Ohlsonomics**

- Ohlson, J. 1995. Earnings, book values and dividends in equity valuation. *Contemporary Accounting Research* 11(2): 661-687

#### **Rational Expectations**

- Grossman, S., and J. Stiglitz. 1980. On the impossibility of informationally efficient markets. *American Economic Review* 70(3): 393-408
- Caskey, J. 2009. Information in equity markets with ambiguity-averse investors. *Review of Financial Studies* 22(9): 3595-
- Guttman, I. 2010. The timing of analysts' earnings forecasts. *The Accounting Review*
- Kyle, A. 1985. Continuous auctions and insider trading. *Econometrica* 53(6): 1315-1335
- Kyle, A. 1989. Informed speculation with imperfect competition. *Review of Economic Studies* 56(3): 317-355

#### **Verifiable disclosure models**

- Verrecchia, R. 1983. Discretionary disclosure. *Journal of Accounting and Economics* 5: 179-194
- Jung, W., and Y. Kwon. 1988. Disclosure when the market is unsure of information endowment of managers. *Journal of Accounting Research* 26(1): 146-153

- Einhorn, E., and A. Ziv. 2012. Biased voluntary disclosure. *Review of Accounting Studies* 17(2): 420-442

### **Cheap Talk**

- Crawford, V., and J. Sobel. 1982. Strategic information transmission. *Econometrica* 50(6): 1431-1451
- Ottaviani, M., and P. Sorensen. 2006. The strategy of professional forecasting. *Journal of Financial Economics* 81(2): 441-466
- Bertomeu, J., and I. Marinovic. 2016. A theory of hard and soft information. *The Accounting Review* 91(1): 1-20

### **Earnings management/Costly signaling**

- Kartik, N. 2009. Strategic communication with lying costs. *Review of Economic Studies* 4(1): 1359-1395
- Caskey, J. 2014. The pricing effects of securities class action lawsuits and litigation insurance. *Journal of Law, Economics and Organization* 30(3): 493-532
- Fischer, P., and R. Verrecchia. 2000. Reporting bias. *The Accounting Review* 75(2): 229-245
- Stein, J. 1989. Efficient capital markets, inefficient firms: A model of myopic corporate behavior. *Quarterly Journal of Economics* 104(4): 655-669

### **Voluntary + mandatory disclosure**

- Einhorn, E. 2018. Competing information sources. *The Accounting Review* 93(4): 151-176
- Cheynel, E., and C. Levine. 2020. Public disclosures and information asymmetry: A theory of the mosaic. *The Accounting Review* 95(1): 79-99

### **Continuous time dynamic models**

- Pastor, L., and P. Veronesi. 2003. Stock valuation and learning about profitability. *Journal of Finance* 58(5): 1749-1789

## ***Solutions to Homework Assignments and Previous Theory Comps***

### **Prof. Brett Trueman Assignments:**

- Assignment 1 (Normal Update of Expected Value)
- Assignment 2 (Rational Expectations)
- Assignment 3 (Costly Signaling)
- Assignment 4 (Inverting Sender's Function)
- Assignment 5 (Uncertain Information Endowment)
- Assignment 6 (Herding)

### **Prof. Judson Caskey Assignments:**

- Assignment 1 Q1 (Rational Expectations)
- Assignment 1 Q4 (Guttman [2010])
- Assignment 2 Q3 (Signal Jamming and Cheap Talk)

### **Previous Theory Comp**

- Theory 1: Comp 2019
- Theory 1: Comp 2020
- Theory 2: Comp 2019
- Theory 2: Comp 2020

# Grossman & Stiglitz (1980)

## I. Rational Expectations (Partially Revealing Rational Expectations Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



## 1 Demand is independent of wealth

The goal of this section is to show that CARA utility makes it so the demand for risky asset does not depend on wealth, explaining why endowments  $\bar{M}_i$  &  $\bar{X}_i$  and wealth  $W_{ti}$  do not enter the systems found in further sections.

**Show demand is independent of wealth**

Knowing that  $W_{1i} \sim \text{Normal}$

Trader's expected utility can be written as:

$$E[U_i(W_{1i})|I_i] = -\exp\{-a[E[W_{1i}|I_i] - \frac{a}{2}\text{var}[W_{1i}|I_i]]\}$$

maxing utility is thus equiv. to maxing:

$$E[W_{1i}|I_i] - \frac{a}{2}\text{var}[W_{1i}|I_i]$$

note that:  $E[W_{1i}|I_i] = W_{0i} + (E[\mu|I_i] - P)X_i$

and:  $\text{var}[W_{1i}|I_i] = X_i^2\text{var}(\mu|I_i)$ ,

substituting, we get objective function:

$$W_{0i} + (E[\mu|I_i] - P)X_i - \frac{a}{2}X_i^2\text{var}(\mu|I_i)$$

FOC: demand  $X_i = \frac{E(\mu|I_i) - P}{a \cdot \text{var}(\mu|I_i)}$

CARA coef. of risk-aversion:  $a \rightarrow$

$U_i(W_{1i}) = -\exp\{-aW_{1i}\} \rightarrow$

risk-free endowment  $\bar{M}_i \rightarrow$

risky endowment  $\bar{X}_i \rightarrow$

price  $P \rightarrow$

constraint  $PX_i + M_i = W_{0i} \equiv \bar{M}_i + \bar{X}_i \rightarrow$

information  $I_i \rightarrow$

$t=1$  wealth:  $W_{1i} = M_i + \mu X_i \rightarrow$

knowing that  $\mu \sim N(\bar{\mu}, \sigma_x^2) \rightarrow$

## 2 Equilibrium when fraction $\lambda$ are informed

The equilibrium threats  $\lambda$  as given thus cost of information acquisition  $c$  is ignored.

<b>Show <math>\exists</math> Partially Revealing Rational Expectations Equilibria</b>	
<p>share of informed: <math>\lambda \rightarrow</math></p> <p>CARA coef. of risk-aversion: <math>a \rightarrow</math></p> <p>per-capita liquidity <math>x \sim N(0, \sigma_x^2) \rightarrow</math></p> <p>private signal: <math>\theta = \mu + \epsilon \rightarrow</math></p> <p><math>\mu \sim N(\bar{\mu}, \sigma_\mu^2) \perp \epsilon \sim N(0, \sigma_\epsilon^2) \rightarrow</math></p> <p>signal conjecture form: <math>P = \alpha + \beta\theta + \gamma x \rightarrow</math></p>	<p>Uninformed Conjecture:</p> $z \equiv \theta + bx = (\mu + \epsilon) + bx$ <p>where <math>b = \gamma/\beta</math></p> $\text{demand } X_U = \frac{E[\mu z] - P}{a \cdot \text{var}(\mu z)}$ $E[\mu z] = \bar{\mu} + \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_\epsilon^2 + b^2 \sigma_x^2} (z - \bar{\mu})$ $\text{Var}(\mu z) = \sigma_\mu^2 - \frac{\sigma_\mu^4}{\sigma_\mu^2 + \sigma_\epsilon^2 + b^2 \sigma_x^2}$ $\text{demand } X_I = \frac{E[\mu \theta] - P}{a \cdot \text{var}(\mu \theta)}$ $E[\mu \theta] = \bar{\mu} + \frac{\sigma_\mu^2}{\sigma_\mu^2 + \sigma_\epsilon^2} (\theta - \bar{\mu})$ $\text{Var}(\mu \theta) = \sigma_\mu^2 - \frac{\sigma_\mu^4}{\sigma_\mu^2 + \sigma_\epsilon^2}$ <p>Thus we have as a function of:</p> $X_U(\bar{\mu}, a, b, \theta, \sigma_\mu^2, \sigma_\epsilon^2, \sigma_x^2)$ $X_I(\bar{\mu}, a, b, \theta, \sigma_\mu^2, \sigma_\epsilon^2, \sigma_x^2)$ <p>plugging into <b>Demand = Supply</b> :</p> $\lambda X_I + (1 - \lambda) X_U = x$ <p>gives: <math>P = \alpha + \beta\theta + \gamma x</math></p> <p>So conjecture terms converge to specific, terms that are expressions of setting parameters. This means <math>\exists</math> a partially revealing price equilibrium in this setting.</p>
	<p>→ informed demand <math>X_I</math></p> <p>→ uninformed demand <math>X_U</math></p> <p>→ <math>\beta</math> as a function of parameters</p> <p>→ <math>\gamma</math> as a function of parameters</p> <p>→ <math>\alpha</math> as a residual constant</p> <p>→ <math>P = \alpha + \beta\theta + \gamma x</math></p> <p>→ price informativeness proxy: <math>\frac{\beta^2}{\gamma^2}</math></p> <p>→ <math>\exists</math> partially revealing equilibrium</p>

### Additional notes:

The process makes use the previous section's optimal demand function and the following properties of normally distributed random variables:

$$E[\mu|z] = E[\mu] + \frac{\text{cov}(\mu, z)}{\text{var}(z)} (z - E[z])$$

$$\text{var}(\mu|z) = \text{var}[\mu] - \frac{\text{cov}^2(\mu, z)}{\text{var}(z)}$$

Law of total variance:  $\text{var}(X) = \mathbf{E}[\text{var}(X | Y)] + \text{var}(\mathbf{E}[X | Y])$

$$q(\mathbf{E}[v] - p) - \frac{\gamma}{2}\text{var}(v)q^2 \quad FOC: 0 = \mathbf{E}[v] - p - \gamma\text{var}(v)q \Rightarrow \begin{cases} p = \mathbf{E}[v] - \gamma\text{var}(v)q \\ q = \frac{\mathbf{E}[v] - p}{\gamma\text{var}(v)} \end{cases}$$

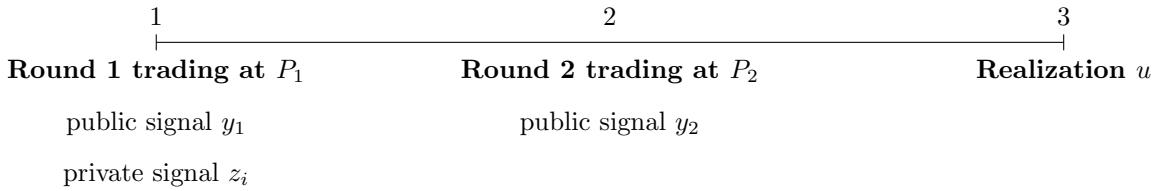
$$\begin{pmatrix} x \\ y \end{pmatrix} \sim \mathcal{N} \left( \begin{pmatrix} \mu_x \\ \mu_y \end{pmatrix}, \begin{pmatrix} \sigma_x^2 & \rho_{xy}\sigma_x\sigma_y \\ \rho_{xy}\sigma_x\sigma_y & \sigma_y^2 \end{pmatrix} \right) \quad \text{where } \rho_{xy} = \frac{\text{cov}(x, y)}{\sigma_x\sigma_y}$$

$$x | y \sim \mathcal{N} \left( \mu_x + \frac{\text{cov}(x, y)}{\text{var}(y)} (y - \mu_y), \sigma_x^2 - \frac{\text{cov}(x, y)^2}{\sigma_y^2} \right) \quad \text{Note: } \text{var}(x | y) = \sigma_x^2 (1 - \rho_{xy}^2)$$

# Kim & Verrecchia (1991)

## I. Rational Expectations (Partially Revealing Rational Expectations Equilibria )

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



Kim & Verrecchia examine the price and volume reactions to a public announcement. This paper has individual traders (subindexed  $i$ ) with their own particular endowments  $E_i$  & levels of risk tolerance  $r_i$ .

**Show  $\exists$  Partially Revealing Equilibria Through Price Amid Random Supply**

$i$ 's assessment of mean of  $u$  at t=2:

$$u_{2i} = E[u|Info_{t=2}]$$

$i$ 's assessment of precision of  $u$  at t=2:

$$K_{2i} = var^{-1}[u|Info_{t=2}]$$

thus trader  $i$ 's t=2 demand:

$$D_{2i} = r_i K_{2i} (u_{21} - P_2)$$

(demand  $D_{1i}$  is complex & omitted)

CARA risk-tolerance coef:  $r_i \rightarrow$

risky asset  $u \sim N(\bar{\mu}, h^{-1}) \rightarrow$

random supply  $x \sim N(0, t^{-1}) \rightarrow$

(t=1) public signal  $y_1 = u + \eta ; \eta \sim N(0, m^{-1}) \rightarrow$

(t=1) private signal  $z_i = u + \epsilon_i ; \epsilon_i \sim N(0, s_i^{-1}) \rightarrow$

(t=2) public signal  $y_2 = u + \nu ; \nu \sim N(0, n^{-1}) \rightarrow$

conjecture form of  $P_1 = \alpha_1 + \theta_1 y_1 + \beta_1 u - \gamma_1 x \rightarrow$

conjecture form of  $P_2 = \alpha_2 + \theta_{21} y_1 + \beta_2 u - \gamma_2 x \rightarrow$

Notes omit this explicitly, but as in previous Rational Expecations models, demand is then used to solve for the coefficients on price conjectures, (ie:  $\alpha_1, \theta_1, \beta_1, \gamma_1$  for  $P_1$ , and  $\alpha_2, \theta_{21}, \beta_2, \gamma_2$  for  $P_2$ ) as functions of economic parameters. This also lets us solve for  $P_1$  &  $P_2$  as functions of economic parameters.

Thus  $\exists$  a Partially Revealing Rational Expectations Equilibrium here. QED.

→  $i$ 's demand at t=2:  $D_{2i}$

→  $P_2$  &  $P_1$  as functions of params

→ Thus  $\exists$  partial price equilibrium  $P_2 - P_1 = \frac{n}{K_2} [Surprise + Noise]$

→  $Surprise = y_2 - \frac{h + m y_1 + s u + r^2 s^2 t q}{K_1}$

→  $Noise = \frac{x}{r K_1}$

→ Volume generated by  $i$  at t=2 is:

→  $|D_{2i} - D_{1i}| = |r_i(s_i - s)(P_2 - P_1)|$

→ Total Volume of all traders is:

→  $(\frac{1}{2} \int r_i |s_i - s| di) |P_2 - P_1|$

Notice in the notes that while trader  $i$ 's riskless endowment  $E_i$  is specified and eq.(4) elaborates  $i$ 's wealth, given CARA utility, they do not play a role above.

# Kyle (1985)

## I. Rational Expectations (Partially Revealing Rational Expectations Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



Kyle derives the rational expectations equilibrium using a different set of assumptions about how the market operates from those of Grossman & Stiglitz (1980), eg: traders are **risk neutral**.

### Show $\exists$ Partially Revealing Rational Expectations Equilibria

$$\max_x E^I[(v - P)x|\theta] = \max_x E^I[(v - E[v|y])x|\theta]$$

$$= \max_x E^I[(v - \mu + \lambda(x + u))x|\theta]$$

$$= \max_x (E^I[v|\theta]x - (\mu + \lambda x)x)$$

$$\text{FOC: } x = \frac{E^I[v|\theta] - \mu}{2\lambda}$$

$$\text{where } E[v|\theta] = \bar{v} + \frac{\sigma_v^2}{\sigma_v^2 + \sigma_\epsilon^2}(\theta - \bar{v})$$

$$\text{from conjecture on } x: \beta = \frac{\sigma_v^2}{2\lambda(\sigma_v^2 + \sigma_\epsilon^2)}$$

$$\alpha = \frac{\sigma_\epsilon^2 \bar{v}}{2\lambda(\sigma_v^2 + \sigma_\epsilon^2)} - \frac{\mu}{2\lambda}$$

$$\text{knowing that: } y = x + u = \alpha + \beta\theta + u$$

$$\text{now: } P = E(v|y) = \bar{v} + \frac{\text{cov}(v,y)}{\text{var}(y)}(y - E(y))$$

$$= \bar{v} + \frac{\beta^2 \sigma_v^2 (y - \alpha - \beta \bar{v})}{\beta^2(\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2}$$

$$\text{from } P \text{ conjecture: } \mu = \frac{(\beta^2 \sigma_\epsilon^2 + \sigma_u^2) \bar{v} - \alpha \beta \sigma_v^2}{\beta^2(\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2}$$

$$\text{and } y \text{ coefficient: } \lambda = \frac{\beta \sigma_v^2}{\beta^2(\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2}$$

Plug in  $x$  above to get rid of  $P$ 's

conjecture elements in conjecture of  $x$ :

$$\beta = \left[ \frac{\sigma_u^2}{\sigma_v^2 + \sigma_\epsilon^2} \right]^{0.5} \text{ and } \lambda = \frac{\sigma_v^2}{2\sigma_u^2(\sigma_v^2 + \sigma_\epsilon^2)^{0.5}}$$

would then solve for  $\alpha = -\beta \bar{v}$  and  $\mu = \bar{v}$

So conjecture terms are convergent,

this means  $\exists$  partially revealing

price equilibrium in this setting. QED.

Note: Mkt Maker makes 0 expected profits.

liquidity trader demand  $u \sim N(0, \sigma_u^2) \rightarrow$

signal of informed trader:  $\theta = v + \epsilon \rightarrow$

where  $v \sim N(\bar{v}, \sigma_v^2) \perp \epsilon \sim N(0, \sigma_\epsilon^2) \rightarrow$

$y$  is the  $t=0$  order flow:  $y = x + u \rightarrow$

competitive mkt maker sees order flow as a signal  $\rightarrow$

and he sets  $P = E^{mkt}[v|y]$  then clears the market  $\rightarrow$

risk neutral obj. function:  $\max_x E^I[(v - P)x|\theta] \rightarrow$

Informed trader can only place market orders and  $\rightarrow$

he considers mkt maker's price setting strategy  $\rightarrow$

to predict the effect his demand will have on price  $\rightarrow$

conjecture form of informed demand:  $x = \alpha + \beta\theta \rightarrow$

conjecture form of price:  $P = \mu + \lambda y \rightarrow$

show  $\exists$  a partial equilibrium in this setting  $\rightarrow$

## Verrecchia (1982)

### I. Rational Expectations (Partially Revealing Rational Expectations Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



The setting here is more general than in Grossman and Stiglitz (1980).

$T$  Traders ( $T \rightarrow \infty$ ) differ in their risk aversion and in the private signals they receive.

The subindex  $t$  index that the variable is at the individual trader level.

**Show  $\exists$  Partially Revealing Rational Expectations Equilibria**

CARA coef. of risk-tolerance:  $r_t \rightarrow$   $E_t(\mu_t|y_t, P) = y_0 + \frac{s_t \gamma^2 V(y_t - y_0) + \beta(P - y_0)}{(h_0 + s_t) \gamma^2 V + \beta^2}$

risky cashflow  $\mu \sim N(y_0, h_0^{-1}) \rightarrow$   $Var(\mu_t|y_t, P) = \frac{1}{h_0 + s_t + \beta^2 / \gamma^2 V}$

private signal of trader  $t$ :  $y_t = \mu + \epsilon_t \rightarrow$   $D_t(y_t, P) = \frac{r_t(E_t(\mu_t|y_t, P) - P)}{Var(\mu_t|y_t, P)}$

where  $\epsilon_t \sim N(0, s_t^{-1}) \rightarrow$

cost of signal with precision  $s_t$ :  $c(s_t) \rightarrow$  Notes skip exact derivation of the above

per capita liquidity  $x \sim N(0, V) \rightarrow$  but it's said to be done in the same

signal conjecture form:  $P = \alpha y_0 + \beta \mu + \gamma x \rightarrow$  manner as Grossman & Stiglitz (1980)

note:  $\beta^2 / \gamma^2 V$  is proxy  
for price informativeness

# Akerlof (1978)

## II. Signaling (Pooling Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)

$$\begin{array}{c} 1 \\ \hline + \\ Trading \end{array}$$

Akerlof shows that in presence of asymmetric information about quality, markets can break down when private information is unverifiable.

### 1 Asymmetric information case

Show this is a Pooling Equilibrium  
where markets break down (Demand=0)

**Supply:**  $S = S(p)$

$$S_1 = \begin{cases} \frac{p}{2}N & \text{if } p \leq 2 \text{ (if quality} \leq \text{ price)} \\ N & \text{if } p > 2 \end{cases}$$

$$S_2 = 0$$

Used car quality is unverifiable →

$E[\text{quality of used car on sale}]: \mu = p/2$

Group 1 Utility:  $U_{1i} = M + \sum_{1i} x_{1i} \rightarrow$

Group 2 Utility:  $U_{2i} = M + \frac{3}{2} \sum_{2i} x_{2i} \rightarrow$

**Demand:**

$$D_1 = \begin{cases} \frac{Y_1}{p} & \text{if } \mu \geq p \text{ (quality} \geq \text{ price)} \\ 0 & \text{if } \mu < p \text{ (quality} < \text{ price)} \end{cases}$$

Priors of used car quality:  $x \sim Unif(0, 2) \rightarrow$

income of group  $j$ :  $Y_j \rightarrow$

(M is other consumption) →

$$D_2 = \begin{cases} \frac{Y_2}{p} & \text{if } \frac{3}{2}\mu \geq p \\ 0 & \text{if } \frac{3}{2}\mu < p \end{cases}$$

→ Total Demand = 0

**Total demand:**

$$D(p, \mu) = \begin{cases} \frac{Y_1 + Y_2}{p} & \text{if } p \leq \mu \\ \frac{Y_2}{p} & \text{if } \mu < p \leq \frac{3}{2}\mu \\ 0 & \text{if } p > \frac{3}{2}\mu \end{cases}$$

but for a given  $p, \mu$  is  $p/2$ , plugging

$p = 2\mu$  above renders  $D(p, \mu) = 0$

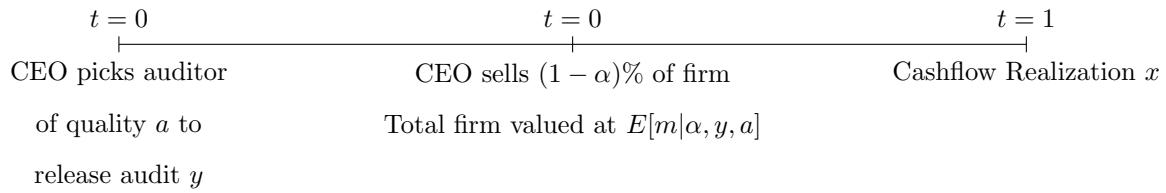
## 2 Symmetric information case (no one knows a car's quality)

	<b>Find price and demand in this Pooling Equilibria.</b>
	<b>Supply:</b> $S = S(p)$
	$S_1 = \begin{cases} N & \text{if } p \geq 1 \text{ (avg quality} \leq \text{price)} \\ 0 & \text{if } p < 1 \end{cases}$
	$S_2 = 0$
Group 1 Utility: $U_{1i} = M + \sum_{1i} x_{1i} \rightarrow$	
Group 2 Utility: $U_{2i} = M + \frac{3}{2} \sum_{2i} x_{2i} \rightarrow$	E[quality of used car on sale]: $\mu = 1$
Group 1 endowed with $N$ cars $\rightarrow$	We knew this was a Pooling Equilibrium.
Group 2 endowed with 0 cars $\rightarrow$	
Priors of used car quality: $x \sim Unif(0, 2) \rightarrow$	Price $p$
income of group $j$ : $Y_j \rightarrow$	then demand $D(p, \mu)$
(M is other consumption) $\rightarrow$	
	<b>Total demand:</b> $D(p, \mu) = \begin{cases} \frac{Y_1+Y_2}{p} & \text{if } p \leq 1 \\ \frac{Y_2}{p} & \text{if } 1 < p \leq \frac{3}{2} \\ 0 & \text{if } p > \frac{3}{2} \end{cases}$
	<b>In equilibrium:</b> $\begin{cases} p = 1 & \text{if } Y_2 \leq N \\ p = \frac{Y_2}{N} & \text{if } \frac{2Y_2}{3} < N \leq Y_2 \\ p = \frac{3}{2} & \text{if } N < \frac{2Y_2}{3} \end{cases}$

# Datar, Feltham & Hughes (1991)

## II. Signaling (Fully Revealing Signaling Equilibria)

This paper takes Leland & Pyle's setting of a CEO signaling with his level of retained ownership and augments it with an auditor, allowing him to potentially signal firm value without retaining as much.



CEO's CARA risk- <b>aversion</b> coefficient is $b \rightarrow$	
Investors are risk- <b>neutral</b> $\rightarrow$	
Project payoff is $x \sim N(m, \sigma^2)$ & $r_f$ rate = 0% $\rightarrow$	FOC: $m - v(\alpha) + (1 - \alpha)v'(\alpha) - b\sigma^2\alpha = 0$
Where CEO privately knows $m$ perfectly $\rightarrow$	$v(\alpha) = m \Rightarrow v'(\alpha) = \frac{b\alpha\sigma^2}{(1-\alpha)}$
Project requires \$ $k$ outlay $\rightarrow$	then integrating: $m = -b\sigma^2[\log(1 - \alpha) + \alpha] + z$ , $z$ is a constant
CEO sells $(1 - \alpha)\%$ of firm at $t=0$ & keeps $\alpha$ $\rightarrow$	
CEO's end wealth: $W = \alpha x + (1 - \alpha)E^{mkt}[x \alpha] - k \rightarrow$	Letting $\delta \equiv m - \underline{m}(y, a)$
$\max_{\alpha \in [0,1]} \alpha m + (1 - \alpha)E^{mkt}[x \alpha] - b\sigma^2\alpha^2/2 \rightarrow$	$v(\alpha) = -b\sigma^2[\log(1 - \alpha) + \alpha] = \delta$
$\alpha(m)$ denotes $\alpha$ retained by CEO with signal $m$ $\rightarrow$	Totally differentiating both sides (LHS: $d\alpha$ , RHS: $d\delta$ )
Through $\alpha$ , market infers CEO's info $m \rightarrow$	$v'(\alpha)d\alpha = 1d\delta$
$v^{mkt}(\alpha)$ denotes mkt assessed value $m$ given $\alpha \rightarrow$	$\frac{b\alpha\sigma^2}{(1-\alpha)}d\alpha = d\delta$
CEO conjectures his $\alpha$ will perfectly reveal info $\rightarrow$	$\frac{d\alpha}{d\delta} = \frac{1-\alpha}{b\sigma^2\alpha} > 0$
$\exists$ functions $\alpha(m)$ & $v^{mkt}(\alpha)$ that solve $\dagger$ while $\rightarrow$	
fully revealing $m$ through $\alpha$ : $v(\alpha(m)) = m$ , thus: $\rightarrow$	That is, the percentage of retained ownership is an increasing function of the difference between the entrepreneur's type and the worst type from which he must separate himself given the audited report.
$\dagger \max_{\alpha \in [0,1]} \alpha m + (1 - \alpha)v(\alpha(m)) - k - b\sigma^2\alpha^2/2 \rightarrow$	
CEO can hire an auditor of quality $a$ to report $y \rightarrow$	
Audit report $y$ is random & conditional on $m$ & $a \rightarrow$	
$m \supseteq y$ has no new info about $x$ for CEO given $m \rightarrow$	
$y$ is released alongside $\alpha$ & changes $z$ to $\underline{m}(y, a) \rightarrow$	
such that $-b\sigma^2[\log(1 - \alpha) + \alpha] = m - \underline{m}(y, a) \rightarrow$	

## Leland & Pyle (1977)

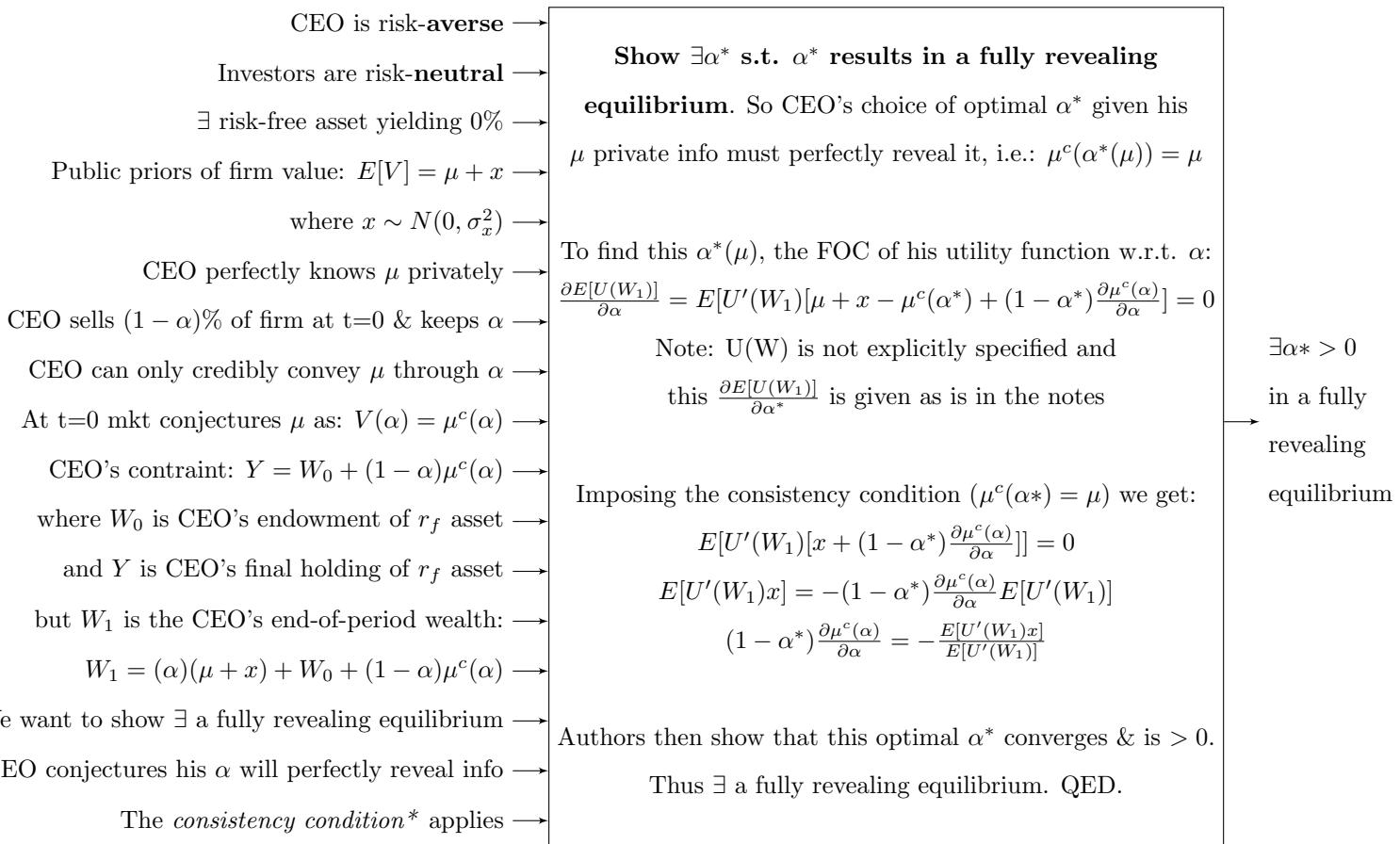
### II. Signaling (Fully Revealing Signaling Equilibria)

Based on Winter 2019 Accounting Theory (simplified) class notes on the paper (Prof. Trueman)



## 1 Fully Revealing Signaling Equilibrium

Firm value  $\mu$  is unknown to investors and a financial variable ( $\alpha$ : share of ownership retained by the CEO) is used to credibly communicate  $\mu$ .



\*The *consistency* (or incentive compatibility) *condition* is a necessary condition for equilibrium. Here it relates to the investors. Their conjecture of  $\mu$ , given that they observe  $\alpha$ , is equal to the actual value of  $\mu$ . That is, in equilibrium their conjecture is correct.

## 2 Solution to question in Section IV of the notes

Solve CEO's problem and find market's optimal conjecture schedule in a fully revealing signaling equilibrium.

Knowing that CEO is risk averse:  $E[U(W_1)] = E[W_1] - \frac{b}{2}\sigma^2(W_1)$

and given that  $W_1 = (\alpha)(\mu + x) + W_0 + (1 - \alpha)\mu^c(\alpha)$ :

$$E[U(W_1)] = E[\alpha(\mu + x) + W_0 + (1 - \alpha)\mu^c(\alpha)] - \frac{b}{2}\sigma^2(\alpha\mu + \alpha x + W_0 + (1 - \alpha)\mu^c(\alpha))$$

Note:  $E[x] = 0$  and  $\mu$  is deterministic.  $\mu^c(\alpha)$  is also deterministic because manager knows that market's conjecture function (which has no random component). Thus:

$$\max_{\alpha} E[U(W_1)] = \max_{\alpha} E[\alpha\mu + W_0 + (1 - \alpha)\mu^c(\alpha)] - \frac{b}{2}\alpha\sigma_x^2$$

$$FOC: \quad \frac{\partial E[U(W_1)]}{\partial \alpha} = \mu - \mu^c(\alpha) + (1 - \alpha)\frac{\partial \mu^c(\alpha)}{\partial \alpha} - \alpha b\sigma_x^2 = 0$$

Now that we've derived the FOC, we can impose the consistency condition letting  $\mu^c(\alpha) = \mu$ :

$$(1 - \alpha)\frac{\partial \mu^c(\alpha)}{\partial \alpha} - \alpha b\sigma_x^2 = 0$$

$$\frac{\partial \mu^c(\alpha)}{\partial \alpha} = \frac{\alpha b\sigma_x^2}{(1 - \alpha)}$$

Integrating:

$$\mu^c(\alpha) = -ab\sigma_x^2(\alpha + \log(|\alpha - 1|)) + constant$$

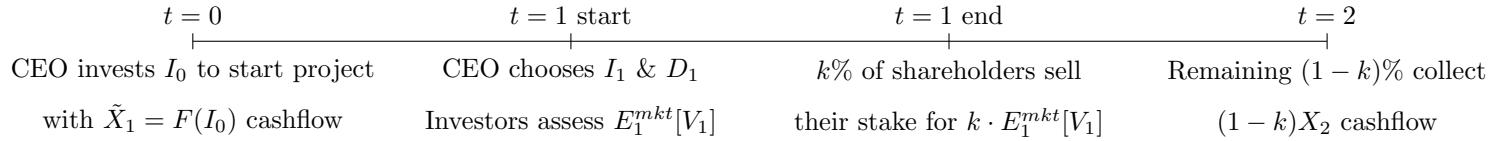
Now because we know that  $0 < \alpha < 1$  (how do we know  $< 1$ ? not sure), the term inside the RHS parenthesis is always negative, making the whole RHS positive given that  $-ab\sigma_x^2$  is also always negative.

From this we can know that:  $\min(\alpha) = 0$  gives  $\mu^c(\alpha = 0) = constant$  and that  $\mu^c(\alpha)$ , which is the investors' conjecture of the value of the firm given  $\alpha$  does indeed rise with with  $\alpha$ .

# Miller & Rock (1985)

## II. Signaling (Fully Revealing Signaling Equilibria)

In this setting the firm uses its dividend to signal its value. Firms can cheat by investing below optimal  $I_0 < I_0^*$  and releasing proceeds as dividend  $D_1$  to fool investors into estimating a higher persistent earnings surprise factor  $\epsilon_1$ . This, along with then private info  $I_0$  &  $\epsilon_1$ , is only made obvious to investors at t=2.



t=0: Firm invest  $I_0$  in production function  $F(I_0)$  →

Firm's earnings at t=1:  $\tilde{X}_1 = F(I_0) + \tilde{\epsilon}_1$ , and →

t=2:  $\tilde{X}_2 = F(I_1) + \tilde{\epsilon}_2 = F(\tilde{X}_1 - D_1) + \tilde{\epsilon}_2$  →

where  $D_1$  are dividends paid →

$\tilde{\epsilon}$ 's are random with mean 0 and  $E(\tilde{\epsilon}_2 | \epsilon_1) = \gamma \epsilon_1$  →

$V_1 = D_1 + \frac{1}{1+i} E[\tilde{X}_2] = D_1 + \frac{1}{1+i} [F(I_1) + \gamma \epsilon_1]$  →

s.t.  $X_1 = I_1 + D_1 \Leftrightarrow X_1 - I_1 = D_1$  →

t=1 firm value:  $V_1 = X_1 - I_1 + \frac{1}{1+i} [F(I_1) + \gamma \epsilon_1]$  →

$F$  is concave:  $F(I) \geq 0, F(0) = 0, F' > 0, F'' < 0$  →

$I_1^*$  is optimal investment level s.t.  $F'(I_1^*) = 1 + i$  →

or  $\frac{F'(I_1^*)}{(1+i)} - 1 = 0$ , where  $i$  is firm's WACC →

$E_0^{mkt}[V_1] = E_0^{mkt}[\tilde{X}_1 - I_1] + \frac{1}{1+i} E_0^{mkt}[F(I_1)]$  →

at t=0:  $E_0^{mkt}[V_1] = F(I_0) - I_1^* + \frac{1}{1+i} [F(I_1^*)]$  →

realizes at t=1:  $V_1 = X_1 - I_1^* + \frac{1}{1+i} [E_0^{mkt}(X_2)]$  →

$= F(I_0) + \epsilon_1 - I_1^* + \frac{1}{1+i} [F(I_1^*) + \gamma \epsilon_1]$  →

Earnings announcement effect:  $V_1 - E_0^{mkt}[V_1]$  →

$= (X_1 - E_0^{mkt}[\tilde{X}_1])[1 + \frac{\gamma}{1+i}] = E_0^{mkt}[\epsilon_1][1 + \frac{\gamma}{1+i}]$  →

So market can back out  $\epsilon_1$  from t=1 financial flows: →

$\epsilon_1 = D_1 - E_0^{mkt}[D_1] = X_1 - E_0^{mkt}[\tilde{X}_1]$  →

CEO can cheat by investing  $I_0 < I_0^*$  to increase  $D_1$  →

this inflates  $E_1^{mkt}[\epsilon_1], E_1^{mkt}[\epsilon_2 = \gamma \epsilon_1]$  and  $E_1^{mkt}[V_2]$  →

$(1 - k)\%$  of shareholders are LT investors (*keepers*) →

$k\%$  sell their stock when  $D_1$  is announced (*sellers*) →

*sellers* get  $V_1^m(D_1) = D_1 + \frac{1}{1+i} E_1^{mkt}[F(I_1) + \gamma \tilde{\epsilon}_1]$  →

*keepers* get  $V_1^d(X_1, D_1) = D_1 + \frac{1}{1+i} [F(I_1) + \gamma \epsilon_1]$  →

CEO weights both interests in objective function: →

$\max_{D_1, I_1} W_1 = kV_1^m + (1 - k)V_1^d$  s.t.  $D_1 + I_1 = X_1$  →

Show  $\exists D^*$  s.t.  $D^* \Rightarrow$  fully revealing equilibrium →

We are looking for a signaling schedule where

$\forall X_1 \exists$  an optimal  $D_1$  which the market uses to conjecture  $X_1$  from the signal  $D_1$  when valuing the firm. We assume this schedule is homogenous among rational shareholders such that:

$$V_1^m(D_1) = V_1^d(X_1(D_1), D_1) = V_1^d(X_1, D_1)$$

Substituting above into objective function:

$$\max_{D_1, I_1} W_1 = kV_1^d(X_1, D_1) + (1 - k)V_1^d(X, D_1)^d$$

s.t.  $D_1 + I_1 = X_1 \Leftrightarrow I_1 = X_1 - D_1$  (eliminating  $I_1$ )

$$\text{FOC } \frac{\partial W_1}{\partial D_1}: k \frac{\partial V^d(X(D_1), D_1)}{\partial X_1} \frac{\partial X_1(D_1)}{\partial D_1} + k \frac{\partial V^d(X_1(D_1), D_1)}{\partial D_1} \\ + (1 - k) \frac{\partial V_d(X, D_1)}{\partial D_1} = 0$$

Changing notation this is equivalent to:

$$kV_x^d(X_1(D_1), D_1)X'_1(D_1) + kV_d^d(X_1(D_1), D_1)$$

$$(1 - k)V_d^d(X_1, D_1) = 0$$

And since conjectures are fulfilled in equilibrium,

i.e.:  $X_1(D_1) = X_1$ :

$$kV_x^d(X_1(D_1), D_1)X'_1(D_1) + V_d^d(X_1(D_1), D_1) = 0$$

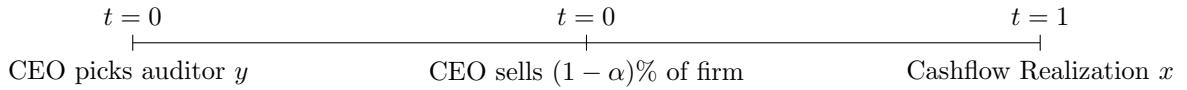
This is an ODE which uniquely describes the signaling schedule  $X_1(D_1)$  which perfectly reveals  $X_1$

$\exists D^*$  st  $D^* \Rightarrow$   
fully revealing  
equilibrium

# Ross (1977)

## II. Signaling (Fully Revealing Signaling Equilibria)

This paper uses a level of maximum debt  $D^*$  as a threshold for CEOs to be able to perfectly signal whether their firm is good (Type A) or bad (Type B).



$$M = \gamma_0 V_0(1 + r) + \gamma_1 \begin{bmatrix} V_1 & \text{if } V_1 \geq D \\ V_1 - C & \text{if } V_1 < D \end{bmatrix}$$

$M$  is CEO compensation,  $V_1$  is future firm value  $\rightarrow$

$V_0 = V_1/(1 + r)$  is current firm value  $\rightarrow$

$\gamma_0$  and  $\gamma_1$  are positive constants  $\rightarrow$

$r$  is the one period interest rate  $\rightarrow$

$D$  is the face value of firm debt  $\rightarrow$

C: penalty paid if bankruptcy occurs ( $V_1 < D$ )  $\rightarrow$

Two types of firms: good type A & bad type B  $\rightarrow$

A firms are worth more than Type B:  $V_a > V_b$   $\rightarrow$

$D^*$  is max debt B can have without bankruptcy  $\rightarrow$

CEO uses  $D^*$  as a threshold to signal firm type  $\rightarrow$

$D > (<)D^*$  means CEO signals it is type A(B)  $\rightarrow$

CEO earns more if he accurately signals type:  $\rightarrow$

A truth:  $M_a = \gamma_0 V_{1a} + \gamma_1 V_{la}$  if  $D^* < D \leq V_{la}$   $\rightarrow$

A lie:  $M_a = \gamma_0 V_{1b} + \gamma_1 V_{la}$  if  $D < D^*$   $\rightarrow$

B lie:  $M_b = \gamma_0 V_{1a} + \gamma_1 (V_{1b} - C)$  if  $D^* < D \leq V_{1a}$   $\rightarrow$

B truth:  $M_b = \gamma_0 V_{1b} + \gamma_1 V_{lb}$  if  $D < D^*$   $\rightarrow$

Clearly, CEOs of firms of type A will tell the truth by choosing  $D > D^*$  because they earn more compensation if they do because  $V_b < V_a$ .

For CEOs of type B firms, compensation must be set s.t. they earn more by correctly choosing  $D < D^*$ .

This is true when:  $\gamma_0 (V_{1a} - V_{1b}) < \gamma_1 C$

or  $\gamma_0 V_{1a} + \gamma_1 (V_{1b} - C) < \gamma_0 V_{1b} + \gamma_1 V_{1a}$

$\exists D^*$  st  $D^* \Rightarrow$

fully revealing equilibrium

Compensation gain from lying must be less than the compensation loss. The LHS is the B CEO's share of gain from initially lying that firm is a type A.

The RHS is the penalty due to bankruptcy.

Unsuccessful firms don't have enough cash to avoid bankruptcy if  $D > D^*$  and lying CEOs will earn less.

**Thus the signaling equilibrium is fully revealing.**

## Spence (1973)

### II. Signaling (Fully Separating Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



Spence (1973) shows how an unknown characteristic can be publicly revealed through the use of a signal. **Note:** The model actually doesn't have multiple dates, but I added the above for clarity.

## 1 Perfectly Revealing/Separating Signaling Equilibrium

Here we show a set of employer conjectures on education of workers that: (1) is fulfilled by the worker's optimal action, and (2) allows for education level to perfectly reveal information on worker's productivity.

### Show this setting leads to a Fully Separating Equilibrium.

A worker, knowing his MPL, will choose education level of:

$$\begin{cases} 0 \text{ if } w(MPL^c|\text{educ}^{y^*}) - c(y^*|\text{MPL}) < w(MPL^c|\text{educ}^0) - c(0|\text{MPL}) \\ y^* \text{ if } w(MPL^c|\text{educ}^{y^*}) - c(y^*|\text{MPL}) \geq w(MPL^c|\text{educ}^0) - c(0|\text{MPL}) \end{cases}$$

This will be a fully revealing separating equilibrium if:

$$\text{For } MPL = 1 \text{ worker : } 2 - y^* < 1 - 0 \Rightarrow 1 < y^*$$

$$\text{For } MPL = 2 \text{ worker : } 2 - y/2^* > 1 - 0 \Rightarrow 2 > y^*$$

Thus conjectures will be fulfilled if  $1 < y^* < 2$

### OUTCOME:

- All workers get either exactly 0 or exactly  $y^*$  units of education

- The signaling schedule is thus:
 
$$\begin{cases} MPL^c = 1 \text{ if } y < y^* \\ MPL^c = 2 \text{ if } y \geq y^* \end{cases}$$

- The signal perfectly reveals a worker's  $MPL$ .

### Fully Separating Equilibrium will result

from value of  $y^*$  in  $1 < y^* < 2$ . QED.

- Risk neutral agents →
- Worker's marginal prod. of labor:  $MPL \in \{1, 2\} \rightarrow$
- Worker knows his  $MPL$ ; employer doesn't →
- Worker population  $\text{Prob}(MPL = 1) = q \rightarrow$
- Worker population  $\text{Prob}(MPL = 2) = 1 - q \rightarrow$
- cost of  $y$   $\text{educ}^y$  units:  $c(y|\text{MPL} = 1) = y \rightarrow$
- cost of  $y$   $\text{educ}^y$  units:  $c(y|\text{MPL} = 2) = y/2 \rightarrow$
- Employer conjecture:  $\begin{cases} MPL^c = 1 \text{ if } y < y^* \\ MPL^c = 2 \text{ if } y \geq y^* \end{cases} \rightarrow$
- Employer offers wage:  $w(MPL^c|\text{educ}) = MPL^c \rightarrow$
- We want a **fully separating equilibrium** →

**Note for section 1:** The inverse relation between productivity and cost of education is a necessary condition for education to serve as a signal for productivity.

## 2 Pooling Equilibrium

A pooling equilibrium can exist in which it would be suboptimal for any worker to get education:

	A worker, knowing his MPL, will choose education level of: $\begin{cases} 0 & \text{if } w(MPL^c \text{educ}^{y^*}) - c(y^* MPL) < w(MPL^c \text{educ}^0) - c(0 MPL) \\ y^* & \text{if } w(MPL^c \text{educ}^{y^*}) - c(y^* MPL) \geq w(MPL^c \text{educ}^0) - c(0 MPL) \end{cases}$
Risk neutral agents →	This will be a pooling equilibrium if:
Worker's marginal prod. of labor: $MPL \in \{1, 2\} \rightarrow$	For $MPL = 1$ worker : $2 - y^* < (2 - q) - 0 \Rightarrow y^* > q$
Worker knows his $MPL$ ; employer doesn't →	For $MPL = 2$ worker : $2 - y^*/2 < (2 - q) - 0 \Rightarrow y^* > 2q$
Worker population $\text{Prob}(MPL = 1) = q \rightarrow$	Thus conjectures will be fulfilled if $y^* > 2q$
Worker population $\text{Prob}(MPL = 2) = 1 - q \rightarrow$	<b>OUTPUT RESULTS:</b>
Thus $E[MPL] = q(1) + (1 - q)(2) = 2 - q \rightarrow$	· All workers get either exactly 0 education.
cost of $y$ <b>educ</b> <sup>y</sup> units: $c(y MPL = 1) = y \rightarrow$	· All workers are better off than in the separating equilibrium:
cost of $y$ <b>educ</b> <sup>y</sup> units: $c(y MPL = 2) = y/2 \rightarrow$	$\begin{cases} \text{For } MPL = 1 \text{ worker: } 2 - q > 1 \\ \text{For } MPL = 2 \text{ worker: } 2 - q > 2 - y^*/2 \end{cases},$
New emp. conj.: $\begin{cases} y < y^* : MPL^c = E[MPL] = 2 - q \\ y \geq y^* : MPL^c = 2 \end{cases} \rightarrow$	<b>Pooling Equilibrium will result from value of <math>y^*</math> in <math>y^* &gt; 2q</math>. QED.</b>
Employer offers wage: $w(MPL^c \text{educ}) = MPL^c \rightarrow$	
We want a <b>pooling equilibrium</b> →	

## Alternative Pooling Equilibrium

Risk neutral agents →  
 Worker's marginal prod. of labor:  $MPL \in \{1, 2\}$  →  
 Worker knows his  $MPL$ ; employer doesn't →  
 Worker population  $Prob(MPL = 1) = q$  →  
 Worker population  $Prob(MPL = 2) = 1 - q$  →  
 Thus  $E[MPL] = q(1) + (1 - q)(2) = 2 - q$  →  
 cost of  $y$   $\text{educ}^y$  units:  $c(y|MPL = 1) = y$  →  
 cost of  $y$   $\text{educ}^y$  units:  $c(y|MPL = 2) = y/2$  →  
 Alt. emp. conj.:  $\begin{cases} y < y^*: MPL^c = 1 \\ y \geq y^*: MPL^c = E[MPL] = 2 - q \end{cases}$  →  
 Employer offers wage:  $w(MPL^c|\text{educ}) = MPL^c$  →  
 We want a **pooling equilibrium** →

A worker, knowing his  $MPL$ , will choose education level of:

$$\begin{cases} 0 \text{ if } w(MPL^c|\text{educ}^{y^*}) - c(y^*|MPL) < w(MPL^c|\text{educ}^0) - c(0|MPL) \\ y^* \text{ if } w(MPL^c|\text{educ}^{y^*}) - c(y^*|MPL) \geq w(MPL^c|\text{educ}^0) - c(0|MPL) \end{cases}$$

This will be **another** pooling equilibrium if:

$$\text{For } MPL = 1 \text{ worker : } (2 - q) - y^* > 1 - 0$$

$$\Rightarrow 2 - q - y^* > 1$$

$$\text{For } MPL = 2 \text{ worker : } (2 - q) - y^*/2 > 1 - 0$$

$$\Rightarrow 2 - q - y^*/2 > 1$$

Thus conjectures will be fulfilled if  $2 - q - y^* > 1$

### OUTPUT RESULTS:

- All workers get either exactly  $y^*$  education.
- and are worse off than in previous pooling equilibrium:
 
$$\begin{cases} \text{For } MPL = 1 \text{ worker: } (2 - q) - y^* < (2 - q) \\ \text{For } MPL = 2 \text{ worker: } (2 - q) - y^* < (2 - q) \end{cases}$$

If however an employer offers  $(2 - q)$  regardless of education, all workers would take that option instead, and that employer would get the same expected level of productivity.

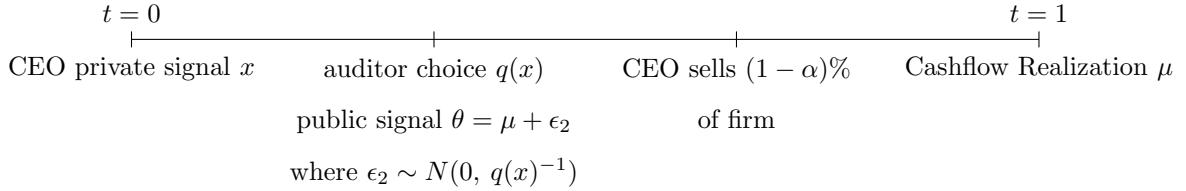
### Pooling Equilibrium will result

**from value of  $y^*$  in  $2 - q - y^* > 1$ . QED.**

# Titman & Trueman (1986)

## II. Signaling (Fully Separating Signaling Equilibria)

Based on Winter 2019 Accounting Theory (simplified) class notes on the paper (Prof. Trueman)



## 1 Fully Revealing Signaling Equilibrium

The goal of this paper is to show how the quality of the chosen auditor can signal the firm's value.

**Show  $\exists$  a Fully Separating Signaling Equilibria by proving  $\exists$  a signal inference schedule**

CEO has CARA coef. of risk-aversion  $a \rightarrow$   
 Investors are risk-neutral  $\rightarrow$

CEO sells  $(1 - \alpha)\%$  of firm &  $\alpha$  is exogenous  $\rightarrow$   
 $\exists$  risk-free asset yielding 0%  $\rightarrow$

Diffuse priors of firm value  $\mu \rightarrow$

CEO gets private signal  $x = \mu + \epsilon_1 \rightarrow$   
 public signal  $\theta = \mu + \epsilon_2 \rightarrow$   
 $\epsilon_1 \sim N(0, h^{-1}) \perp \epsilon_2 \sim N(0, q^{-1}) \rightarrow$

precision  $q$  depends on CEO's auditor quality  $\rightarrow$   
 auditor quality is function of CEO's  $x : q(x) \rightarrow$

mkt posterior of  $\mu$  given auditor quality:  $f(q*) \rightarrow$   
 $E[U^{CEO}(W)|x] = \int \int U(W)g(\theta, \mu|x)d\mu d\theta \rightarrow$

$W = W_0 - I + (1 - \alpha)E^{mkt}[\mu|\theta, q] - c(q) + \alpha\mu \rightarrow$   
 where:  $W_0$  is CEO's initial wealth  $\rightarrow$

$c(q)$  is cost of auditor of quality  $q$  &  $c'(q) > 0 \rightarrow$

and  $I$  is amount CEO invested in the project  $\rightarrow$

CEO's objective function is to maximize:  
 $E[W|x] - \frac{a}{2}var(W|x)$  given his CARA utility

Given that priors are diffuse:  $E^{CEO}[\mu|x] \sim N(x, h^{-1})$   
 and firm value:  $E^{mkt}[\mu|\theta, q^*] = \frac{q^*}{q^*+h}\theta + \frac{h}{q^*+h}f(q^*)$

Notes simply give **equilibrium** signaling schedule:  

$$f(q^*) = \frac{\int_{q_{min}}^{q^*} c'(y)(y+h)dy}{(1-\alpha)h} + \frac{a(1+\alpha)[\ln(q^*+h) - \ln(q_{min}+h)]}{2h} + z$$
from which we can tell that  $f(q^*)$  is  $\nearrow$  in  $x$  (in equilib.)  
 $(z$  is not elaborated on in the notes).

CEO's expectation of market price in equilibrium:  
 $E^{CEO}[E^{mkt}[\mu|\theta, q^*]] = \frac{q^*}{q^*+h}E^{CEO}[\theta|q^*] + \frac{h}{q^*+h}f(q^*)]$   
 using consistency condition:  $E^{CEO}[\theta|q^*] = f(q^*)$   
 gives  $E^{CEO}[E^{mkt}[\mu|\theta, q^*]] = f(q^*)$

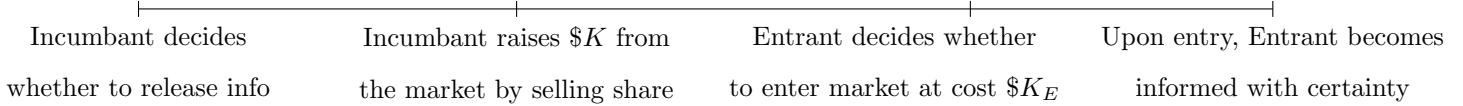
$f(q^*)$  is  $\nearrow$  in  $x$   
 in equilibrium.

The equilibrium signaling schedule shows that the CEO with the least favorable info consistent with taking on the project signals with minimum precision  $q_{min}$ .

# Darrough & Stoughton (1990)

## III. Voluntary Disclosure (Partial Signaling Equilibria)

This paper looks at equilibria in the context of an Incumbent who has to trade off between losing profit share to a potential Entrant by revealing info to raise money in the market for a profitable project.

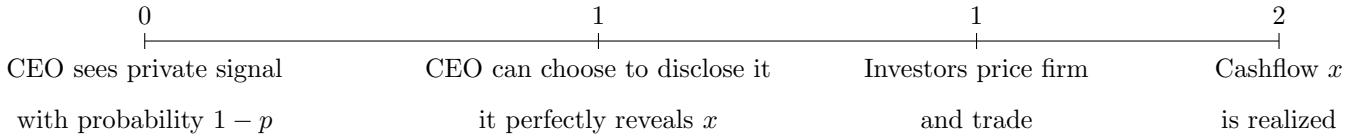


If Entrant $E$ enters Monopolist $M$ 's industry: →	$\exists \text{ A fully revealing equilibrium if } q > \mu \Rightarrow e = 1$ defined by disclosure/entry policies $\{d_1 \in [0, 1], d_2 = 1, e = 1\}$		
then $M$ becomes Incumbant $I$ in a duopoly →	It is sufficient to show that both types of $M$ ( $L$ & $H$ ) would prefer diclosure & $q = 1$ is a consistent belief:		
To finance project $M \in \{\text{type } H, L\}$ raises \$ $K$ →	for $L : V(d) = \begin{cases} \underline{I}_M - K & \text{if } d = L \text{ (} E \text{ won't enter)} \\ (1 - \frac{K}{V}) \underline{\Pi}_I & \text{if } d = ND \end{cases}$		
$M$ can credibly ( $d$ )isclose its type before →	$V = e(\underline{\Pi}_I + q(\bar{\Pi}_I - \underline{\Pi}_I)) + (1 - e)(\underline{I}_M + q(\bar{\Pi}_M - \underline{I}_M)) = \bar{\Pi}_I$		
selling fraction $\alpha(d) = \frac{K}{V(d)}$ of the firm where →	Thus $K \leq \frac{\bar{\Pi}_I(\underline{I}_M - \underline{I}_I)}{\bar{\Pi}_I - \underline{I}_I}$ knowing that $\frac{(\Pi_M - \Pi_I)}{(\Pi_I - \Pi_I)} \geq 1$		
$V(d)$ : market's valuation given available info →	And type $H$ cannot prevent $E$ from entering given $e = 1$ :		
Afterwards $E$ decides whether to enter →	$\exists \text{ A pooling equilibrium if } q < \mu \Rightarrow e = 0$ defined by disclosure/entry policies $\{d_1 = 0, d_2 = 0, e = 0\}$		
It costs \$ $K_E$ for $K$ to enter $M$ 's industry →	Type $H$ will prefer $ND$ if $(1 - \frac{K}{V}) \bar{\Pi}_M \geq \bar{\Pi}_I - K$		
Upon entry $M$ 's revenue falls from $\Pi_M$ to $\Pi_I$ →	$K \leq \frac{V(\bar{\Pi}_M - \bar{\Pi}_I)}{\bar{\Pi}_M - V}$ holds since we know that $\bar{\Pi}_I \leq \underline{\Pi}_M < V$		
and $E$ learns $M$ 's type if previously undisclosed →	Investor posteriors become: $q = \frac{p(1-d_1)}{p(1-d_1)+(1-p)(1-d_2)} = p$		
then $E$ earns $\bar{\Pi}_E$ . For $i = M, I, E$ profits $\Pi_i$ are: →	And since $e = 0$ , $L$ types will prefer $ND$ with certainty.:		
$\bar{\Pi}_i$ (if $M$ was type $H$ ) > $\underline{\Pi}_i$ (if $M$ was type $L$ ) →	$\exists \text{ A partially revealing equilibrium if } q = \mu$ defined by policies: $\{d_1 = 0, d_2 = \frac{\mu-p}{(1-p)\mu}, e = e^* \in (0, 1)\}$		
Assume $\underline{\Pi}_M \geq \bar{\Pi}_I$ and $\bar{\Pi}_I > \underline{\Pi}_I > K > 0$ →	This will only occur if posterior given $ND$ leave $E$ indifferent to entering, meaning $q = \mu$ . Since here only the type $H$ incumbents always choose $ND$ : $q > p$ . This yields the necessary condition that $\mu > p$ . Note that when $d_2 = \frac{\mu-p}{(1-p)\mu}$ , the posterior is $q = \frac{p}{p+(1-\frac{\mu-p}{(1-p)\mu})(1-p)} = \frac{p}{p+1-p-\frac{\mu-p}{\mu}} = \frac{p}{p/\mu} = \mu$		
$d_j$ is the probability $M$ discloses given type →	$V = e^*(\underline{\Pi}_I + \mu(\bar{\Pi}_I - \underline{\Pi}_I)) + (1 - e^*)(\underline{\Pi}_M + \mu(\bar{\Pi}_M - \underline{\Pi}_M))$		
where $d_1 = P(d = H H)$ , $d_2 = P(d = L L)$ →	Since $L$ is indifferent between $d = L$ & $ND$ , $e^*$ much be s.t.: $\underline{\Pi}_M - K = (1 - \frac{K}{V})(e^*\underline{I}_I + (1 - e^*)\underline{\Pi}_M)$ . Given $\Pi_M \geq \bar{\Pi}_I$ , $\exists$ entry probability $e^*$ s.t. above holds. To verify that at $e^*$ , type $H$ prefers $ND$ : $\bar{\Pi}_I - K < (1 - \frac{K}{V})(e^*\bar{\Pi}_I + (1 - e^*)\bar{\Pi}_M)$		
$e = P(\text{entry} ND)$ where $ND = \text{No disclosure}$ →	But: $\bar{\Pi}_I - K \leq \underline{\Pi}_M - K = (1 - \frac{K}{V})(e^*\underline{\Pi}_I + (1 - e^*)\underline{\Pi}_M)$		
$E$ enters if $d = H$ since $\bar{\Pi}_E > K_E > \underline{\Pi}_E > 0$ →	$(1 - \frac{K}{V})(e^*\underline{I}_I + (1 - e^*)\underline{\Pi}_M) < (1 - \frac{K}{V})(e^*\bar{\Pi}_I + (1 - e^*)\bar{\Pi}_M)$		
$p = P(H)$ is prior belief $M$ is $H$ type while →			
$q = P(H d = ND)$ is posterior given $ND$ →			
Sequential equilibrium defined by $\{d_1, d_2, e, q\}$ →			
Even with no disclosure ( $ND$ ), $E$ would enter →			
if: $E(\Pi_E ND) = q\bar{\Pi}_E + (1 - q)\underline{\Pi}_E > K_E$ , or if: →			
$q > \frac{K_E - \underline{\Pi}_E}{\bar{\Pi}_E - \underline{\Pi}_E} \equiv \mu$ can be intepreted as entry cost →			
$e = 1$ if $q > \mu$ , $q = \mu$ if $e \in (0, 1)$ , $e = 0$ if $q < \mu$ →			
$V(d) = E(\Pi_I d, e)$ : value of $I$ given $E$ 's entry →			
$M$ optimizes $\max_{d \in \{D, ND\}} E(\Pi_I   d, e) - K$ →			
Note: $V(d) = e(1 - \frac{K}{V})\bar{\Pi}_I + (1 - e)(1 - \frac{K}{V})\underline{\Pi}_M$ →			
is the valuation under $ND$ , where $V = e(\underline{\Pi}_I + q(\bar{\Pi}_I - \underline{\Pi}_I)) + (1 - e)(\underline{I}_M + q(\bar{\Pi}_M - \underline{I}_M))$ →			
Show that all 3 types of signaling equilibria exist →			
depending on whether $q > \mu$ , $q = \mu$ or $q < \mu$ →			

Dye (1985) + Jung & Kwon (1988)

### III. Voluntary Disclosure (Partial Signaling Equilibria)

Here partial disclosure can arise even when there are no disclosure costs.



- Firm has risky cashflow with prior mean  $\mu \rightarrow$
- Investors are risk **neutral**  $\rightarrow$
- 0, CEO sees private signal with prob.  $1 - p \rightarrow$ 
  - If received, it perfectly reveals  $x \rightarrow$
  - CEO can choose to credibly disclose it  $\rightarrow$
  - Disclosure is costless  $\rightarrow$
  - EO cannot credibly convey he's uninformed  $\rightarrow$
  - CEO's goal is to maximize firm valuation  $\rightarrow$
  - Show  $\exists$  a partially revealing equilibrium  $\rightarrow$

If CEO discloses  $x$ , t=1 price will equal  $x$

otherwise price considering that:

$$Prob(\text{CEO uninformed}) \equiv p$$

$$Prob(\text{CEO withholding}) \equiv (1 - p)F(x)$$

$$Prob(\text{CEO discloses}) = (1 - p)[1 - F(x)]$$

where  $F(\hat{x}) = \text{Prob}(x \leq \hat{x}) = \int_{\underline{x}}^{\hat{x}} dF(x)$

Now given non-disclosure (*ND*):

$$\begin{aligned} E[x|ND] &= \frac{p}{p+(1-p)P(x < \hat{x})} E[x] + \frac{(1-p)P(x < \hat{x})}{p+(1-p)P(x < \hat{x})} E[x|x < \hat{x}] \\ E[x|ND] &= \frac{p}{p+(1-p)F(\hat{x})} \mu + \frac{(1-p)F(\hat{x})}{p+(1-p)F(\hat{x})} \int_{\underline{x}}^{\hat{x}} \frac{xdF(x)}{F(\hat{x})} \\ &= \frac{p}{p+(1-p)F(\hat{x})} \mu + \frac{(1-p)F(\hat{x})}{p+(1-p)F(\hat{x})} \int_{\underline{x}}^{\hat{x}} x dF(x) \end{aligned}$$

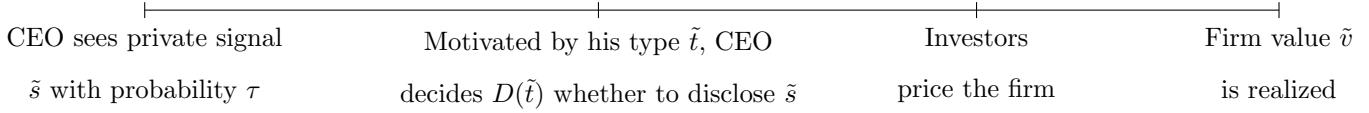
Authors then show threshold level of  $\hat{x}$  is unique.

$\exists$  a unique  
 $\hat{x}^*$  s.t.  $\hat{x}^*$   
 $\Rightarrow$  partially  
revealing  
equilibrium

# Einhorn (2007)

## III. Voluntary Disclosure (Partial Signaling Equilibria)

This paper departs from the typical setting of voluntary disclosure in a rational expectations model by the addition of an assumption that investors are uncertain about the manager's reporting objective.



We start by case where  $\tau = 1$ , in equilibrium :  $\forall s \in \mathbb{R}$  :

$$D(\text{inf}, s) = \hat{D}(\text{inf}, s) = \begin{cases} \text{dis if } s \geq s^* \\ \text{nd if otherwise} \end{cases}$$

$$D(\text{def}, s) = \hat{D}(\text{def}, s) = \begin{cases} \text{dis if } s \leq s^* \\ \text{nd if otherwise} \end{cases}$$

$$P(\text{dis}, s) = \hat{P}(\text{dis}, s) = \mu + \rho\sigma s$$

$$P(\text{nd}, s) = \hat{P}(\text{nd}, s) = E[\tilde{v} | \text{nd}] = \mu + \rho\sigma E[\tilde{s} | \text{nd}]$$

$$= \mu + \rho\sigma [\text{Prob}(\text{inf}|\text{nd})E[\tilde{s} | \text{inf}, \text{nd}] + \text{Prob}(\text{def}|\text{nd})E[\tilde{s} | \text{def}, \text{nd}]]$$

$$= \mu + \rho\sigma [\text{Prob}(\text{inf}|\text{nd})E[\tilde{s} | \tilde{s} \leq s^*] + \text{Prob}(\text{def}|\text{nd})E[\tilde{s} | \tilde{s} \geq s^*]]$$

$$\text{Prob}(\text{inf}|\text{nd}) = \frac{\pi\Phi(s^*)}{\pi\Phi(s^*) + (1-\pi)(1-\Phi(s^*))}; E[\tilde{s} | \tilde{s} \leq s^*] = -\frac{z(s^*)}{\Phi(s^*)}$$

$$\text{Prob}(\text{def}|\text{nd}) = \frac{(1-\pi)(1-\Phi(s^*))}{\pi\Phi(s^*) + (1-\pi)(1-\Phi(s^*))}; E[\tilde{s} | \tilde{s} \geq s^*] = \frac{z(s^*)}{1-\Phi(s^*)}$$

$$P(\text{nd}, s) = \mu + \rho\sigma \left[ \frac{-\pi\Phi(s^*)z(s^*)(1-\Phi(s^*))/\Phi(s^*) + (1-\pi)(1-\Phi(s^*))z(s^*)}{(1-\Phi(s^*))(\pi\Phi(s^*) + (1-\pi)(1-\Phi(s^*)))} \right]$$

$$= \mu + \rho\sigma \frac{-\pi z(s^*) + (1-\pi)z(s^*)}{\pi\Phi(s^*) + (1-\pi)(1-\Phi(s^*))} = \mu - \rho\sigma \frac{(2\pi-1)z(s^*)}{(2\pi-1)\Phi(s^*) + (1-\pi)}$$

$\exists$  a

unique

$\hat{s}^*$

s.t.  $\hat{s}^*$

$\Rightarrow$

partial

reveal

equilib.

With uncertain info endowment  $0 < \tau \leq 1$ , the following changes:

$$\text{Prob}(\text{inf}|\text{nd}) = \frac{\tau\pi\Phi(s^*)}{(1-\tau) + \tau\pi\Phi(s^*) + \tau(1-\pi)(1-\Phi(s^*))}$$

$$\text{Prob}(\text{def}|\text{nd}) = \frac{\tau(1-\pi)(1-\Phi(s^*))}{(1-\tau) + \tau\pi\Phi(s^*) + \tau(1-\pi)(1-\Phi(s^*))}$$

$$P(\text{nd}, s) = \mu - \rho\sigma \frac{\tau(2\pi-1)z(s^*)}{\tau(2\pi-1)\Phi(s^*) + (1-\tau)\pi}$$

Disclosure costs: if  $\tau = 0$  & instead cost to acquire  $\tilde{s}$  is  $c > 0$  then

the points of indifference for an inflating and deflating CEO are:

$$P(\text{nd}, s_{\text{inf}}^*) = P(\text{inf}, s_{\text{def}}^*) - c; -P(\text{nd}, s_{\text{def}}^*) = -P(\text{dis}, s_{\text{def}}^*) - c$$

$$\text{For nd: } P(\text{nd}, s_{\text{inf}}^*) = P(\text{nd}, s_{\text{def}}^*) \Rightarrow P(\text{dis}, s_{\text{inf}}^*) - P(\text{dis}, s_{\text{def}}^*) = 2c$$

$$P(\text{nd}, s) : (\mu + \rho\sigma s_{\text{inf}}^*) - (\mu + \rho\sigma s_{\text{def}}^*) = 2c \Leftrightarrow \rho\sigma(s_{\text{inf}}^* - s_{\text{def}}^*) = 2c$$

$$s_{\text{inf}}^* = s_{\text{def}}^* + \frac{2c}{\rho\sigma} \quad \text{then:}$$

$$\text{Prob}(\text{inf}|\text{nd}) = \frac{\pi\Phi(s_{\text{inf}}^*)}{\pi\Phi(s_{\text{inf}}^*) + (1-\pi)(1-\Phi(s_{\text{inf}}^*))}; E[\tilde{s} | \tilde{s} \leq s_{\text{inf}}^*] = -\frac{z(s_{\text{inf}}^*)}{\Phi(s_{\text{inf}}^*)}$$

$$\text{Prob}(\text{def}|\text{nd}) = \frac{(1-\pi)(1-\Phi(s_{\text{def}}^*))}{\pi\Phi(s_{\text{def}}^*) + (1-\pi)(1-\Phi(s_{\text{def}}^*))}; E[\tilde{s} | \tilde{s} \geq s_{\text{def}}^*] = \frac{z(s_{\text{def}}^*)}{1-\Phi(s_{\text{def}}^*)}$$

$$P(\text{nd}, s) = \mu - \rho\sigma \frac{\tau\pi z(s_{\text{inf}}^*) - \tau(1-\pi)z(s_{\text{def}}^*)}{1-\tau + \tau\pi\Phi(s_{\text{inf}}^*) + \tau(1-\pi)(1-\Phi(s_{\text{def}}^*))}$$

In equilibrium,  $-P(\text{nd}, s_{\text{def}}^*) = -P(\text{dis}, s_{\text{def}}^*) - c$ :

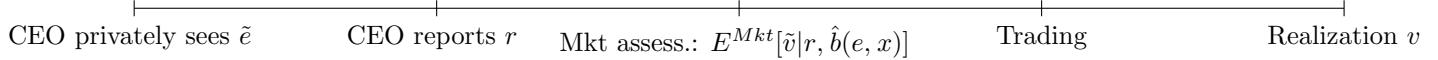
$$s_{\text{def}}^* + \frac{\tau\pi z(s_{\text{def}}^* + 2c/\rho\sigma) - \tau(1-\pi)z(s_{\text{def}}^*)}{1-\tau + \tau\pi\Phi(s_{\text{def}}^* + 2c/\rho\sigma) + \tau(1-\pi)(1-\Phi(s_{\text{def}}^*))} + \frac{c}{\rho\sigma} = 0.$$

Solving this gives unique  $s_{\text{def}}$  yielding a partial equilibrium. QED.

# Fischer & Verrechia (2000)

## III. Voluntary Disclosure (Fully Revealing Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)



This paper shows that uncertainty regarding managerial incentives can result in biasing activity by the managers that affects firm value.

**Show that the coefficients on the linear conjectures converge to specific economic parameter based terms, thus proving that  $\exists$  a Fully Separating Equilibrium in this setting**

- CEO is risk **neutral** →
- Public priors on  $\tilde{v} \sim N(0, \sigma_v^2)$  →
- CEO privately sees signal  $\tilde{e} = \tilde{v} + \tilde{n}$  →  
where  $\tilde{n} \sim N(0, \sigma_n^2)$  →
- Commonly known that  $\tilde{v} \perp \tilde{n}$  →
- After seeing  $\tilde{e} = e$ : CEO reports  $r$  →  
 $r = e + b$  where  $b$  is CEO report bias →
- Given  $x$  CEO maxes:  $x\hat{P} - \frac{cb^2}{2}$  →  
where random event  $\tilde{x} \sim N(\mu_x, \sigma_x^2)$  →  
has outcome only CEO sees:  $\tilde{x} = x$  →
- $c > 0$  is publicly known cost parameter →  
 $cb^2/2$  is CEO's known report bias cost →
- CEO choice of bias is  $b(e, x)$  →
- $\hat{b}(e, x)$ : Mkt conjecture of CEO's  $b(e, x)$  →  
has linear form  $\hat{b}(e, x) = \hat{\lambda}_e e + \hat{\lambda}_x x + \hat{\delta}$  →
- In equilibrium:  $\hat{b}(e, x) = b(e, x) \quad \forall \{e, x\}$  →
- Mkt assess:  $P(r) = E^{Mkt}[\tilde{v} | r, \hat{b}(e, x)]$  →
- $\hat{P}$ : CEO conject. of market pricing rule →  
has linear form  $\hat{P}(r) = \hat{\beta}r + \hat{\alpha}$  →
- In equil.:  $\hat{P}(r) = P(r) = \beta r + \alpha \quad \forall r$  →

CEO's conjecture of market price dynamics given his report  $r$ :  $P = \hat{\beta}r + \hat{\alpha} = \hat{\beta}e + \hat{\beta}b + \hat{\alpha}$

CEO maximizes  $x(\hat{\beta}e + \hat{\beta}b + \hat{\alpha}) - \frac{cb^2}{2}$  w.r.t.  $b$   
 $0 = \hat{\beta}x - cb \Rightarrow b(e, x) = \frac{\hat{\beta}}{c}x$  : optimal bias function.

Knowing this, investors estimate the coefficients of their conjecture of CEO's bias function:  $b(e, x) = \lambda_e e + \lambda_x x + \delta$   
as:  $\lambda_e = 0, \quad \lambda_x = \frac{\hat{\beta}}{c}, \quad \delta = 0$

Given this, investors assess value of firm as:

$$E[\tilde{v} | r, \hat{b}(e; x)] = E[\tilde{v} | r] = E[\tilde{v} | e + \hat{\lambda}_x x]$$

$$= E[\tilde{v} | \tilde{v} + \tilde{n} + \hat{\lambda}_x x] = E[\tilde{v}] + \frac{\text{cov}(\tilde{v}, \tilde{v} + \tilde{n} + \hat{\lambda}_x x)}{\text{var}(\tilde{v} + \tilde{n} + \hat{\lambda}_x x)}(r - E[r])$$

$$E[\tilde{v} | r, \hat{b}(e, x)] = \frac{\sigma_v^2}{\sigma_v^2 + \sigma_n^2 + \hat{\lambda}_x^2 \sigma_x^2}(r - \hat{\lambda}_x \mu_x)$$

Knowing this, CEO estimates the coefficients in his conjecture of investor's pricing strategy,  $\hat{P}(r) = \hat{\beta}r + \hat{\alpha}$ ,  
as  $\beta = \frac{\sigma_v^2}{\sigma_v^2 + \sigma_n^2 + \hat{\lambda}_x^2 \sigma_x^2}$  and  $\alpha = -\beta \hat{\lambda}_x \mu_x$

Plugging  $\lambda_x = \frac{\beta}{c}$ , in equilibrium,  $\beta = \frac{\sigma_v^2}{\sigma_v^2 + \sigma_n^2 + (\frac{\beta}{c})^2 \sigma_x^2}$   
Giving:  $\beta^3 \sigma_x^2 + \beta (\sigma_v^2 + \sigma_n^2) c^2 - \sigma_v^2 c^2 = 0$

This gives us a solution that also lets us solve for  $\alpha$  as a function of economic parameters, proving that  $\exists$  a Fully Separating Equilibrium in this setting. QED.

# Hughes & Pae (2004)

## III. Voluntary Disclosure (Partially Revealing Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)

This paper explores the real effects of partial disclosure equilibria on the firm.

### Mandatory Disclosure (subscript: $man$ )

CEO decides whether to pay $c$ to be $CEO_I$ & privately know precision $s$ perfectly	CEO observes private signal $y = x + \epsilon$ where $\epsilon \sim N(0, s^{-1})$	It's <b>mandatory</b> that CEO releases all private info he acquired	To finance project CEO sells firm for $E^{mkt}[x info]$	Project realization $x$
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Show that this fully revealing equilibrium has real effects on whether the project is undertaken

Let  $D$ = disclosure of  $s$ ,  $ND$ = no disclosure of  $s$ .

Under the **mandatory** disclosure case we have:

$$Prob(CEO_I \Leftrightarrow D) = \lambda \text{ & } Prob(CEO_U \Leftrightarrow ND) = 1 - \lambda$$

$$E^{CEO}(x|y, D) = \frac{s}{s+h}y + \frac{h}{s+h}\theta$$

$$E^{CEO}(x|y, ND) = E[\frac{s}{s+h}|y, ND]y + (1 - E[\frac{s}{s+h}|y, ND])\theta$$

CEO & traders are risk **neutral** →

CEO has project that requires outlay  $k$  →

It's worthless without the investment →

Project has positive NPV ( $\theta > k$ ) →

Public priors on PV:  $x \sim N(\theta, h^{-1})$  →

CEO sells firm to investors to raise  $k$  →

Mandatory disclos. mkt posterior:  $E^{mkt}[x|info]$  →

Before sale: CEO sees private signal:  $y = x + \epsilon$  →

where  $\epsilon \sim N(0, s^{-1})$ ; no one knows value of  $s$  →

Before seeing  $y$ : CEO can buy info on  $s$  →

This info costs  $c$  and perfectly reveals  $s$  →

$c$  is random & only known to the CEO →

Denote CEO who got info on  $s$ :  $CEO_I$  →

Denote CEO who didn't get info on  $s$ :  $CEO_U$  →

All disclosures are assumed truthful →

$Prob(CEO_I) = \lambda$  is exogenously imposed →

$E[\frac{s}{s+h}|y, ND]$ : posterior of  $\frac{s}{s+h}$  given  $y$  &  $ND$  →

it's small s.t.:  $E[\frac{s}{s+h}|y, ND] << E[\frac{h}{s+h}|y, ND]$  →

This is a fully separating equilibrium →

Then investor expectation of net value of the firm is:

#### Case 1: Given Disclosure of $s$ (D)

$$\begin{aligned} \text{Here } CEO_I \Leftrightarrow D, \text{ so } E^{mkt}[x - k|y, D] &= \frac{s}{s+h}y + \frac{h}{s+h}\theta - k \\ &= \frac{s}{s+h}y + (1 - \frac{s}{s+h})\theta - k = \frac{s}{s+h}(y - \theta) + (\theta - k) \end{aligned}$$

Real effects

(A) and (B)

(A) Case 1 relations show that project will be taken regardless of the value of  $\frac{s}{s+h}$  if  $y > k$ . But if  $y < k$ , project will be taken only if  $\frac{s}{s+h}$  is small enough

#### Case 2: Given No-Disclosure of $s$ (ND)

$$\begin{aligned} \text{Here } CEO_U \Leftrightarrow ND, \text{ so } E^{mkt}[x - k|y, ND] &= E^{mkt}[\frac{s}{s+h}|y, ND]y + (1 - E^{mkt}[\frac{s}{s+h}|y, ND])\theta - k \\ E^{mkt}[x - k|y, ND] &= E^{mkt}[\frac{s}{s+h}|y, ND](y - \theta) + (\theta - k) \end{aligned}$$

(B) Case 2 relations show that project will be taken regardless of the value of  $E^{mkt}[\frac{s}{s+h}|y, ND]$  if  $y > k$ .

But if  $y < k$ , project will be taken

only if  $E^{mkt}[\frac{s}{s+h}|y, ND]$  is small enough

## Voluntary Disclosure (subscript: $vol$ )

Here investors do not know whether no disclosure ( $ND$ ) means that the CEO did not collect information on  $s$  ( $CEO_U|ND$ ) or if CEO is informed but withholding said information on  $s$  ( $CEO_I|ND$ ).

CEO decides whether to pay $c$ to be $CEO_I$ & privately know precision $s$ perfectly	CEO observes private signal $y = x + \epsilon$ where $\epsilon \sim N(0, s^{-1})$	CEO discloses level of $y$ and decides whether to <b>voluntarily</b> disclose info on $s$ if he acquired it	To finance project CEO sells firm for $E^{mkt}[x info]$	Project realization $x$
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### Show that this partially revealing equilibrium has real effects on whether the project is undertaken

Let  $D$  = disclosure of  $s$ ,  $ND$  = no disclosure of  $s$ .

$$E^{CEO}(x|y, D) = \frac{s}{s+h}y + \frac{h}{s+h}\theta$$

$$E^{CEO}(x|y, ND) = E[\frac{s}{s+h}|y, ND]y + (1 - E[\frac{s}{s+h}|y, ND])\theta$$

Then investor expectation of net value of the firm is:

#### Case 1: Given Disclosure of $s$ (D)

$$\begin{aligned} E_{vol}^{mkt}[x - k|y, D] &= E_{man}^{mkt}[x - k|y, D] = \frac{s}{s+h}y + \frac{h}{s+h}\theta - k \\ &= \frac{s}{s+h}y + (1 - \frac{s}{s+h})\theta - k = \frac{s}{s+h}(y - \theta) + (\theta - k) \end{aligned}$$

(A) Case 1 has the same result as in mandatory disclosure: The relations show that project will be

taken regardless of the value of  $\frac{s}{s+h}$  if  $y > k$ . But if

$y < k$ , project will be taken only if  $\frac{s}{s+h}$  is small enough

→ Real effects

→ (A) and (B)

#### Case 2: Given No-Disclosure of $s$ (ND)

$$E_{vol}^{mkt}[x - k|y, ND]$$

$$= E_{vol}^{mkt}[\frac{s}{s+h}|y, ND]y + (1 - E_{vol}^{mkt}[\frac{s}{s+h}|y, ND])\theta - k$$

$$E_{vol}^{mkt}[x - k|y, ND] = E_{vol}^{mkt}[\frac{s}{s+h}|y, ND](y - \theta) + (\theta - k)$$

(B) Case 2 relations show that project will be taken regardless of the value of  $E_{vol}^{mkt}[\frac{s}{s+h}|y, ND]$  if  $y > k$ .

But if  $y < k$ , project will be taken

only if  $E_{vol}^{mkt}[\frac{s}{s+h}|y, ND]$  is small enough

Now investors aren't sure if CEO is  $CEO_U$  or  $CEO_I$  so they put some weight on  $Prob(CEO_I|ND)$  giving result:

$$E_{man}^{mkt}[\frac{s}{s+h}|y, ND] < E_{vol}^{mkt}[\frac{s}{s+h}|y, ND] \text{ when } y < \theta$$

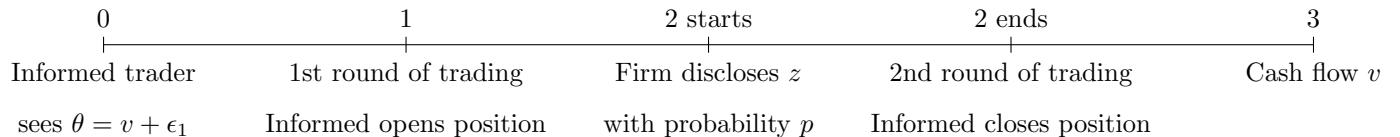
$$E_{man}^{mkt}[\frac{s}{s+h}|y, ND] > E_{vol}^{mkt}[\frac{s}{s+h}|y, ND] \text{ when } y > \theta$$

# McNichols & Trueman (1994)

## III. Voluntary Disclosure (Fully Revealing Signaling Equilibria)

This paper shows the effects of information acquisition and trading before public disclosure.

Basic framework follows Kyle (1985) with 2 changes: (1) informed trader now has a short term holding strategy and (2) after he has made his purchase, there is a possible public signal before he closes his position which he now needs to consider.



risky asset  $v$  prior:  $v \sim N(\bar{v}, \sigma^2) \rightarrow$   
 $t=1$ : informed sees signal  $\theta = v + \epsilon_1 \rightarrow$   
 where  $\epsilon_1 \sim N(0, \sigma_1^2) \rightarrow$

Risk neutral informed  $t=1$  demand:  $x \rightarrow$

Liquidity  $t=1$  demand:  $\mu \sim N(0, \sigma_u^2) \rightarrow$

Market maker sets price  $P_1 = E(v|y) \rightarrow$

where  $y$  is order flow  $y = x + \mu \rightarrow$

$t=2$ : with probability  $p$  firm discloses:  $\rightarrow$

$z = v + \epsilon_2$  where  $\epsilon_2 \sim N(0, \sigma_2^2) \rightarrow$

and  $cov(\epsilon_1, \epsilon_2) = \sigma_{1,2} \rightarrow$

Then both traders close their positions  $\rightarrow$

Market maker sets price  $P_2 = E(v|y, z) \rightarrow$

We start with Kyle (1985)'s framework:  $\rightarrow$

where informed trader's conjecture of  $P_1 = \mu + \lambda y \rightarrow$

depends on mkt maker's conjecture of  $x = \alpha + \beta\theta \rightarrow$

with both knowing each other's conjecture forms  $\rightarrow$

have coeffs:  $\beta = \left[ \frac{\sigma_u^2}{\sigma_v^2 + \sigma_\epsilon^2} \right]^{0.5}$ ;  $\lambda = \frac{\sigma_v^2}{2\sigma_u^2(\sigma_v^2 + \sigma_\epsilon^2)^{0.5}}$ ,  $\rightarrow$

and again  $x = \alpha + \beta\theta = -\beta\bar{v} + \beta\theta = \beta(\theta - \bar{v}) \rightarrow$

$P_1 = E(v | y) = \bar{v} + \lambda y \rightarrow$

if  $z$  not disclosed:  $P_1 = P_2 \rightarrow$

Informed can acquire info on  $\theta$ 's  $\sigma_1^2$  for  $C_1(\frac{1}{\sigma_1^2}) \rightarrow$

### Show $\exists$ Partially Revealing Rational Expectations Equilibria

The new price at  $t=2$  conjecture is:

$$P_2 = E(v | y, z) = \bar{v} + b_1 y + b_2(z - \bar{v})$$

which the paper solves (assuming  $b_2 > 0$ ):

$$b_1 = \frac{(b_2)\sigma_1^2(\sigma^2 + \sigma_1^2)^{1/2}(\sigma_2^2 - \sigma_{12})}{\sigma_u[2(\sigma^2 + \sigma_1^2)(\sigma^2 + \sigma_2^2) - (\sigma^2 + \sigma_{12})^2]}$$

$$b_2 = \frac{\sigma^2(2\sigma_1^2 + \sigma^2 - \sigma_{12})}{2(\sigma^2 + \sigma_1^2)(\sigma^2 + \sigma_2^2) - (\sigma^2 + \sigma_{12})^2}$$

Conjecture coefficients as functions of economic parameters proves that  $\exists$  a Partial Rational Expectations Equilibrium in this setting

Plugging into informed demand's:

problem:  $\max_x E[(P_2 - P_1)x|\theta]$  where  $x = \alpha + \beta\theta$

Paper integrates over all possible values of  $\theta$  to get

Profits:  $E(\pi) = p[\lambda - b_1]\sigma_u = p[cov(P_2 - P_1, \theta - \bar{v})\beta]$

Profits occur if  $z$  disclosed, otherwise  $P_1 = P_2$

With costly info acquisition, informed optimizes

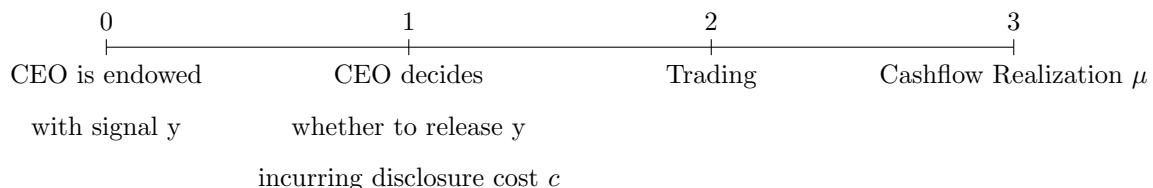
for his choice of  $\frac{1}{\sigma_1^{2*}}$ :  $\max_{\frac{1}{\sigma_1^2}} E(\pi) - C_1(\frac{1}{\sigma_1^2})$

## Verrecchia (1983)

### III. Voluntary Disclosure (Partial Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)

*Relative to Verrecchia (1983), traders here are risk neutral as opposed to risk averse.*



This paper shows that under reasonable conditions, a partial disclosure equilibrium will arise: good news disclosed and bad news withheld.

**Show  $\exists \hat{x}$  s.t.  $\hat{x}$  results in a Partial Equilibrium**

Let  $D =$  disclosure,  $ND =$  no disclosure.

Since we are looking for a partial equilibrium,

we want to show that  $\exists \hat{x}$  such that:

(1)  $\forall y \geq \hat{x}$ : CEO discloses  $y$

(2)  $\forall y < \hat{x}$ : no disclosure of  $y$

Traders are risk neutral  $\rightarrow$   
Risky cashflow  $\mu \sim N(y_0, h_0^{-1}) \rightarrow$

Private signal:  $y = \mu + \epsilon \rightarrow$

where  $\epsilon \sim N(0, s^{-1}) \rightarrow$

Disclosure cost:  $c \rightarrow$

Trader's priors  $\Omega$  are diffuse  $\rightarrow$

We are looking for a partial equilibrium  $\rightarrow$

For above cases, value of firm =

(1)  $E(\mu|D) = E(\mu - c | y)$

(2)  $E(\mu|ND) = E(\mu | y < \hat{x})$

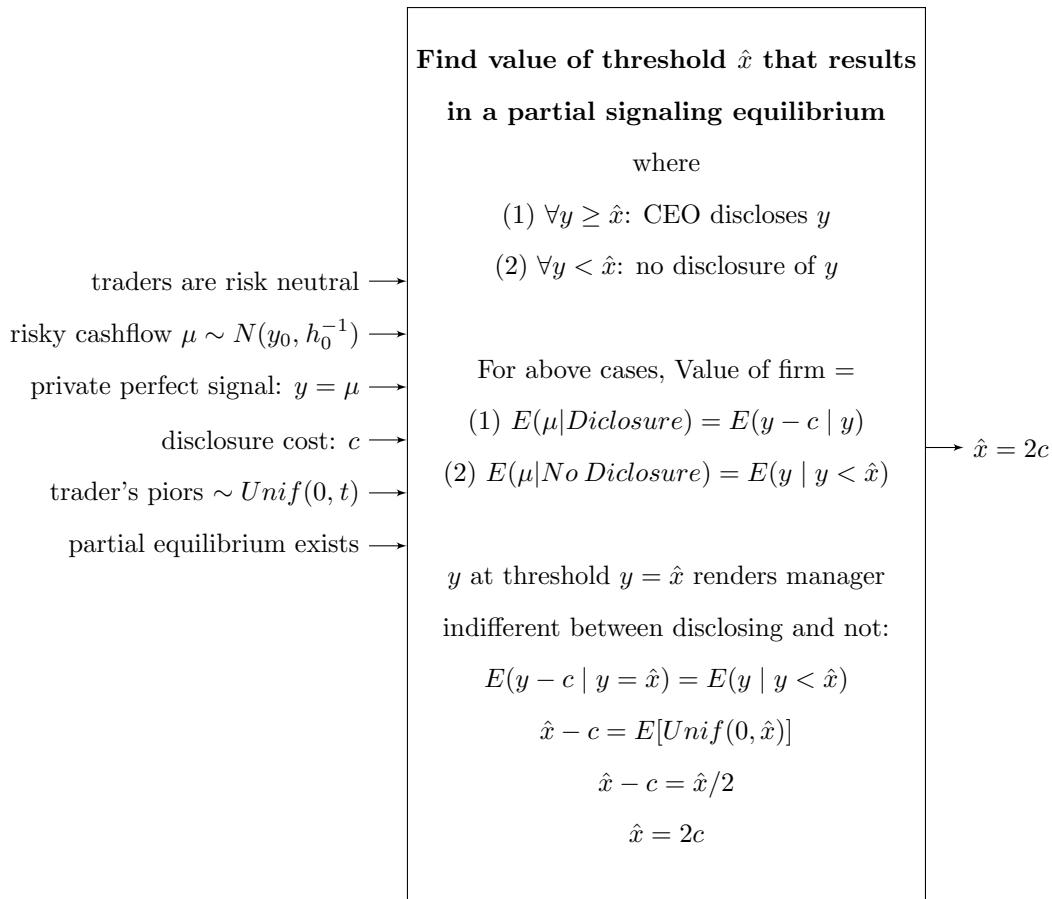
To solve, we consider that  $y$  at threshold  $y = \hat{x}$  will render CEO indifferent between disclosing and not:

$$E(\mu - c | y = \hat{x}) = E(\mu | y < \hat{x})$$

Author shows  $\exists$  unique  $\hat{x}$  solving above relation, proving that  $\exists$  a partial equilibrium in this setting

$\exists$  unique  $\hat{x}$  as a function of params resulting in partial signaling equilibrium

## Example p.5 (here we find specific term for threshold $\hat{x}$ )



# Beyer (2008)

## VI. Asymmetric Information (Fully Separating Signaling Equilibrium)

This paper studies analyst forecasting strategies and CEO's earnings management policy. When reporting earnings, CEO trades off disutility he gets from falling short of the analyst forecast against costs of manipulating earnings.

$t = 1$	$t = 2$	$t = 3$
The analyst observes a private signal $\omega_1$ , and issues his first forecast $AF_1$ .	With probability $\lambda$ the analyst observes a second private signal $\omega_2$ , and issues a revised forecast $AF_2$	The manager privately observes $x$ and issues the earnings report $R$ .

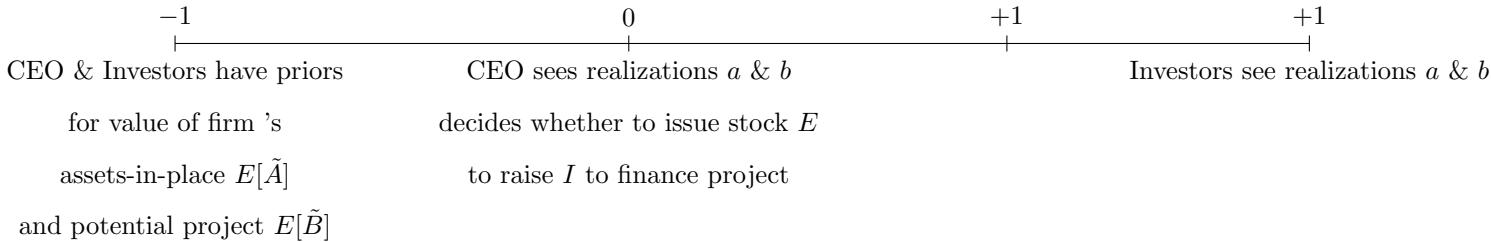
Firm has unmanaged earnings  $\tilde{x} \sim N(\mu_x, \tau_x^{-1})$  →  
 Realizations of  $\tilde{x}$  only observable to CEO →  
 Analyst's private signal  $\omega_1$  is a noisy measure of  $\tilde{x}$  →  
 $\omega_1 = x + \varepsilon_1 + \varepsilon_2; \varepsilon_1 \sim N(0, \tau_1^{-1}) \perp \varepsilon_2 \sim N(0, \tau_2^{-1})$  → which he bases himself on to release forecast  $AF_1$  →  
 He has a  $\lambda\%$  chance of getting another private signal →  
 $\omega_2 = x + \varepsilon_2$  leading him to release 2nd forecast  $AF_2$  →  
 Forecasts minimize error:  $\min_{CF} E[|\tilde{R} - CF||\{\omega\}]$  → where  $CF \in \{AF_1, AF_2\}$  is his *latest* forecast →  
 CEO sees forecasts and then observes realized  $x$  → Finally, CEO releases a reports  $R$  on  $x$  →  
 If  $R$  diverges from  $x$ , CEO incurs costs  $\frac{k_m}{2}(R - x)^2$  →  
 If  $R$  below analyst expectations, cost:  $\frac{k_f}{2}(R - CF)^2$  →  
 $U^{CEO}(R - AF) - \frac{k_m}{2}d^2$  and  $U' > 0, U'' < 0$  →  
 $d = R - x$  is CEO's choice of discretionary accruals →  
 Equilib. defined by policies:  $\{AF_1(), AF_2(), R(), P()\}$  →  
 $P()$  is market's pricing strategy:  $P(\Omega) = E[\tilde{x}|info]$  →  
 $R()$  is CEO's reporting rule  $R(x, AF_1, AF_2)$  which → minimizes  $\frac{k_f}{2}(R - CF)^2 + \frac{k_m}{2}(R - x)^2$  if  $R < CF$  → minimizes  $\frac{k_m}{2}(R - x)^2$  if  $R \geq CF$  →  
 Show  $\exists$  a fully revealing equilibrium in this setting →

**Proof by backwards induction**  $\exists$  fully revealing equilibrium  
 If the realization of unmanaged earnings exceed the analyst's most current forecast the CEO has no incentives to manipulate earnings  
 $Hence R(x, CF) = x$  if  $x \geq CF$   
 If  $x < CF$  then the manager solves  $\min_R \frac{k_f}{2}(R - CF)^2 + \frac{k_m}{2}d^2$  and hence  $R(x, CF) = x + \frac{k_f}{k_f + k_m}(CF - x)$  let  $g_1(\tilde{x}) = g(\tilde{x} | \omega_1)$  and  $g_2(\tilde{x}) = g(\tilde{x} | \omega_1, \omega_2)$  Then at  $t = 1, 2$  the analyst minimizes  
 $\int_{-\infty}^{\infty} |R(\tilde{x}, AF_t) - AF_t| g_t(\tilde{x}) dx$  which is equivalent to  
 $\frac{k_m}{k_f + k_m} \int_{-\infty}^{AF_t} (AF_t - \tilde{x}) g_t(\tilde{x}) dx + \int_{AF_t}^{\infty} (\tilde{x} - AF_t) g_t(\tilde{x}) dx$  The FOC to these optimization problem simplifies to  
 $G_t(AF_t) = \frac{k_f + k_m}{k_f + 2k_m}$  Note that the solution to this condition,  $AF_t(\omega_t)$ , is a function of  $\omega_t$  because  $G_t(\cdot)$  depends on  $\omega_t$ . The second order condition to the analyst's optimization problem is  $g_t'(AF_t) > 0$  and hence  $G_t'(AF_t)$  characterizes a unique solution. since the RHS of  $G_t(AF_t)$  exceeds  $\frac{1}{2}$ , it follows that  $AF_t > E[\tilde{x} | \omega_t]$ . Thus we can rewrite  $G_t(AF_t)$  as  
 $\frac{2}{\sqrt{\pi}} \int_0^{\frac{AF_t - E[\tilde{x} | \omega_t]}{\sqrt{2}\sigma_t}} e^{-y^2} dy = \frac{k_f + k_m}{k_f + 2k_m} - \frac{1}{2}$  which implies  $AF_t(\omega_t, \sigma_t) = E[\tilde{x} | \omega_t] + \sqrt{2}\sigma_t \operatorname{erf}^{-1}\left(\frac{k_f}{k_f + 2k_m}\right)$  since  $AF_1(\omega_1, \sigma_1)$  is strictly increasing in  $\omega_1$ , it is one-to-one and therefore observing  $AF_1$  is informationally equivalent to observing  $\omega_1$  (this follows because  $\sigma_1$  is a known parameter of the model). Therefore, the capital market price at  $t = 1$  is satisfies  
 $P(AF_1) = E[\tilde{x} | AF_1] = E[\tilde{x} | \omega_1]$ . similarly,  $P(AF_1, AF_2) = E[\tilde{x} | \omega_2]$   
 At  $t = 3$  the price can be derived by solving  $R(x, CF)$  for  $x$ . QED.

# Myers & Majluf (1982)

## VI. Asymmetric Information (Partial Signaling Equilibrium)

This paper describes investment and financing decisions when managers have private information about firm value which cannot be credibly convey to risk **neutral** investors. Acting in the interests of old shareholders', CEOs may pass up profitable projects as their positive NPV would be outweighed by the dilution. Thus the market interprets the decision of (not) issuing as negative (positive).



Firm has project requiring  $\$I$  outlay →

$I$  can be partly financed by known cash ( $S$ ) →

$S < I$ , so firm would need to issue  $E = I - S$  →

t=-1 expected value of assets:  $\bar{A} = E_{-1}[\tilde{A}]$  →

at t=0 CEO sees realized value of assets:  $a \geq 0$  →

t=-1 expected NPV of investment:  $\bar{B} = E_{-1}[\tilde{B}]$  →

at t=0 CEO sees realized value of project:  $b \geq 0$  →

CEOs act in interest of *old* (t=-1) shareholders →

Thus CEOs maximize  $V_0^{old} = V(a, b, E)$  →

P : market value at t = 0 if stock is not issued →

$P'$  : mkt value at t = 0 of *old* shares if stock issued →

If firm forfeits investment:  $V^{old} = S + a$  →

If it issues  $E$ :  $V^{old} = \frac{P'}{P'+E}(E + S + a + b)$  then →

*old* only gain if  $S + a < \frac{P'}{P'+E}(E + S + a + b) \Leftrightarrow$  →

$\frac{E}{P'+E}(S + a) < \frac{P'}{P'+E}(E + b) \Leftrightarrow \frac{E}{P'}(S + a) < E + b$  →

if  $\frac{E}{P'}(S + a) > E + b$ : firm does nothing → M →

if  $\frac{E}{P'}(S + a) < E + b$ : it issues & invests → M' →

Firm is indifferent at line where  $\frac{E}{P'}(S + a) = E + b$  →

Issue fairly priced if:  $P' = S + \bar{A}(M') + \bar{B}(M')$  →

where  $\bar{A}(M') \equiv E^{mkt}(\tilde{A} | E = I - S)$  →

and  $\bar{B}(M') \equiv E^{mkt}(\tilde{B} | E = I - S)$  →

Note that under no issue/project:  $P = S + \bar{A}(M)$  →

Likelihood of forfeiting project:  $F(M)$  →

Expected loss due to forfeit:  $L \equiv F(M)\bar{B}(M)$  →

$L$  increases with  $E = I - S$  →

### Special Case 1: Pooling equilibrium ( $a$ is known)

If  $a$  is also known to investors ( $\bar{A}(M') = a$ ) we have:

$P' = S + a + \bar{B}(M')$  and since  $\bar{B}(M') \geq 0$  and  $P' \geq S + a$

The firm will issue if  $E \left( \frac{S+a}{P'} \right) < E + b$

Which is satisfied given  $b \geq 0$ , because  $\frac{(S+a)}{P'} \leq 1$

Thus firm will issue whenever  $b > 0$ , and  $P' = S + a + \bar{B}$

### Special Case 2: Issues always dilute ( $P' \leq P$ )

Let  $a^*$  be the threshold of assets that renders the firm indifferent between  $M'$  and  $M$  (issuing and not). Using the indifference line, threshold  $a^*$  solves:  $a^* + S = P'(1 + b/E)$

Note that  $\bar{A}(M) + S > a^* + S$ , because any  $a < a^*$  would lead the firm

to issue ( $a < a^*$  implies  $a^* + S < P'(1 + b/E)$ ).

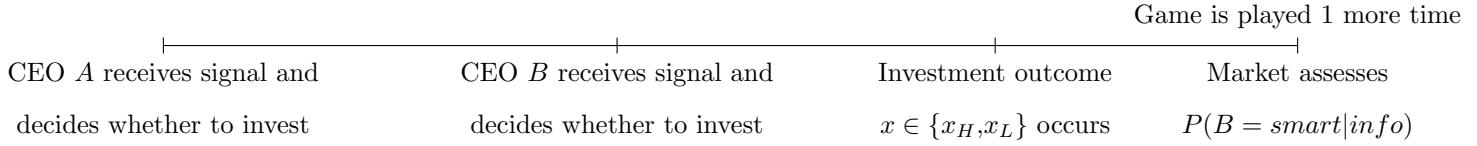
Since  $P = \bar{A}(M) + S$ ,  $P > P'(1 + b/E)$ .

Since  $b > 0$ ,  $P'(1 + b/E) \geq P'$  and  $P \geq P'$

# Scharfstein & Stein (1990)

## VI. Asymmetric Information (No Partial Signaling Equilibrium)

This paper shows the dynamics of herding in a situation where a manager is being judged by the accuracy of his private signal which he reports only after another manager has reported his.



<p>Possible investment outcome <math>x \in \{x_H, x_L\} \rightarrow</math>  <math>x_H &gt; 0</math> occurs with probability <math>P(x_H) = \alpha \rightarrow</math>  <math>x_L &lt; 0</math> occurs with probability <math>P(x_L) = 1 - \alpha \rightarrow</math>          CEOs have private signal <math>s \in \{s_g, s_b\} \rightarrow</math>  <math>\text{smart}</math> CEOs get informative signal: <math>\rightarrow</math>  <math>P(s_G   x_H, \text{smart}) = p &gt; P(s_G   x_L, \text{smart}) = q \rightarrow</math>  <math>\text{dumb}</math> CEOs get uninformative diffuse signal: <math>\rightarrow</math>  <math>P(s_G   x_H, \text{dumb}) = P(s_G   x_L, \text{dumb}) = z \rightarrow</math>          CEOs don't know their type, priors are public: <math>\rightarrow</math>          Priors: <math>P(\text{smart}) = \theta</math> and <math>P(\text{dumb}) = 1 - \theta \rightarrow</math>          There are 2 CEOs: <i>A</i> invests before <i>B</i> <math>\rightarrow</math>          Before making his investment decision <math>\rightarrow</math>  <i>B</i> can consider his signal &amp; <i>A</i>'s decision <math>\rightarrow</math>          Outcomes are public even if no one invests <math>\rightarrow</math>          Market updates its belief <math>\hat{\theta}</math> about CEO <i>B</i> <math>\rightarrow</math>              where <math>\hat{\theta}</math> is likelihood CEO <i>B</i> is smart <math>\rightarrow</math>  <math>\hat{\theta}(A's \text{ signal}, B's \text{ signal}, \text{ex-post } s \in \{s_g, s_b\}) \rightarrow</math>              based on investment decision outcome <math>\rightarrow</math>              and if <i>B</i> followed <i>A</i>'s decision (herding) <math>\rightarrow</math>          Both types equally as likely to see positive <math>s_g \rightarrow</math>  <math>P(x_H   s_G) = \mu_G ; P(x_H   s_B) = \mu_B \rightarrow</math>          Project is attractive if good signal is received <math>\rightarrow</math>  <math>\mu_G x_H + (1 - \mu_G) x_L &gt; 0 &gt; \mu_B x_H + (1 - \mu_B) x_L \rightarrow</math>          Thus if <i>A</i> goes first, <i>B</i> would know <i>A</i>'s signal <math>\rightarrow</math>          Signal draws are independent unless both <math>\rightarrow</math>          CEOs are <i>smart</i> then they're identical <math>\rightarrow</math>          CEOs are risk neutral &amp; maximize <math>\hat{\theta} \rightarrow</math>          Assume that <math>\alpha = 1/2</math> and show that there is no <math>\rightarrow</math>          fully revealing equilibrium for <i>B</i> in this setting <math>\rightarrow</math></p>	<p><b>Proof by contradiction:</b></p> <p>Since <i>smart</i> &amp; <i>dumb</i> are equally as likely to see positive signal <math>s_g</math>:  <math display="block">P(s_G   \text{smart}) = P(s_G \cap x_H   \text{smart}) + P(s_G \cap x_L   \text{smart})</math> <math display="block">\text{Knowing that } P(B A) = \frac{P(A \cap B)}{P(A)} \Leftrightarrow P(A \cap B) = P(A)P(B A)</math> <math display="block">P(s_G   \text{smart}) = P(x_H)P(s_G   x_H, \text{smart}) + P(x_L)P(s_G   x_L, \text{smart})</math> <math display="block">P(s_G   \text{smart}) = \alpha p + (1 - \alpha)q = P(s_G   \text{dumb}) = z = P(s_G)</math></p> <p><math display="block">\mu_G = P(x_H   s_G) = \frac{P(s_G   x_H)P(x_H)}{P(s_G)} = \frac{\theta p + (1 - \theta)z}{z} \alpha</math> <math display="block">\mu_B = P(x_H   s_B) = \frac{P(s_B   x_H)P(x_H)}{P(s_B)} = \frac{\theta(1-p) + (1-\theta)(1-z)}{1-z} \alpha</math></p> <p>Probability 2 <i>dumb</i> managers observe <math>s_G</math> is <math>z^2</math>  <math display="block">P(s_G, s_G   x_H \text{ or } x_L, \text{dumb}, \text{dumb}) = z^2</math></p> <p>Probability 2 <i>smart</i> managers observe <math>s_G</math> when state is high is <math>p</math>  <math display="block">P(s_G, s_G   x_H, \text{smart}, \text{smart}) = p \quad (\text{not } p^2)</math></p> <p>Note that <math>p = 1 - q &gt; q = 1 - p</math>. If <math>\alpha = 1/2</math> this implies that:  <math display="block">\alpha p + (1 - \alpha)q = z \Leftrightarrow p/2 + (1/2)(1 - p) = 1/2 = z</math></p> <p>What market surmises if following high state <math>x_H</math>,          it knew that <i>A</i> had a bad signal &amp; <i>B</i> had a good signal          (which is what <i>B</i> would let on if he breaks with <i>A</i>):          If <math>\hat{\theta}(A's \text{ signal}, B's \text{ signal}, s \in \{s_g, s_b\}) = \hat{\theta}(s_B, s_G, x_H)</math>  <math display="block">\hat{\theta}(s_B, s_G, x_H) = \frac{\frac{1}{2}(1-\theta)p\theta}{\frac{1}{2}(1-\theta)p\theta + \frac{1}{2}(1-\theta)(1-p)\theta + \frac{1}{4}(1-\theta)^2} = 2\theta p / (1 + \theta)</math></p> <p>If instead <i>A</i> saw <math>s_B</math> &amp; didn't invest, <i>B</i> would communicate the following if he followed <i>A</i>  <math display="block">\hat{\theta}(s_B, s_B, x_H) = \frac{2\theta(1-p)(1+\theta)}{4\theta(1-p)+(1-\theta)^2} ; \hat{\theta}(s_B, s_B, x_L) = \frac{2\theta p(1+\theta)}{4\theta p+(1-\theta)^2}</math></p> <p>Whereas if he broke with <i>A</i>:  <math display="block">\hat{\theta}(s_B, s_G, x_H) = 2\theta p / (1 + \theta) ; \hat{\theta}(s_B, s_G, x_L) = 2\theta(1 - p) / (1 + \theta)</math></p> <p style="text-align: right;"><i>continued on next page...</i></p>
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**Note that**  $P(s_G|smart) = P(x_H)P(s_G|x_H, smart) + P(x_L)P(s_G|x_H, smart) = \alpha p + (1 - \alpha)q$   
**and**  $P(s_L|smart) = P(x_L)P(s_L|x_L, smart) + P(x_H)P(s_L|x_H, smart) = (1 - \alpha)p + \alpha q = p - \alpha p + \alpha - \alpha p = \alpha + p - 2\alpha p = 1 - \alpha p - 1 - \alpha p + \alpha + p = 1 - \alpha p - (1 - \alpha)(1 - p) = 1 - (\alpha p + (1 - \alpha)(1 - p)) = 1 - P(s_G|smart)$   
**Just to confirm that**  $P(s_L|x_L, smart) = p$  **and**  $P(s_L|x_H, smart) = q = 1 - p$

### Proof of equation 12 :

I will use this relation:  $P(A) = P(A \& B) + P(A \& B^c)$  and  $P(B | A) = \frac{P(A \& B)}{P(A)} = \frac{P(A \& B)}{P(A \& B) + P(A \& B^c)}$   
Thus:  $P(B | A \& C) = \frac{P((A \& C) \& B)}{P(A \& C)} = \frac{P((A \& C) \& B)}{P((A \& C) \& B) + P((A \& C) \& B^c)}$   
So:  $P(x_H | s_B, s_G) = \frac{P(s_B, s_G, x_H)}{P(s_B, s_G, x_H) + P(s_B, s_G, x_H^c)} = \frac{P(s_B, s_G, x_H)}{P(s_B, s_G, x_H) + P(s_B, s_G, x_L)}$   
 $P(s_B, s_G, x_L) = \theta p(1 - \theta)(1 - z) + (1 - \theta)(z)\theta(1 - p) + (1 - \theta)z(1 - \theta)(1 - z)$   
 $P(s_B, s_G, x_L) = \frac{1}{2}\theta(1 - \theta)p + \frac{1}{2}\theta(1 - \theta)(1 - p) + \frac{1}{4}(1 - \theta)^2 = \frac{1}{2}\theta(1 - \theta) + \frac{1}{4}(1 - \theta)^2$  (given that  $z=1/2$ )  
 $P(s_B, s_G, x_H) = \theta(1 - p)(1 - \theta)\frac{1}{2} + (1 - \theta)\frac{1}{2}\theta p + \frac{1}{4}(1 - \theta)^2$   
 $P(s_B, s_G, x_H) = \theta(1 - p)(1 - \theta)\frac{1}{2} + (1 - \theta)\frac{1}{2}\theta p + \frac{1}{4}(1 - \theta)^2 = \frac{1}{2}\theta(1 - \theta) + \frac{1}{4}(1 - \theta)^2$   
Thus:  $P(x_H | s_B, s_G) = \frac{P(s_B, s_G, x_H)}{P(s_B, s_G, x_H) + P(s_B, s_G, x_L)} = \frac{1}{2}$  and  $P(x_L | s_B, s_G) = \frac{1}{2}$ .

### Proof of equation 13 : Similarly,

$$\begin{aligned} P(s_B, s_B, x_H) &= \theta^2(1 - p) + \theta(1 - p)(1 - \theta)(1 - \frac{1}{2}) + (1 - \theta)(1 - \frac{1}{2})\theta(1 - p) + \frac{1}{4}(1 - \theta)^2 \\ P(s_B, s_B, x_H) &= \theta^2(1 - p) + \theta(1 - \theta)(1 - p) + \frac{1}{4}(1 - \theta)^2 = \theta(1 - p) + \frac{1}{4}(1 - \theta)^2 \\ P(s_B, s_B, x_L) &= \theta^2 p + 2\theta p(1 - \theta)\frac{1}{2} + \frac{1}{4}(1 - \theta)^2 = \theta^2 p + \theta(1 - \theta)p + \frac{1}{4}(1 - \theta)^2 = \theta p + \frac{1}{4}(1 - \theta)^2 \\ P(x_H | s_B, s_B) &= \frac{P(s_B, s_B, x_H)}{P(s_B, s_B, x_H) + P(s_B, s_B, x_L)} = \frac{\theta(1-p)+\frac{1}{4}(1-\theta)^2}{\theta+\frac{1}{2}(1-\theta)^2} = \frac{4\theta(1-p)+(1-\theta)^2}{4\theta+2(1-\theta)^2} \text{ and} \\ P(x_L | s_B, s_B) &= \frac{P(s_B, s_B, x_L)}{P(s_B, s_B, x_H) + P(s_B, s_B, x_L)} = \frac{\theta p+\frac{1}{4}(1-\theta)^2}{\theta+\frac{1}{2}(1-\theta)^2} = \frac{4\theta p+(1-\theta)^2}{4\theta+2(1-\theta)^2} = 1 - P(x_H | s_B, s_B) \end{aligned}$$

**Inequality (14)** states the requirement that if  $B$  receives signal  $s_G$ , he prefers to invest (and identify himself as someone who had observed  $s_G$ ), rather than not invest (and masquerade as someone who had received signal  $s_B$ ). There for break with  $A$  in order to reveal his signal. Note from previous page:

$$\begin{aligned} \hat{\theta}(s_B, s_G, x_H) &= \hat{\theta}(s_G, s_B, x_L) = \frac{2\theta p}{1+\theta} & \hat{\theta}(s_B, s_G, x_L) &= \hat{\theta}(s_G, s_B, x_H) = \frac{2\theta(1-p)}{1+\theta} \\ \hat{\theta}(s_B, s_B, x_H) &= \hat{\theta}(s_G, s_G, x_L) = \frac{2\theta(1-p)(1+\theta)}{4\theta(1-p)+(1-\theta)^2} & \hat{\theta}(s_B, s_B, x_L) &= \hat{\theta}(s_G, s_G, x_H) = \frac{2\theta p(1+\theta)}{4\theta p+(1-\theta)^2} \end{aligned}$$

$$\begin{aligned} \hat{\theta}(s_B, s_G, x_H) P(x_H | s_B, s_G) + \hat{\theta}(s_B, s_G, x_L) P(x_L | s_B, s_G) &\geq \hat{\theta}(s_B, s_B, x_H) P(x_H | s_B, s_G) + \hat{\theta}(s_B, s_B, x_L) P(x_L | s_B, s_G) \\ \frac{1}{2}\hat{\theta}(s_B, s_G, x_H) + \frac{1}{2}\hat{\theta}(s_B, s_G, x_L) &\geq \frac{1}{2}\hat{\theta}(s_B, s_B, x_H) + \frac{1}{2}\hat{\theta}(s_B, s_B, x_L) \\ \frac{2\theta p}{1+\theta} + \frac{2\theta(1-p)}{1+\theta} &\geq \frac{2\theta(1-p)(1+\theta)}{4\theta(1-p)+(1-\theta)^2} + \frac{2\theta p(1+\theta)}{4\theta p+(1-\theta)^2} \Leftrightarrow \frac{2\theta}{1+\theta} \geq 2\theta(1+\theta) \left[ \frac{1-p}{4\theta(1-p)+(1-\theta)^2} + \frac{p}{4\theta p+(1-\theta)^2} \right] \\ 1 &\geq (1+\theta)^2 \left[ \frac{(1-p)[4\theta p+(1-\theta)^2] + (p)[4\theta(1-p)+(1-\theta)^2]}{[4\theta p+(1-\theta)^2][4\theta(1-p)+(1-\theta)^2]} \right] = (1+\theta)^2 \frac{4\theta p+(1-\theta)^2}{[4\theta p+(1-\theta)^2][4\theta(1-p)+(1-\theta)^2]} = (1+\theta)^2 \frac{1}{4\theta(1-p)+(1-\theta)^2} \\ 4\theta(1-p) + (1-\theta)^2 &\geq (1+\theta)^2 \Leftrightarrow 4\theta - 4\theta p + 1 - 2\theta + \theta^2 \geq 1 + 2\theta + \theta^2 \Leftrightarrow 4\theta - 4\theta p \geq 4\theta \Leftrightarrow -4\theta p \geq 0 \rightarrow \leftarrow \end{aligned}$$

The inequality in equation 14 is indeed violated. Thus there is no fully revealing equilibrium in which manager  $B$ 's investment decision depends on the signal he observes. Thus the only possible equilibria are those where manager  $B$  mimics manager  $A$  regardless of the signal, or where manager  $B$  does the opposite of manager  $A$  regardless of the signal. This is Proposition 1 in the paper.

# Trueman (1994)

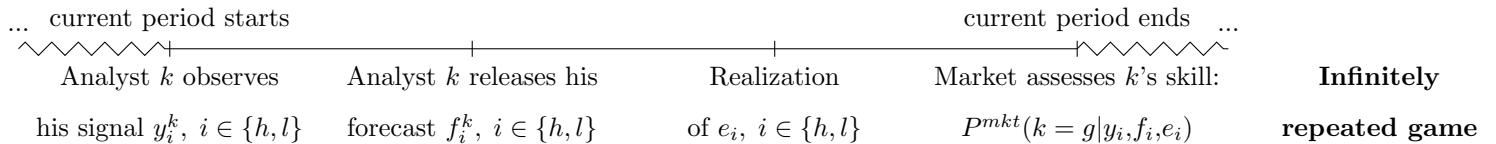
## IV. Asymmetric Information (Partially Revealing Signaling Equilibria)

Based on Winter 2019 Accounting Theory class notes on the paper (Prof. Trueman)

This paper looks at the incentives driving forecasting analysts to herd on their earnings forecasts and to examine the implications of this herding behavior on forecast accuracy and the return-earnings relation.

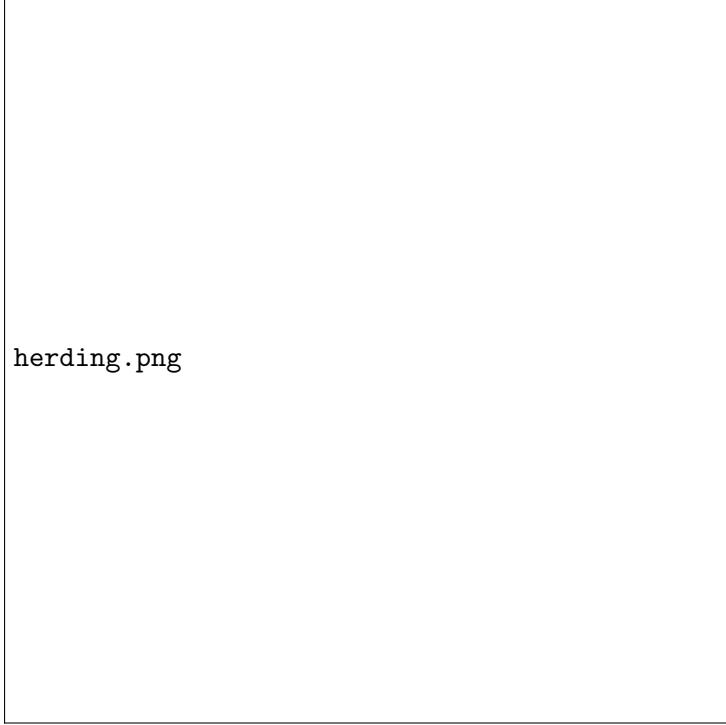
**Note:** While the notes introduce EPS sign  $\{-, +\}$ , only high/low level  $\{h, l\}$  of EPS matters in analysis.

### Equilibrium with simultaneous forecast release



<b>Show <math>\exists \alpha^*</math> which optimizes a <math>b</math> analyst's policy and yields a Partially Separating Equilibrium</b>	
Analyst knows his skill $k$ , investors don't $\rightarrow$	Market Assessment is $P^mkt(g   f_l, e_l) =$
$k \in \{b, g\}: k = g$ for good, $k = b$ for bad $\rightarrow$	$P(g   y_l, f_l, e_l) * P(y_l   f_l, e_l) + P(g   y_h, f_l, e_l) * P(y_h   f_l, e_l)$
$k$ forecasts $e \in \{e_l, e_h\}$ as $f^k(y_i^k) \in \{f_l, f_h\} \rightarrow$	where $P(g   y_h, f_l, e_l) = 0$ ( $\dagger$ ) and $P(g   y_l, f_l, e_l) = \frac{P(g, y_l, f_l, e_l)}{P(y_l, f_l, e_l)}$
$P(e_l) = 2t, P(e_h) = 1 - 2t, 2t > 1 - 2t \rightarrow$	$= \frac{P(y_l, f_l, g   e_l) P(e_l)}{P(y_l, f_l   e_l) P(e_l)} = \frac{P(y_l, f_l   g, e_l) P(g   e_l)}{g P(g) + b P(b)} = \frac{g P(g)}{g P(g) + b P(b)}$
$k$ observes his signal for level of $e$ : $y_i^k \rightarrow$	
$P^k(y_i^k   e_i) = k; P^k(y_j^k   e_i) = 1 - k \rightarrow$	
where $i \in \{h, l\}; k \in \{g, b\}$ with $g > b: \rightarrow$	Noting that: $P(y_l   f_l, e_l) = P(y_l, g   f_l, e_l) + P(y_l, b   f_l, e_l)$
e.g.: $P^b(y_i^b   e_i) = b; P^b(y_j^b   e_i) = 1 - b \rightarrow$	$P(y_l, g   f_l, e_l) = \frac{g P(g)}{g P(g) + (b + (1-b)\alpha) P(b)} = \frac{g P(g)}{g P(g) + b P(b) + \alpha(1-b) P(b)}$
$\dagger k$ truthfully reports his signal: $f^k(y_i^k) = f_i \rightarrow$	Similarly, $P(y_l, b   f_l, e_l) = \frac{b P(b)}{g P(g) + b P(b) + \alpha(1-b) P(b)}$
$\dagger$ except $b$ can lie at a rate $\alpha$ about his $y_h^b \rightarrow$	Thus, $P(y_l   f_l, e_l) = \frac{g P(g) + b P(b)}{g P(g) + b P(b) + \alpha(1-b) P(b)}$
such that $f^b(y_h^b) = (1 - \alpha)f_h + \alpha f_l \rightarrow$	And: $P^mkt(g   f_l, e_l) = \frac{g P(g)}{g P(g) + b P(b)} \frac{g P(g) + b P(b)}{g P(g) + b P(b) + \alpha(1-b) P(b)}$
Market assesses $k$ 's skill: $P_t^mkt(g   y_i, f_i, e_i, P_{t-1}^mkt) \rightarrow$	So: $P^mkt(g   f_l, e_l) = \frac{g P(g)}{g P(g) + b P(b) + \alpha(1-b) P(b)}$
$P^mkt(g   f_i, e_i, P_{t-1}^mkt)$ assessed over $\infty$ periods $\rightarrow$	Similarly: $P^mkt(g   f_l, e_h) = \frac{(1-g)P(g)}{(1-g)P(g) + (1-b)P(b) + \alpha b P(b)}$
As $T \rightarrow \infty$ : $P_t \rightarrow P_{t-1} \Rightarrow P^mkt(g   f_i, e_i) \rightarrow$	$P^mkt(g   f_h, e_l) = \frac{(1-g)P(g)}{(1-g)P(g) + (1-\alpha)(1-b)P(b)}$
$\Rightarrow b$ seeks policy to maximize $E^b[P^mkt(g   f_i, e_i)] \rightarrow$	$P^mkt(g   f_h, e_h) = \frac{g P(g)}{g P(g) + (1-\alpha)b P(b)}$
Market conjectures that $b$ uses $\alpha^*$ to solve for $\rightarrow$	
optimal policy $\max_\alpha E^b[P^mkt(g   f^b(y_h^b), e_i)   y_i^b] \rightarrow$	$E^b[P^mkt(g   f_h, e_i)   y_h^b] = \frac{(1-g)P(g)}{(1-g)P(g) + (1-\alpha)(1-b)P(b)} P^b(e_l   y_h^b)$
Show $\exists$ a separating equilibrium in this setting $\rightarrow$	$+ \frac{g P(g)}{g P(g) + (1-\alpha)b P(b)} P^b(e_h   y_h^b)$
	$E^b[P^mkt(g   f_l, e_i)   y_h^b] = \frac{g P(g)}{g P(g) + b P(b) + \alpha(1-b)P(b)} P^b(e_l   y_h^b)$
	$+ \frac{(1-g)P(g)}{(1-g)P(g) + (1-b)P(b) + \alpha b P(b)} P^b(e_h   y_h^b)$
	Given above, if market conjectures that upon seeing $y_h^b$ , analyst $b$ lies with probability $\alpha^*$ then at equilibrium,
	$b$ is indifferent between the two, thus equating:
	$E^b[P^mkt(g   f_h, e_i)   y_h^b] = E^b[P^mkt(g   f_l, e_i)   y_h^b]$
	gives an expression for $\alpha^*$ in a Partial Equilibrium. QED.

In order to lower the number of variables, I've simplified the possible  $e$  levels of  $e_h^-$ ,  $e_l^-$ ,  $e_l^+$ ,  $e_h^+$ , where:  $e_h^- < e_l^- < 0 < e_l^+ < e_h^+$  to simply  $e_h$  and  $e_l$  while maintaining the result given that  $e$ 's pdf is symmetric &  $E[e]=0$  (See Fig.).



$$P(e^l) = 2P(e_l^-) = 2P(e_l^+) = 2t, P(e_l) = 2t > 0.5 > 1 - 2t = P(e_h)$$

I also extended the use of the probabilities of accurate symbols ( $b/g$ ) to serve as indicators for weak/strong analysts ( $w/s$ , in the notes) again to lower the number of variables.

† This is the market's set of conjectures (S1)-(S6) in the paper. It stems from the fact that: Since low signals are more common in nature, for  $k \in \{g, b\}$ :  $E^k[P^{mkt}(g | E^k[e|y_l^k], f_l)] > E^k[P^{mkt}(g | E^k[e|y_l^k], f_h)]$   
So neither analyst has incentive to misreport low  $y_l$ 's

A good  $g$  analyst has no incentive to misreport  $y_h^g$ :  $E^g[P^{mkt}(g | E^g[e|y_h^g], f_h)] > E^b[P^{mkt}(g | E^b[e|y_l^b], f_h)]$   
Thus  $g$ 's optimal policy is never lying  $\alpha^{g*} = 0$ .

Some of the conditional probability relationships used here are listen below:

$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

$$P(B | A) = \frac{P(A \& B)}{P(A)} \Leftrightarrow P(B|A)P(A) = P(A \& B)$$

$$P(A | B) = \frac{P(B | A)P(A)}{P(B | A)P(A) + P(B | A^c)P(A^c)}$$

Mostly not used but listed for reference:

$$P(A \mid B \cap C) = \frac{P(B \mid A \& C)P(A \mid C)}{P(B \mid C)}$$

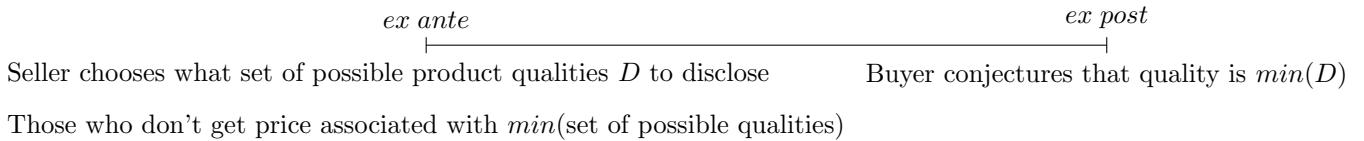
$$\frac{P(A \& B \& C)}{P(B \& C)} = \frac{P(A \& B \& C)P(A \mid C)}{P(A \& C)P(B \mid C)}$$

$$\frac{P(A \& B \& C)}{P(B \& C)} = \frac{P(A \& B \& C)P(A \& C)}{P(A \& C)P(B \mid C)P(C)}$$

Grossman (1981)

## 1. Background (Unraveling)

Under some circumstances, agents will voluntarily fully disclose credible private information. Though sometimes it may be in their best interest for a seller to for instance keep product quality undisclosed, we consider an equilibrium where the seller has incentives to disclose quality, even when it's low. Here, the way adverse selection works against sellers is by forcing full disclosure, instead of a market breakdown.



Let  $I$  be consumer's income  $\rightarrow$

$\gamma_i$  is belief about likelihood of quality  $q_i \rightarrow$

Expect. util.  $V(\gamma, p) = \sum_i \gamma_i U(q_i, I - p) \rightarrow$

Given willingness to pay price  $p(\gamma)$ :  $\rightarrow$

price  $p$  solves  $V(\gamma, p) = \bar{u} \rightarrow$

where  $\bar{u}$  is his baseline utility from not  $\rightarrow$

consuming the product →

Seller can make **credible & true** disclosure →

of privately known product quality  $q \in Q \rightarrow$

This is costlessly verifiable ex-post →

Let  $D(q)$  be disclosure if true quality is  $q \rightarrow$

Two ways consumer can infer quality  $q$ :  $\rightarrow$

1)  $q$  is fully revealed, and if not ... →

2) he conjectures value for unrevealed  $q \rightarrow$

$p_i$ : price buyer's willing to pay for quality  $q_i \rightarrow$

Let  $p_1 < p_2 < p_3 < \dots$  for  $q_1 < q_2 < q_3 < \dots \rightarrow$

If seller says quality  $q$  is in  $D = \{q_3, q_6, q_7\} \rightarrow$

yer will infer that  $q = \min(D) = q_3 \rightarrow$

Here, it is consumers' conjectures about the seller's quality out of equilibrium which forces the seller to choose a particular equilibrium (ie: off-equilibrium beliefs).

The seller is unable to mislead consumers because of the rational conjectures consumers have regarding what his quality must be if he deviates from full-information.

This occurs because buyers are unable to directly conjecture the quality through other means (eg: a costly signal) and because seller's disclosure is credible but costless.

Thus consumer hold the following beliefs  $\gamma_i^e$  about expected quality given disclosed possible quality set  $D$ :

$$Prob(q = q_i \in D) : \gamma_i^e(D) = \begin{cases} 0 & \text{if } q_i \text{ is not lowest quality in } D \\ 1 & \text{if } q_i \text{ is the lowest quality in } D \end{cases}$$

**Proof:** let  $\underline{q}_i$  be lowest quality in  $D$

A seller with quality  $k > q_i$  would get higher price if his disclosed set of possible qualities  $D^*$  excluded all qualities below  $k$ . Thus a profit maximizing seller would always disclose a set  $D^*$  which had as its lowest quality member actual quality  $k$  s.t.  $\min(D) = k$ . His profit from signaling  $D^*$  would be greater than for signaling  $D$ .

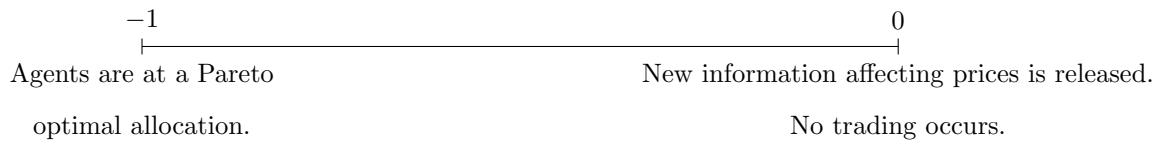
Thus this is a fully revealing equilibrium. “Off-equilibrium” beliefs would include that any seller who does not disclose has product with minimal possible product quality  $\min(Q)$ . QED

# Milgrom & Stokey (1982), Milgrom (1981a), Hirshleifer (1971)

## 1. Background (Common knowledge and trade)

Based on Spring 2020 Accounting Theory class notes on the paper (Prof. Caskey)

This paper can be described as the equivalent of M&M's take on how capital structure effects investment but applied to how info affects trade: it should not. If a population of risk-averse traders begins at a Pareto-optimal allocation then new private info alone will cause prices to change but generate no trade.



Pure exchange economy (no production) →  
 Individuals  $i$  trade consumption bundles →  
 state-space  $\Omega = \Theta \times X$  is partitioned into: →  
 -realization/payoff-relevant states  $\theta \in \Theta$  →  
 -purely informational (re:  $\theta$ ) states  $x \in X$  →

giving state of the world  $\omega = \{\theta, x\}$  →  
 $i$  has state-contingent endowment  $e_i(\theta)$  →  
 assume a single consumption good →  
 utility  $U_i(\theta, e_i(\theta))$ , prior beliefs  $p_i(\omega)$  →  
 $i$  has particular information partition  $\hat{P}_i$  →  
 $q(\theta)$ : ex ante price of consumption in state  $\theta$  →  
 $q(\theta)$  was a Pareto-optimal allocation →

In a competitive equilibrium prices satisfy: →

$$\frac{q(\theta)}{q(\theta')} = \underbrace{\frac{p_i(\theta)U'_i(\theta, e_i(\theta))}{p_i(\theta')U'_i(\theta', e_i(\theta'))}}_{MRS \text{ between } \theta \text{ & } \theta'}, \text{ in other words:} \rightarrow$$

relative prices of allocations  $\theta$  &  $\theta'$  equal relative marginal benefits of consuming those bundles →

Payoffs depend on the state  $\omega_k \in \Omega$  →  
 where  $\Omega = \{\omega_1, \dots, \omega_K\}$ ,  $K < \infty$  →

$i$ 's information set is defined by a partition  $P_i$  →  
 where if the true state is  $\omega$ ,  $i$  knows  $\omega$  is in  $P_i(\omega)$  →  
 $\wedge$  : ‘meet’ of partitions is finest common *coarsening* →  
 $\vee$  : ‘join’ of partitions is coarsest common *refinement* →

Beliefs are *concordant* if all agents agree on the likelihood of payoff-relevant state  $\theta$  yielding purely informational state  $x$ :  $p_1(x | \theta) = \dots = p_n(x | \theta) \forall x, \theta$  →

### Example:

$$P_1 : \begin{array}{c|cc|cc|cc|cc} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ \hline 1 & 2 & | & 3 & 4 & | & 5 & | \\ \end{array}$$

$$P_2 : \begin{array}{c|cc|cc|cc|cc} 1 & 2 & | & 3 & 4 & | & 5 & | & 6 & 7 & 8 \\ \hline 1 & 2 & | & 3 & 4 & | & 5 & | & 6 & 7 & 8 \\ \end{array}$$

$$\text{Meet } P_1 \wedge P_2 : \begin{array}{c|cc|cc|cc|cc} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ \hline 1 & 2 & | & 3 & 4 & | & 5 & | & 6 & 7 & 8 \\ \end{array}$$

$$\text{Join } P_1 \vee P_2 : \begin{array}{c|cc|cc|cc|cc} 1 & 2 & | & 3 & 4 & | & 5 & | & 6 & 7 & 8 \\ \hline 1 & 2 & | & 3 & 4 & | & 5 & | & 6 & 7 & 8 \\ \end{array}$$

Use  $R$  to denote the partitions defined by  $P_1 \wedge P_2$  where  $R(\omega)$  is the element of  $R$  that includes  $\omega$ .

In the above example,  $R(\omega_1) = R(\omega_2) = \dots = R(\omega_5) = \{1, 2, 3, 4, 5\}$  and  $R(\omega_6) = R(\omega_7) = R(\omega_8) = \{6, 7, 8\}$  and the trivial event  $\omega \in \Omega$ .

Milgrom (1981a) shows this is equivalent to an event that satisfies the following axioms where  $K_A = \{\omega \in \Omega : A \text{ is common knowledge at } \omega\}$

·  $K_A \subset A$  :  $A$  can be common knowledge only if it actually occurs

·  $\forall \omega \in K_A, \forall i, P_i(\omega) \subset K_A$  : If  $A$  is common knowledge, then everyone knows it is common knowledge.

·  $B \subset A \Rightarrow K_B \subset K_A$  : When  $B$  is common knowledge, any

consequence of  $B$  is as well. For example, if  $B$  is “the S&P 500 decreased by more than 5%” and  $A$  is “the S&P 500 decreased”, then if it is common knowledge that the S&P decreased by more than 5%, then it is common knowledge that the S&P decreased.

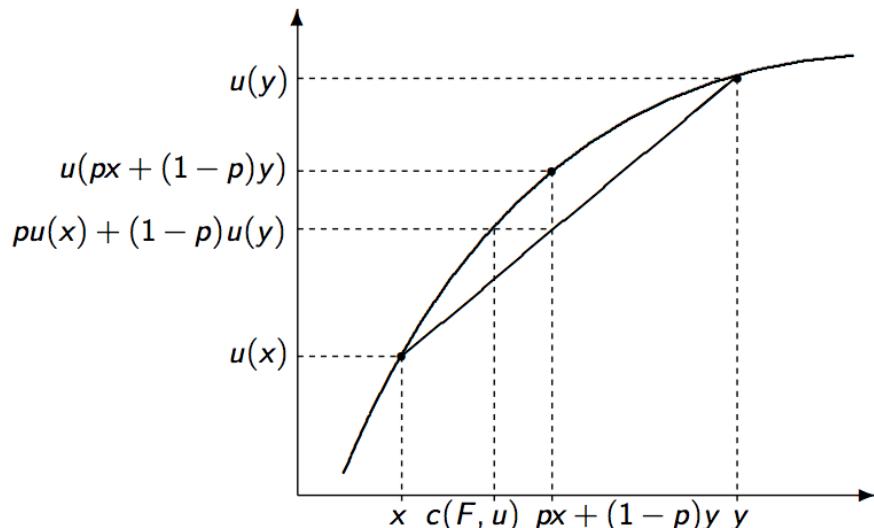
·  $\{\forall i, \forall \omega \in A, P_i(\omega) \subset A\} \Rightarrow A = K_A$  : All public events are common knowledge. When a  $\omega \in A$  occurs, all agents know (from  $P_i(\omega) \subset A$ ).

Concordant beliefs are a key condition in the case of private info. For public information, the condition can be weakened to *essentially concordant* beliefs, which restricts only relative likelihoods of payoff relevant states generating signals:  $\frac{p_1(x|\theta)}{p_1(x|\theta')} = \dots = \frac{p_n(x|\theta)}{p_n(x|\theta')}$ ,  $\forall x, \theta, \theta'$

**Hirshleifer (1971)** shows that information can be detrimental because it eliminates opportunities to insure. For example, assume that there are two types of individuals  $i \in \{1, 2\}$  and two states  $\omega_i \in \{\omega_1, \omega_2\}$  that occur with equal probability. There is one commodity and individuals have utility  $u(c)$ . Type  $i$  individuals have endowments  $c_i$  with payouts:

	$\omega_1$	$\omega_2$
$c_1$	\$2	\$1
$c_2$	\$1	\$2

If no information is released prior to trade, individuals will trade so that each holds  $\frac{1}{2}$  unit of  $c_1$  and  $\frac{1}{2}$  unit of  $c_2$ , giving expected utility of  $u(3/2)$ [1] If individuals could not trade, the expected utility of type  $i$  individuals is  $\frac{1}{2}u(2) + \frac{1}{2}u(1) < u(\frac{1}{2} \times 2 + \frac{1}{2} \times 1) = u(3/2)$  for any concave  $u$ . If information is released prior to trade, prices immediately shift and eliminate chances to insure.



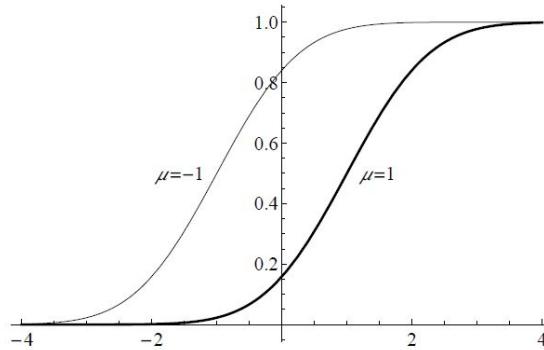
## Milgrom (1981b)

### 1. Background (Notions of ‘good news’: FOSD/SOSD/MLRP)

*Based on Accounting Theory class notes (Prof. Caskey Spring 2020):* A general and unambiguous notion of good news is first-order stochastic dominance (FOSD). If distribution  $F_1 \succ_{\text{FOSD}} F_2$ , then its CDF lies to the right of  $F_2$ . Thus, its mass is pushed to higher values. The following are equivalent definitions:

$$F_1 \succeq_{\text{FOSD}} F_2 \Leftrightarrow F_1(\theta) \leq F_2(\theta) \forall \theta \Leftrightarrow \underbrace{\int u(\theta) dF_1(\theta)}_{E_1[\theta]} \geq \underbrace{\int u(\theta) dF_2(\theta)}_{E_2[\theta]} \quad \forall \text{ increasing } u$$

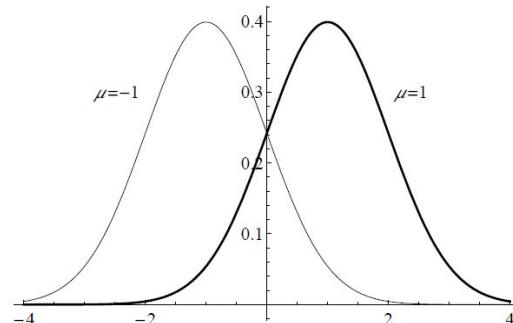
For example, if  $F_1$  and  $F_2$  are both normal distributions with the same variance, the one with the higher mean dominates. If we take an observation  $x$  to be a signal about the mean  $\theta \in \{-1, 1\}$ , a higher  $x$  is ‘good news’ in the sense that  $P(\theta = 1 | x)$  is increasing in  $x$



Monotone likelihood ratio property (MLRP) is a less general, but sometimes more useful, definition of good news. MLRP implies FOSD but not vice versa. A distribution with signal  $x$  has the MLRP if the ratio  $\frac{f(x|\theta_1)}{f(x|\theta_2)}$  is increasing in  $x$  for every  $\theta_1 > \theta_2$ . This is equivalent to:

$$f(x_1 | \theta_1) f(x_2 | \theta_2) - f(x_1 | \theta_2) f(x_2 | \theta_1) > 0, \forall x_1 > x_2, \theta_1 > \theta_2$$

MLRP implies FOSD. It also implies that the PDFs cross at a single point. For example, if  $F_1$  and  $F_2$  are both normal distributions with the same variance, the one with the higher mean MLRP dominates and the PDFs cross at a single point.



# Ohlson (1995)

## 2. Ohlsonomics (Asset prices and earnings)

*Based on material seen in Accounting Theory (Prof. Caskey Spring 2020)*

Ohlson shows how contemporaneous and future earnings, book values and dividends are price into the firm's market value.

$$\mathbf{z}_t = \begin{pmatrix} x_t^a \\ \mathbf{v}_t \end{pmatrix} = \boldsymbol{\Gamma} \begin{pmatrix} x_{t-1}^a \\ \mathbf{v}_{t-1} \end{pmatrix} + \boldsymbol{\epsilon}_t \text{ where } \boldsymbol{\Gamma} = \begin{pmatrix} \omega & 1 \\ 0 & \gamma \end{pmatrix}$$

This implies that  $E_t [\mathbf{z}_{t+s}] = \boldsymbol{\Gamma}^s \mathbf{z}_t$

To compute the infinite sum  $\sum_{s=1}^{\infty} R^{-s} E_t [\mathbf{z}_{t+s}]$ , write the eigenvalue decomposition  $\boldsymbol{\Gamma} = \mathbf{X}^{-1} \boldsymbol{\Lambda} \mathbf{X}$

so that  $\boldsymbol{\Gamma}^s = \mathbf{X}^{-1} \boldsymbol{\Lambda}^s \mathbf{X}$  and:

$$\sum_{s=1}^{\infty} R^{-s} E_t [\mathbf{z}_{t+s}] = (\sum_{s=1}^{\infty} R^{-s} \boldsymbol{\Lambda}^s) \mathbf{z}_t = \mathbf{X}^{-1} (\sum_{s=1}^{\infty} R^{-s} \boldsymbol{\Lambda}^s) \mathbf{X} \mathbf{z}_t$$

Assuming the sum converges:  $\sum_{s=1}^{\infty} R^{-s} \boldsymbol{\Lambda}^s = R^{-1} \boldsymbol{\Lambda} (\mathbf{I} - R^{-1} \boldsymbol{\Lambda})^{-1}$

$$\sum_{s=1}^{\infty} R^{-s} E_t [\mathbf{z}_{t+s}] = \mathbf{X}^{-1} R^{-1} \boldsymbol{\Lambda} (\mathbf{I} - R^{-1} \boldsymbol{\Lambda})^{-1} \mathbf{X} \mathbf{z}_t$$

$$\sum_{s=1}^{\infty} R^{-s} E_t [\mathbf{z}_{t+s}] = R^{-1} \boldsymbol{\Gamma} (\mathbf{I} - R^{-1} \boldsymbol{\Gamma})^{-1} \mathbf{z}_t$$

This gives:  $\Rightarrow P_t = y_t + \begin{pmatrix} 1 & \mathbf{0}' \end{pmatrix} R^{-1} \boldsymbol{\Gamma} (\mathbf{I} - R^{-1} \boldsymbol{\Gamma})^{-1} \mathbf{z}_t$

$$(1 \ \mathbf{0}') R^{-1} \boldsymbol{\Gamma} (\mathbf{I} - R^{-1} \boldsymbol{\Gamma})^{-1} = \begin{pmatrix} \frac{\omega}{R-\omega} & \frac{R}{(R-\omega)(R-\gamma)} \end{pmatrix}$$

$$\Rightarrow P_t = y_t + \frac{\omega}{R-\omega} x_t^a + \frac{R}{(R-\omega)(R-\gamma)} v_t$$

where  $\alpha_1 = \frac{\omega}{R-\omega} \geq 0$  is weight on abnormal earnings

and  $\alpha_2 = \frac{R}{(R-\omega)(R-\gamma)} \geq 0$  is weight on other info

The  $R - \omega$  in the denominators is like the  $r - g$  in a growing perpetuity where  $R = 1 + r$  and  $\omega = 1 + g$ . The  $\omega$  in the numerator for the abnormal earnings  $x_t^a$  coefficient appears because perpetuity begins after one period of growth.

The multiplier on 'other information'  $v_t$  has  $R$  in the numerator because of the lag for including in price and the denominator is a perpetuity-of-perpetuities: The 'other information' has growth rate  $\gamma - 1$

*continued on next page...*

By moving one period forward, we have an expression for stock return (proof below):

$$\frac{P_{t+1} + d_{t+1}}{P_t} = R_f + (1 + \alpha_1) \frac{\epsilon_{1,t+1}}{P_t} + \alpha_2 \frac{\epsilon_{2,t+1}}{P_t}$$

$P_t$  could also be expressed as a weighted average of earnings-multiplier value and book value:

$$P_t = k(\varphi x_t - d_t) + (1 - k)y_t + \alpha_2 v_t \quad \text{where } \varphi = \frac{R_f}{R_f - 1}, \quad k = \frac{(R_f - 1)\omega}{R_f - \omega}$$

If no other information could be used to predict earnings, i.e.,  $v_t = 0$ , assumed persistent abnormal earnings where  $\omega = 1$ , stock will be valued based on earnings multiplier. Moreover, assumed random walk of abnormal earnings where  $\omega = 0$ , stock will be valued based on book value.

**Proof of equation (6)** in the paper:  $\frac{P_{t+1} + d_{t+1}}{P_t} = R_f + (1 + \alpha_1) \frac{\epsilon_{1,t+1}}{P_t} + \alpha_2 \frac{\epsilon_{2,t+1}}{P_t}$

Derive the expression for  $P_{t+1} + d_{t+1} - R_f P_t$  and subsequently divide by  $P_t$

$$\begin{aligned} P_{t+1} + d_{t+1} - R_f P_t &= y_{t+1} + d_{t+1} + \alpha_1 x_{t+1}^a + \alpha_2 v_{t+1} - R_f (y_t + \alpha_1 x_t^a + \alpha_2 v_t) \\ &= (\alpha_1 + 1) x_{t+1}^a + \alpha_2 v_{t+1} - R_f \alpha_1 x_t^a - R_f \alpha_2 v_t \end{aligned}$$

Substituting  $x_{t+1}^a = \omega x_t^a + v_t + \varepsilon_{1t+1}$  and  $v_{t+1} = \gamma v_t + \varepsilon_{2t+1}$

$$P_{t+1} + d_{t+1} - R_f P_t = (\alpha_1 + 1) \varepsilon_{1t+1} + \alpha_2 \varepsilon_{2t+1} + \beta_1 x_t^a + \beta_2 v_t$$

where  $\beta_1 \equiv (\alpha_1 + 1) \omega - \alpha_1 R_f$  and  $\beta_2 \equiv (\alpha_1 + 1) + \alpha_2 \gamma - \alpha_2 R_f$

Since  $\alpha_1 = \omega / (R_f - \omega)$  and  $\alpha_2 = R_f / (R_f - \gamma) (R_f - \omega)$  we thus have  $\beta_1 = \beta_2 = 0$  then

$$(P_{t+1} + d_{t+1}) / P_t = R_f + (\alpha_1 + 1) \varepsilon_{1t+1} / P_t + \alpha_2 \varepsilon_{2t+1} / P_t$$

$$E[v_{t+\tau}] = \gamma^\tau v_t$$

$$E[x_{t+\tau}^a] = \omega^\tau x_t + \sum_{j=0}^{\tau-1} \omega^j \gamma^j v_{t+\tau-j}$$

# Caskey (2009)

## 3. Rational Expectations Model (with Ambiguity Aversion)

Based on material seen in Accounting Theory (Prof. Caskey Spring 2020)

This paper shows that persistent mispricing is consistent with a market that includes ambiguity-averse investors. In particular, ambiguity-averse investors may prefer to trade based on aggregate signals that reduce ambiguity at the cost of a loss in information.

- Paper uses Smooth ambiguity aversion:  $f \succeq g \Leftrightarrow \int_{p \in P} h(E_p[u(f)]) dQ(p) \geq \int_{p \in P} h(E_p[u(g)]) dQ(p)$
- Ambiguity-neutrality can be expressed by  $h(u) = u$ , giving:  $f \succeq g \Leftrightarrow \int_{\omega \in \Omega} u(f) dp^*(\omega) \geq E_{p^*}[u(g)]$   
where  $dp^*(\omega) = \int_{p \in P} dp(\omega) dQ(p)$
- Just as risk-aversion (concave utility) places more weight on adverse *outcomes*, ambiguity-aversion (concave  $h(u)$ ) places more weight on adverse *distributions*.
- Competitive noisy rational expectations model (a la Grossman & Stiglitz, 1980).
- The model has payoff  $v = v_1 + v_2$ , with corresponding signals  $s_i = v_i + e_i$
- All investors have risk-**tolerance**  $\rho$ , ie:  $u(c) = \exp\{-\gamma c\}$  and eg:  $E[u(c)] = -\exp\{-\frac{1}{\gamma}\mu_c + \frac{1}{2}\sigma_c^2 \frac{1}{\gamma^2}\}$
- Ambiguity-aversion is represented as the concave function  $h(u) = -(-u)^a$  where  $a = 1$  corresponds to ambiguity-neutrality and  $a > 1$  corresponds to ambiguity-aversion.
- Given a payoff  $v$  that is joint-normal with some ambiguous parameter  $b$ , and conditioning on signal  $s$ , the ambiguity-averse optimization problem can be written as:

$$\max_q \int h \left( \underbrace{\int u(q(v-p)) dF(v | b, s)}_{E[u(q(v-p)) | b, s]} \right) dF(b | s)$$

- Evaluating interior of expectation:  $E[u(q(v-p)) | b, s] = -\exp\left\{-\frac{1}{\rho} \left( q(E[v | b, s] - p) - \frac{1}{2\rho} q^2 var(v | b, s) \right)\right\}$
- With risk-ambiguity:  $h(E[u(q(v-p)) | b, s]) = -\exp\left\{-\frac{a}{\rho} \left( q(E[v | b, s] - p) - \frac{1}{2\rho} q^2 var(v | b, s) \right)\right\}$

$$E[h(E[u(q(v-p)) | b, s]) | s] = -\exp\left\{-\frac{a}{\rho} \left( q(E[v | s] - p) - \frac{1}{2\rho} q^2 (a var(E[v | b, s] | s) + var(v | b, s)) \right)\right\}$$

Using Law of Total Variance:  $var(v | s) = var(v | b, s) + var(E[v | b, s] | s)$ , we get the following certainty equivalent after substituting  $var(v | b, s) = var(v | s) - var(E[v | b, s] | s)$ :

$$q(E[v | s] - p) - \frac{1}{2\rho} q^2 (var(v | s) + (a-1) var(E[v | b, s] | s)) = q(E[v | s] - p) - \frac{1}{2\rho} q^2 var(v | s) (1 + (a-1) R(v, b | s))$$

where  $R(v, b | s) = \frac{var(E[v | b, s] | \theta)}{var(v | \theta)} = \frac{var(v | \theta) - var(v | b, s)}{var(v | s)}$  reflects the contribution of  $b$  to uncertainty about  $v$  given  $s$ ,

- If we were to aggregate signals  $s_1$  and  $s_2$ , we could compare  $R(v, b | s_1, s_2)$  to  $R(v, b | s^*)$  where  $s^* = \mathbb{E}[v | s_1, s_2]$ .
  - Conditional variance is unchanged from this aggregation of signals:
$$\text{var}(v | s_1, s_2) = \text{var}(v) - \text{var}(\mathbb{E}[v | s_1, s_2]) = \text{var}(v) - \text{var}(\mathbb{E}[v | s^*]) = \text{var}(v | s^*)$$
  - Thus:  $R(v, b | s_1, s_2) > R(v, b | s^*) \Leftrightarrow \text{var}(\mathbb{E}[v | b, s_1, s_2] | s_1, s_2) > \text{var}(\mathbb{E}[v | b, s^*] | s^*)$
  - Using the Kalman filter:  $\mathbb{E}[v | b, s_1, s_2] = \mathbb{E}[v | s_1, s_2] + \frac{\text{cov}(v, b | s_1, s_2)}{\text{var}(b | s_1, s_2)} (b - \mathbb{E}[b | s_1, s_2])$
  - so that:  $\text{var}(\mathbb{E}[v | b, s_1, s_2] | s_1, s_2) = \frac{\text{cov}(v, b | s_1, s_2)^2}{\text{var}(b | s_1, s_2)} = \underbrace{\left( \frac{\sigma_{v1}^2}{\sigma_{s1}^2} \right)^2}_{\text{Impact of } s_1 \text{ on } \mathbb{E}[v | s_1, s_2]} \frac{\sigma_{s1}^2}{\text{var}(s_1 | b)} \cdot \sigma_b^2$
- where  $\frac{\sigma_{s1}^2}{\text{var}(s_1 | b)}$  is the Variation of  $s_1$  caused by  $b$ .
- Similar computations give:  $\text{var}(\mathbb{E}[v | b, s^*] | s^*) = \underbrace{\left( \frac{\sigma_{v1}^2}{\sigma_{s1}^2} \right)^2}_{\text{Impact of } s_1 \text{ on } s^*} \frac{\text{var}(s^* | b)}{\text{var}(s^* | b)} \sigma_b^2$
- where again  $\frac{\text{var}(s^*)}{\text{var}(s^* | b)}$  is the Variation of  $s^*$  caused by  $b$ .
- So that  $R(v, b | s_1, s_2) > R(v, b | s^*) \Leftrightarrow \frac{\sigma_{s1}^2}{\text{var}(s_1 | b)} = \frac{1}{1 - \text{corr}(s_1, b)^2} > \frac{\text{var}(s^*)}{\text{var}(s^* | b)} = \frac{1}{1 - \text{corr}(s^*, b)^2}$   
 $\Leftrightarrow \text{corr}(s_1, b)^2 > \text{corr}(s^*, b)^2$
  - Because  $s^*$  is an aggregate signal,  $b$  causes a lower proportion of its variation, giving  $R(v, b | s_1, s_2) > R(v, b | s^*)$ . Formally:

$$\text{corr}(s_1, b)^2 = \frac{\sigma_b^2}{\sigma_{s1}^2} > \text{corr}(s^*, b)^2 = \frac{\sigma_{v1}^4 / \sigma_{s1}^2}{\text{var}(s^*)} \frac{\sigma_b^2}{\sigma_{s1}^2}$$

where the inequality follows because  $\text{var}(s^*) = \frac{\sigma_{v1}^4}{\sigma_{s1}^2} + \frac{\sigma_{v2}^4}{\sigma_{s2}^2}$

- The commingling of  $s_2$  with  $s_1$  makes ambiguity-averse investors more sure of how to interpret the aggregate signal.
- Because ambiguity-averse investors strictly prefer the aggregate information, they would be willing to pay for an aggregate signal. This holds even if the signal destroys information (Prop. 3 in paper). This can lead to mispricing.

# Guttman (2010)

## 3. Rational expectations (Competitive trader models)

Based on material seen in Accounting Theory (Prof. Caskey Spring 2020)

This study's model endogenizes the timing decision of analysts and analyzes their equilibrium timing strategies. Analysts face a trade-off between the timeliness and the precision of their forecasts.

Investors trade over  $t \in [0, T]$  →  
They have CARA coefficient  $\gamma$  →  
Asset payoff is  $\pi \sim \mathcal{N}(\mu_{\pi 0}, \sigma_{\pi 0}^2)$  →

Public information at time  $t$  has precision: →

$f(t)$  with  $f(0) = \frac{1}{\sigma_{\pi 0}^2} \geq 1$  (i.e.  $\sigma_{\pi 0}^2 \leq 1$ ) →  
 $f(t) \rightarrow \infty$  as  $t \rightarrow T$  and  $f'(t) > 0$  →

Investors can buy signal  $\psi_i(t) = \pi + e_i(t)$  →  
from analyst  $i$  where  $\text{var}(e_i(t)) = \frac{1}{f_i(t)}$  →

Imputing the cost of information  $c$  from →  
the informed/uninformed indifference†: →

† from Grossman & Stiglitz (1985) →  
 $c = \frac{1}{\gamma} \log \sqrt{\frac{\text{var}_u(\pi-p)}{\text{var}_i(\pi-p)}}$  →  
where  $\text{var}_u(\pi-p) = \frac{1}{f(t)}$  →

and  $\text{var}_i(\pi-p) = \frac{1}{f(t)+f_i(t)}$  and so: →

$U^i(f(t)) = C(t) = \frac{1}{\gamma} \log \sqrt{\frac{\text{var}_u(\pi-p)}{\text{var}_{(v-p)}}}$  →

$\frac{1}{2\gamma} \log \frac{f(t)+f_i(t)}{f(t)} \Leftarrow$  analyst  $i$  maximizes this →

By assumption  $f_i(t) = F_i + \alpha_i \log f(t)$  →

where  $\alpha_i$  is speed of analyst learning →

and  $F_i$  is his initial info advantage →

for some constants  $F_i \geq 0$  and  $\alpha_i > 0$  →

Both  $f(t)$  &  $f_i(t)$  increase over time →

$f'_i(t) = \frac{\alpha_i}{f(t)} f'(t)$ : analyst precision grows →

faster than public precision if  $f(t) < \alpha_i$  →

FOC to find optimal  $t^*$  at which analyst's payoff is maximized:

$$\frac{\partial U^i}{\partial f} = \frac{1}{2\gamma} \frac{\alpha_i - \overbrace{(F_i + \alpha_i \log f(t))}^{f_i(t)}}{f(t)(f(t)+f_i(t))} = 0 \Rightarrow \alpha_i - (F_i + \alpha_i \log f(t)) = 0 \\ \Rightarrow \frac{F_i}{\alpha_i} - 1 = -\log f(t) \Rightarrow f(t^*) = e^{1-F_i/\alpha_i} \Rightarrow t^* = f^{-1}(e^{1-F_i/\alpha_i})$$

If  $\alpha_i < f_i(0)$ , then  $\frac{\partial U^i}{\partial f} < 0 \forall f$  and the analyst immediately releases a forecast.

The condition  $\alpha_i < f_i(0)$  is equivalent to  $f(0) > e^{1-F_i/\alpha_i}$

Using the **implicit function theorem** for  $f(t) - e^{1-F_i/\alpha_i} = 0$

we get marginal effects of initial analyst precision  $F_i$  on optimal  $f(t^*)$ :

$$\frac{dt^*}{dF_i} = -\frac{\frac{\partial}{\partial F_i}(f(t)-e^{1-F_i/\alpha_i})}{\frac{\partial}{\partial t}(f(t)-e^{1-F_i/\alpha_i})} = -\frac{1}{\alpha_i} \frac{e^{1-F_i}}{f'(t^*)} = -\frac{1}{\alpha_i} \underbrace{\frac{f(t^*)}{f'(t^*)}}_{>0} < 0$$

and marginal effects of speed of learning  $\alpha_i$  on optimal  $f(t^*)$ :

$$\frac{dt^*}{d\alpha_i} = -\frac{\frac{\partial}{\partial \alpha_i}(f(t)-e^{1-F_i/\alpha_i})}{\frac{\partial}{\partial t}(f(t)-e^{1-F_i/\alpha_i})} = \frac{F_i}{\alpha_i^2} \underbrace{\frac{f(t^*)}{f'(t^*)}}_{>0} \Rightarrow \text{where } \text{sign}\left(\frac{dt^*}{d\alpha_i}\right) = \text{sign}(F_i)$$

$\frac{dt^*}{dF_i} < 0 \Rightarrow$  The more the analyst already knows, the earlier releases his forecast.

Extreme case leads him to release at  $t = 0$  to bank his initial info.

$\frac{dt^*}{d\alpha_i} > 0 \Rightarrow$  analysts who learn faster wait longer to issue forecasts.

**Two analysts  $i, j$ :** To find earliest time  $\tau_i$  at which analyst  $i$  will release a forecast we use the indifference interval defined by the  $t = \tau_i$  that solves:

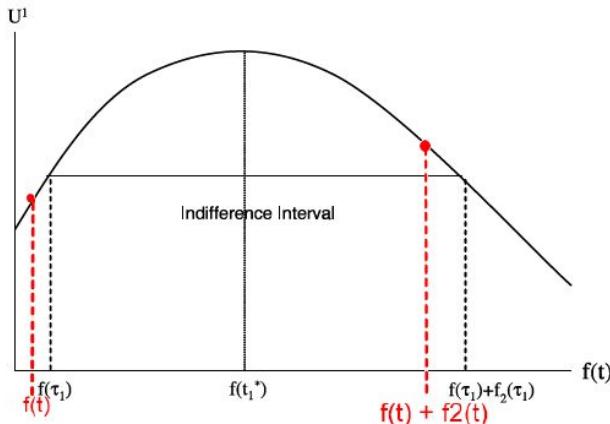
$$\underbrace{\log \frac{f(t) + f_i(t)}{f(t)}}_{i's \text{ payoff at } t \text{ from releasing first}} = \log \underbrace{\frac{f(t) + f_j(t)}{f(t) + f_j(t)}}_{\text{public info, now includes } j's \text{ forecast}} + \underbrace{\frac{F_i + \alpha_i \log(f(t) + f_j(t))}{f(t) + f_j(t)}}_{i's \text{ payoff at } t \text{ from releasing 2nd after } j}$$

Figure 5 shows that the analyst will never unilaterally issue before  $\tau_i$  since even if analyst  $j$  'jumps ahead' he can get a higher payoff by delaying. The main proposition says that the first analyst to forecast will release at the earlier of when the other analyst is indifferent ( $\tau_2$ ) or his own unconstrained optimum ( $t_1^*$ ). The other analyst will either cluster with the first or wait until his unconstrained optimum ( $t_2^*$ ). Fig. 6 & 7 illustrate the proposition.

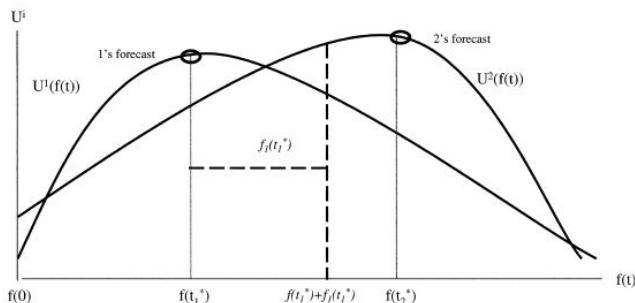
**Implicit Function Theorem:** If you have a function (say:  $x^2 + y^2 + z^2 = e^{xy}$ ) and you want to find a partial derivative of one of these variables w.r.t another (eg:  $\frac{\partial z}{\partial x}$ ) put it in the form  $F(x, y, z) = 0$  (so:  $F(x, y, z) = x^2 + y^2 + z^2 - e^{xy} = 0$ ) then (note the variable inversion):

$$\frac{\partial z}{\partial x} = -\frac{\frac{\partial F(x,y,z)}{\partial x}}{\frac{\partial F(x,y,z)}{\partial z}} \quad \text{eg: } \frac{\partial z}{\partial x} = -\frac{\frac{\partial F(x,y,z)}{\partial x}}{\frac{\partial F(x,y,z)}{\partial z}} = -\frac{2x - ye^{xy}}{2z}$$

**FIGURE 5**  
Interior Indifference Interval

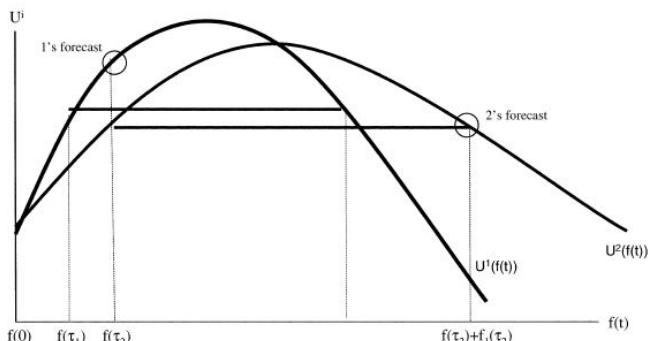


**FIGURE 6**  
Non-Clustering Pattern



Analyst 1 forecasts first at his unconstrained optimum:  $f(t_1^*)$ . Following his forecast, the precision of the investors' beliefs "jumps" to  $f(t_1^*) + f_1(t_1^*)$ , which is still lower than the unconstrained optimum of analyst 2. Analyst 2 waits until the precision of the investors' beliefs equals his unconstrained optimum and then publishes his forecast.

**FIGURE 7**  
Clustering Pattern

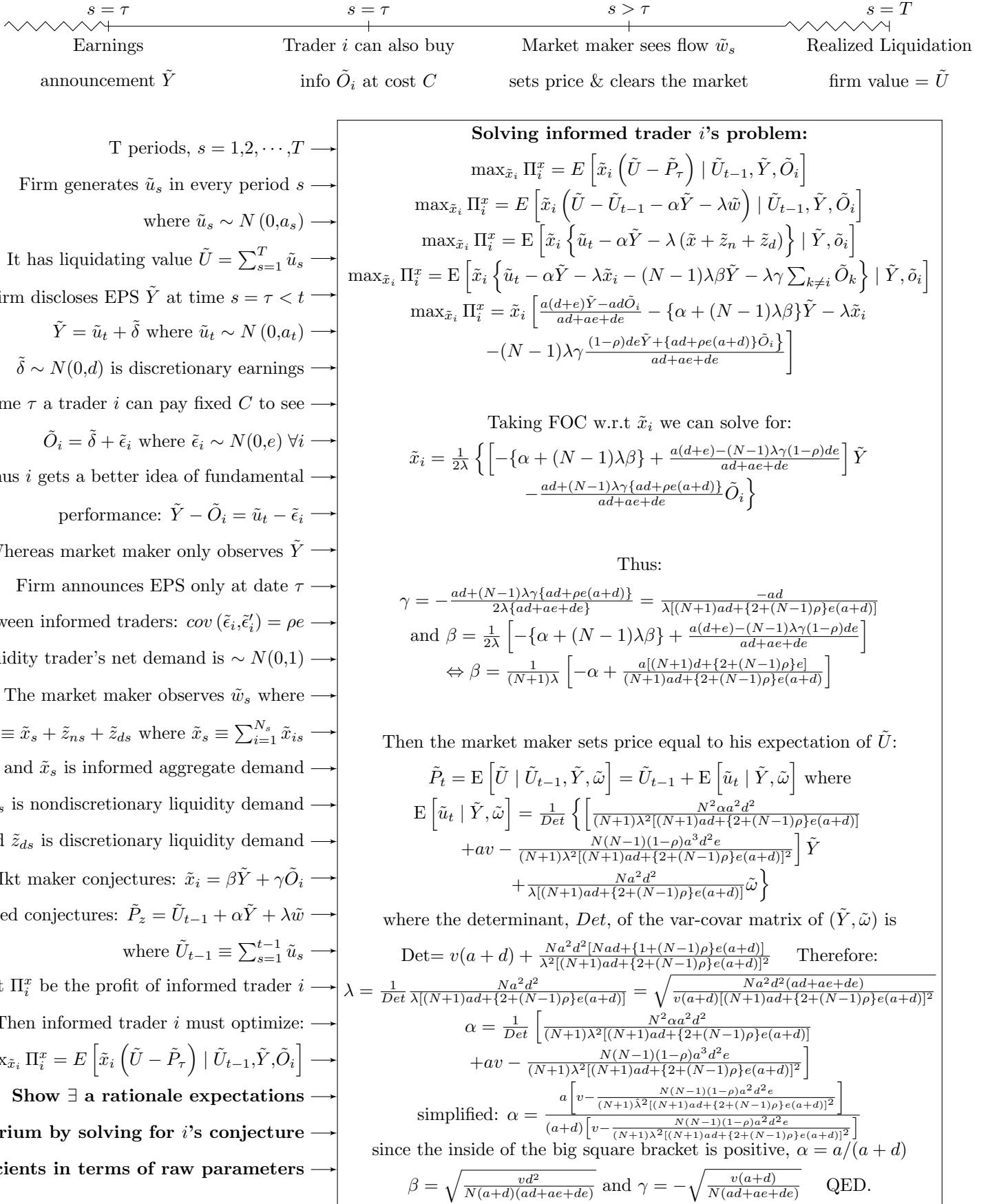


Since  $\tau_1 < \tau_2$ , analyst 1 is the first to forecast. He issues his forecast at  $\tau_1$ . Following his forecast, the precision of the investors' beliefs "jumps" to  $f(\tau_2) + f_1(\tau_2)$ , which is higher than the unconstrained optimum of analyst 2. Since the payoff of analyst 2 at this region is decreasing in the precision of the investors' beliefs, analyst 2 issues his forecast immediately after the forecast of analyst 1.

# Kim & Verrecchia (1994)

## 4. Rational Expectations ( Strategic Trader Models )

This paper suggests that earnings announcements provide information that allows certain traders to make judgments about a firm's performance that are superior to the judgments of other traders.



# Kyle (1989)

## 4. Rational Expectations (Strategic Trader Models)

This paper models investors which consider the price impact of their trades when optimizing their optimal demands.



Liquidity trader demand  $z \sim N(0, \sigma_z^2) \rightarrow$   
 $n$ 's obj. function:  $\max_{q_n} E_n[u(w_n + q_n(v - p))] \rightarrow$   
 places **limit orders** & has risk-aversion:  $r_n \rightarrow$   
 $n$  conjectures his price impact:  $p = p_0 + \lambda_n q_n \rightarrow$   
 $\max_{q_n} E_n[u(w_n + q_n(v - (p_0 + \lambda_n q_n)))]$ , FOC:  $\rightarrow$   
 $0 = E_n[u'(w_n + q_n(v - p))(v - p - \lambda_n q_n)] = \rightarrow$   
 $E_n[u'(w_n + q_n(v - p_0 - \lambda_n q_n))(v - p_0 - 2\lambda_n q_n)] \rightarrow$   
 with certainty equivalence and  $u(c) = -e^{-r_n c} \rightarrow$   
 $0 = r_n e^{-r_n(w_n + q_n' E_n[v - p] - \frac{p_n}{2} q_n \text{var}_n(v - p) q_n)} \rightarrow$   
 $\times (E_n[v - p] - \lambda_n q_n - r_n \text{var}_n(v - p) q_n) \rightarrow$   
 $\Leftrightarrow 0 = E_n[v - p] - \lambda_n q_n - r_n \text{var}_n(v - p) q_n \rightarrow$   
 $\Rightarrow q_n = \frac{E_n[v - p]}{r_n \text{var}_n(v - p) + \lambda_n}$  so  $\lambda_n$  deflates demand  $\rightarrow$   
 $N_I$  informed &  $N_U$  uninformed investors  $\rightarrow$   
 Informed risk aversion:  $r_I$ , uninformed:  $r_U \rightarrow$   
 Use mkt clearing cond.:  $\sum_{n=1}^N q_n(p) + x = 0 \rightarrow$   
 Assume  $p$  is known:  $q_n = \frac{E_n[v] - p}{r_n \text{var}_n(v) + \lambda_n} \rightarrow$   
 Informed  $n$  trader's signal:  $s_n = v + e_n \rightarrow$   
 where  $v \sim N(\mu_v, \tau_v^{-1}) \perp \text{iid } \epsilon \sim N(0, \tau_e^{-1}) \rightarrow$   
 Uninformed conjectures informed  $q_I(p, s_n) \rightarrow$   
 $q_I(p, s_n) = \mu_I + \beta s_n - \gamma_I p \rightarrow$   
 Informed conjectures uninformed  $q_U(p) \rightarrow$   
 $q_U(p) = \mu_U - \gamma_U p \rightarrow$   
 Overall market price conjecture form:  $\rightarrow$   
 $p = p_0 + p_{\bar{s}} \bar{s} + p_x x \rightarrow$

Thus  $E[v|s_n] = \mu_v + \frac{\tau_v^{-1}}{\tau_v^{-1} + \tau_e^{-1}}(s_n - \mu_v)$   
 $= \mu_v + \frac{1}{1 + \frac{\tau_v}{\tau_e}}(s_n - \mu_v) = \mu_v + \frac{\tau_e}{\tau_e + \tau_v}(s_n - \mu_v)$   
 $n$  also has price  $p$  as signal, so 1st consider:  
 $E[v | p] = E[v] + \frac{\text{cov}(v, p)}{\text{var}(p)}(p - E[p])$   
 Since  $s_n$  affects both  $p$  and  $v$  we condition above on it:  
 $E[v|p, s_n] = E[v|s_n] + \frac{\text{cov}(v, p|s_n)}{\text{var}(p|s_n)}(p - E[p | s_n])$   
 where  $p - E[p | s_n] = p - (E[p] + \frac{\text{cov}(p, s_n)}{\text{var}(s_n)}(s_n - \mu_v))$   
 Informed demand:  $q_n = \frac{E[v|s_n, p] - p}{\rho_I \text{var}(v|s_n, p) + \lambda_I}$   
 $= \frac{\frac{\tau_v}{\tau_v + \tau_e} \mu_v - \frac{\text{cov}(v, p|s_n)}{\text{var}(p|s_n)} \left( E[p] - \frac{\text{cov}(p, s_n)}{\text{var}(s_n)} \mu_v \right)}{\rho_I \tau_I^{-1} + \lambda_I}$   
 $+ \underbrace{\frac{\frac{\tau_e}{\tau_v + \tau_e} + \frac{\text{cov}(v, p|s_n)}{\text{var}(p|s_n)} \frac{\text{cov}(p, s_n)}{\text{var}(s_n)}}{\rho_I \tau_I^{-1} + \lambda_I} s_n}_{\beta} - \underbrace{\frac{1 - \frac{\text{cov}(v, p|s_n)}{\text{var}(p|s_n)}}{\rho_I \tau_I^{-1} + \lambda_I} p}_{\gamma_I}$   
 where  $\tau_I = \text{var}(v | s_n, p)^{-1}$   
 Meanwhile uninformed have:  $q_U = \frac{E[v|p] - p}{\rho_U \text{var}(v|p) + \lambda_U}$   
 $= \underbrace{\frac{\mu_U - \frac{\text{cov}(v, p)}{\text{var}(p)} E[p]}{\rho_U \tau_U^{-1} + \lambda_U}}_{\mu_U} - \underbrace{\frac{1 - \frac{\text{cov}(v, p)}{\text{var}(p)}}{\rho_U \tau_U^{-1} + \lambda_U} p}_{\gamma_U}$   
 where  $\tau_U = \text{var}(v | p)^{-1}$   
 Now the market clearing condition where all flows = 0:  
 $0 = \sum_{n=1}^{N_I} (\mu_I + \beta s_n - \gamma_I p) + N_U (\mu_U - \gamma_U p) + x$   
 $= N_I (\mu_I - \gamma_I p) + \beta N_I \bar{s}_I + N_U (\mu_U - \gamma_U p) + x$   
 $\Rightarrow p = \frac{N_I \mu_I + N_U \mu_U}{N_I \gamma_I + N_U \gamma_U} + \frac{\beta N_I}{N_I \gamma_I + N_U \gamma_U} \bar{s}_I + \frac{1}{N_I \gamma_I + N_U \gamma_U} x$   
 where  $\bar{s}_I$  is the average signal:  $\bar{s}_I = \frac{1}{N_I} \sum_{n=1}^{N_I} s_n$   
 Thus price satisfies the following  
 $p = \underbrace{\frac{N_I \mu_I + N_U \mu_U}{N_I \gamma_I + N_U \gamma_U}}_{p_0} + \underbrace{\frac{\beta N_I}{N_I \gamma_I + N_U \gamma_U}}_{p_{\bar{s}}} \bar{s}_I + \underbrace{\frac{1}{N_I \gamma_I + N_U \gamma_U}}_{p_x} x$

- Informed and
- uninformed
- demands:
- $X_I$  and  $X_U$ .
- Price
- function  $p$ 's
- conjecture
- coefficients:
- $p_x, p_{\bar{s}}, p_0$
- as functions
- of parameters.

In Kyle (1985), the price setting mechanism is that market-makers set price equal to the expected value given order flow  $y = x + u$  (equivalent to  $\hat{q}_i = q_i + x$  in here). This is equivalent to the market clearing condition with a single informed trader and an unbounded number of uninformed traders. One way to see this is to assume that uninformed trader  $n$  conjectures that other uninformed traders have demand of the form  $\mu_U - \gamma_U p$  and that the informed trade  $q_i$  is joint-normal with  $v$ . The market clearing condition  $(\sum_{n=1}^N q_n(p) + x = 0)$  then gives (from the point of view of uninformed trader  $n$ ):

$$q_n + (N_U - 1)(\mu_U - \gamma_U p) + q_I + x = 0$$

giving the following conjectures for price and informed demand  $\hat{q}_i$ :

$$p = \frac{(N_u - 1)\mu_u + q + x}{(N_u - 1)\gamma_u} + \underbrace{\frac{1}{(N_u - 1)\gamma_u} q_n}_{\lambda_u}$$

$$\hat{q}_i = (N_u - 1)(\gamma_u p - \mu_u) - q_n = q_i + x$$

$$E[v | p, q_n] = E[v | \hat{q}_i] = \mu_v + \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} (\hat{q}_i - E[q_i] - \mu_x)$$

Uninformed trade quantity is then the following where the second line follows after taking into account the  $q_u$  embedded in  $\hat{q}_i$  and rearranging:

$$q_u = \frac{E[v | \hat{q}_i] - p}{\rho_u var(v | \hat{q}_i) + \lambda_u} = \underbrace{\frac{\mu_v - \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} ((N_u - 1)\mu_u + E[q_i + x])}{\rho_u var(v | \hat{q}_i) + \lambda_u + \frac{cov(v, q_i)}{var(q_i) + \sigma_w^2}}}_{\mu_u} - \underbrace{\frac{1 - \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} (N_u - 1)\gamma_u}{\rho_u var(v | \hat{q}_i) + \lambda_u + \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2}} p}_{\gamma_u}$$

This satisfies the linear conjecture and we can get the expression for price by solving the following expression using the  $\mu_u$  and  $\gamma_u$  expressions above:  $N_u(\mu_u - \gamma_u p) + q_i + x = 0 \Rightarrow p = \frac{\mu_u}{\gamma_u} + \frac{1}{N_u \gamma_u} (q_i + x)$

The expression for the  $\gamma_u$  coefficient implies:  $\gamma_u = \frac{1}{\rho_u var(v | \hat{q}_i) + \lambda_u + \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} N_u}$

and the expression for the  $\mu_u$  coefficient implies:  $\mu_u = \gamma_u \left( \mu_v - \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} E[q_i + x] \right)$

Using the above conjecture for  $\hat{q}_i$ , the relation  $\lambda_u = \frac{1}{(N_u - 1)\gamma_u}$  implies

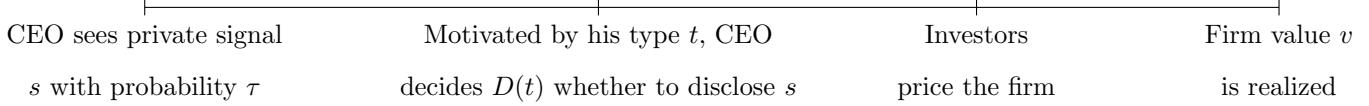
$$\lambda_u = \frac{\rho_u}{N_u - 2} var(v | \hat{q}_i) + \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} \frac{N_u}{N_u - 2} \Rightarrow \gamma_u = \frac{N_u - 2}{N_u - 1} \frac{1}{\rho_u var(v | q_i) + \frac{cov(v, q_i)}{var(q_i) + \sigma_w^2} N_u}$$

$$\Rightarrow p = \mu_v + \underbrace{\frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} (q_i + x - E[q_i + x])}_{E[v | q_i + x]} + \frac{1}{N_u - 2} \frac{cov(v, q_i)}{var(q_i) + \sigma_x^2} (q_i + x) + \frac{N_u - 1}{N_u - 2} \frac{\rho_u \mu_u}{N_u} var(v | \hat{q}_i)$$

which approaches the  $E[v | q_i + x]$  from Kyle (1985) as  $N_u \rightarrow \infty$

# Einhorn & Ziv (2012)

## 5. ‘Semi’ Verifiable Disclosure Models (Signal Jamming)



### Case 1: truthful disclosure ( $c = 0$ ):

Solve by finding threshold  $\hat{v}_{b1}$  that renders CEO indifferent between N & ND:

$$E[v | D] - \lambda = E[v | ND] \text{ where}$$

$$E[v | ND] = P(\text{Uninformed} | ND)E[v] + P(\text{Informed} | ND)E[v | v < \hat{v}_{b1}]$$

$$\text{Thus: } \hat{v}_{b1} - \lambda = \underbrace{\frac{1 - \tau}{1 - \tau + \tau F(\hat{v}_{b1})}}_{\mathbf{P}(\text{ Uninformed} | ND)} \mu + \underbrace{\frac{\tau F(\hat{v}_{b1})}{1 - \tau + \tau F(\hat{v}_{b1})}}_{\mathbf{P}(\text{ Informed} | ND)} \underbrace{\frac{1}{F(\hat{v}_{b1})} \int_{-\infty}^{\hat{v}_{b1}} x dF(x)}_{E[v | v < \hat{v}_{b1}]}$$

Investors are risk **neutral** → Uncertain firm value  $v \sim N(\mu, \sigma^2)$  →

CEO sees signal  $s$  with probability  $\tau$  →

where  $s = v + \varepsilon$  and  $\varepsilon \sim N(0, \sigma_\varepsilon^2)$  →

CEO can make report  $r$  if informed →

or CEO can choose to not disclose (ND) →

This report incurs cost  $\lambda + c(r - s)^2$  →

Let  $\hat{v} = E[v | s]$  and  $\delta = \frac{\sigma^2}{\sigma^2 + \sigma_\varepsilon^2}$  →

where  $var(\hat{v}) = cov(v, \hat{v}) = \frac{\sigma^4}{\sigma^2 + \sigma_\varepsilon^2} = \delta\sigma^2$  →

$\delta \in (0, 1)$  represents extent of CEO info →

$\Phi$  denotes standard normal distribution →

$\phi$  denotes the standard normal density →

with standard normal being:  $\sim N(0, 1)$  →

$$\hat{v}_{b1} - \lambda = \underbrace{\frac{1 - \tau}{1 - \tau + \tau \Phi\left(\frac{\hat{v}_{01} - \mu}{\sigma \sqrt{d}}\right)}}_{\mathbf{P}(\text{ Uninformed} | ND)} \mu + \underbrace{\frac{\tau \Phi\left(\frac{\hat{v}_{b1} - \mu}{\sigma \sqrt{d}}\right)}{1 - \tau + \tau \Phi\left(\frac{\hat{v}_{b1} - \mu}{\sigma \sqrt{d}}\right)}}_{\mathbf{P}(\text{ Informed} | ND)} \underbrace{\left(\mu - \sigma \sqrt{d} \frac{\phi\left(\frac{\hat{v}_{\Delta 1} - \mu}{\sigma \sqrt{d}}\right)}{\Phi\left(\frac{\hat{v}_{b1} - \mu}{\sigma \sqrt{d}}\right)}\right)}_{E[v | v < \hat{v}_{b1}]}$$

$$\hat{v}_{b1} - \lambda = \mu - \frac{\tau \phi\left(\frac{\hat{v}_{01} - \mu}{\sigma \sqrt{d}}\right)}{1 - \tau + \tau \Phi\left(\frac{\hat{v}_{01} - \mu}{\sigma \sqrt{d}}\right)} \sigma \sqrt{d}$$

$$\Rightarrow \frac{\lambda}{\sigma \sqrt{d}} = x + \frac{\tau \phi(x)}{1 - \tau + \tau \Phi(x)} \text{ where } x = \frac{\hat{v}_{b1} - \mu}{\sigma \sqrt{d}}$$

### Case 2: Mandatory disclosure with only lying costs

write the investors’ conjecture as  $r = r_0 + r_v v$  to get CEO’s payoff:

$$\underbrace{\frac{r - r_0}{r_v}}_{E[v|r]} - \frac{c}{2}(r - \hat{v})^2 \Rightarrow 0 = \frac{1}{r_v} - c(r - v) \Rightarrow r = v + \frac{1}{r_v c} \Rightarrow r_v = 1, r_0 = \frac{1}{c}$$

### Case 3: Voluntary disclosure with manipulation

CEO’s objective function:  $\max_r E[v | r] - \lambda - \frac{c}{2}(r - \hat{v})^2$

$$\text{FOC: } 0 = \frac{\partial E[v|r]}{\partial r} - c(r - \hat{v}) \Rightarrow r = \hat{v} + \underbrace{\frac{1}{c} \frac{\partial E[v|r]}{\partial r}}_{b(\hat{v})}$$

To prevent a CEO with  $\hat{v}$  from mimicking one with  $\hat{v} + \Delta\hat{v}$ , it must be that:

$$\begin{aligned} & E[v | r = \hat{v} + b(\hat{v})] - \lambda - \frac{c}{2}b(\hat{v})^2 \\ & \geq E[v | r = \hat{v} + \Delta\hat{v} + b(\hat{v} + \Delta\hat{v})] - \lambda - \frac{c}{2}(\Delta\hat{v} + b(\hat{v} + \Delta\hat{v}))^2 \\ & \text{Then } E[v | r = \hat{v} + b(\hat{v})] = \hat{v} \Rightarrow -b(\hat{v})^2 \geq \frac{2}{c}\Delta\hat{v} - (\Delta\hat{v} + b(\hat{v} + \Delta\hat{v}))^2 \\ & -b(\hat{v})^2 \geq \frac{2}{c}\Delta\hat{v} - (b(\hat{v})^2 + 2b(\hat{v})(\Delta\hat{v} + b(\hat{v})) + (\Delta\hat{v} + b(\hat{v}))^2) \\ & \Leftrightarrow 0 \leq \left(1 + \frac{\Delta b}{\Delta v}\right)^2 + 2\frac{b(\hat{v})}{\Delta v} \left(1 + \frac{\Delta b}{\Delta v}\right) - \frac{2}{c} \frac{1}{\Delta v} \\ & \Leftrightarrow 1 + \frac{\Delta b}{\Delta v} \notin \left(-\frac{1}{\Delta v} \left(\sqrt{\frac{2}{c}\Delta\hat{v} + b(\hat{v})^2} + b(\hat{v})\right), \frac{1}{\Delta v} \left(\sqrt{\frac{2}{c}\Delta\hat{v} + b(\hat{v})^2} - b(\hat{v})\right)\right) \\ & \text{where } \Delta b = b(\hat{v} + \Delta\hat{v}) - b(\hat{v}). \text{ Taking the positive root gives:} \\ & \frac{db}{dv} \geq \lim_{\Delta v \rightarrow 0} \underbrace{\frac{1}{\Delta v} \left(\sqrt{\frac{2}{c}\Delta\hat{v} + b(\hat{v})^2} - b(\hat{v})\right)}_{\Delta b / \Delta v} - 1 \\ & = \lim_{\Delta v \rightarrow 0} \frac{1}{c} \left(\frac{2}{c}\Delta\hat{v} + b(\hat{v})^2\right)^{-1/2} - 1 = \frac{1}{cb(\hat{v})} - 1 \quad (\text{using l'Hopital's rule}) \end{aligned}$$

Similar to the above, to prevent a manager with  $\hat{v} + \Delta\hat{v}$  from mimicking one with  $\hat{v}$ , it must be the case that:

$$\begin{aligned} \mathbb{E}[v \mid r = \hat{v} + \Delta\hat{v} + b(\hat{v} + \Delta\hat{v})] - \lambda - \frac{c}{2}b(\hat{v} + \Delta\hat{v})^2 &\geq \mathbb{E}[v \mid r = \hat{v}] - \lambda - \frac{c}{2}(b(\hat{v}) - \Delta\hat{v})^2 \\ \Leftrightarrow 0 &\geq \left( \frac{\Delta b}{\Delta v} \right)^2 + 2 \frac{b(v)}{\Delta v} \frac{\Delta b}{\Delta v} - \left( \frac{2}{c} \frac{1}{\Delta v} - 2 \frac{b(v)}{\Delta v} + 1 \right) \\ \Leftrightarrow \frac{\Delta b}{\Delta v} &\in \left( -\frac{1}{\Delta v} \left( \sqrt{(b(\hat{v}) - \Delta\hat{v})^2 + \frac{2}{c}\Delta\hat{v}} + b(\hat{v}) \right), \frac{1}{\Delta v} \left( \sqrt{(b(\hat{v}) - \Delta\hat{v})^2 + \frac{2}{c}\Delta\hat{v}} - b(\hat{v}) \right) \right) \end{aligned}$$

Again using the positive root and taking limits gives the following again using l'Hopital's rule:

$$\frac{db}{dv} \leq \lim_{\Delta v \rightarrow 0} \frac{1}{\Delta v} \left( \sqrt{(b(\hat{v}) - \Delta\hat{v})^2 + \frac{2}{c}\Delta\hat{v}} - b(\hat{v}) \right) = \frac{1}{cb(\hat{v})} - 1$$

The two inequalities implies that  $\frac{db}{d\hat{v}} = \frac{1}{cb(\hat{v})} - 1$  so that  $b(\hat{v})$  solves the differential equation  $b(\hat{v})(1 + b'(\hat{v})) = \frac{1}{c}$ . The solution is  $b(\hat{v}) = \frac{1}{c} \left( 1 + \omega \left( -e^{-(1+c\hat{v}-c^2k)} \right) \right)$  where  $\omega$  is the Lambert-W function and the boundary condition  $b(\hat{v}_0) = 0$  implies  $k = \frac{\hat{v}_0}{c}$ . This gives:

$$b(\hat{v}) = \frac{1}{c} \left( 1 + \omega \left( -e^{-(1+c(\hat{v}-\hat{v}_0))} \right) \right)$$

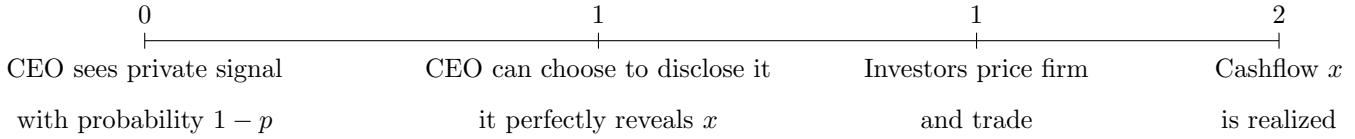
Note that we have  $b(\hat{v}_0) = \frac{1}{c} \left( 1 + \omega(-e^{-1}) \right) = 0$  and  $\lim_{\hat{v} \rightarrow \infty} b(\hat{v}) = \frac{1}{c}(1 + \omega(0)) = \frac{1}{c}$

# Jung & Kwon (1988)

## 5. Verifiable Disclosure Models (Partial Signaling Equilibria)

Based on Spring 2020 Accounting Theory class notes on the paper (Prof. Caskey)

Here partial disclosure can arise even when there are no disclosure costs.



If CEO discloses  $x$ , t=1 price will equal  $x$   
otherwise price considering that:

$$Prob(\text{CEO uninformed}) = p$$

$$Prob(\text{CEO withholding}) = (1 - p)F(\hat{x})$$

$$Prob(\text{CEO discloses}) = (1 - p)[1 - F(\hat{x})]$$

$$\text{where } F(\hat{x}) = Prob(x \leq \hat{x}) = \int_{\underline{x}}^{\bar{x}} dF(x)$$

Now given non-disclosure ( $ND$ ):

$$\begin{aligned} E[x|ND] &= \frac{p}{p+(1-p)P(x<\hat{x})} E[x] + \frac{(1-p)P(x<\hat{x})}{p+(1-p)P(x<\hat{x})} E[x|x < \hat{x}] \\ E[x|ND] &= \frac{p}{p+(1-p)F(\hat{x})} \mu + \frac{(1-p)F(\hat{x})}{p+(1-p)F(\hat{x})} \int_{\underline{x}}^{\hat{x}} \frac{xdF(x)}{F(\hat{x})} \\ &= \frac{p}{p+(1-p)F(\hat{x})} \mu + \frac{(1-p)}{p+(1-p)F(\hat{x})} \left( \hat{x}F(\hat{x}) - \int_{\underline{x}}^{\hat{x}} F(x)dx \right) \end{aligned}$$

Firm has risky cashflow  $x$  with prior mean  $\mu$  →  
 $x$  has support  $[\underline{x}, \bar{x}]$  and distribution  $F(x)$  →

Investors are risk **neutral** →

at t=0, CEO sees private signal with probability  $p$  →

If received, it perfectly reveals  $x$  →

CEO can choose to credibly disclose it →

Disclosure is costless →

CEO cannot credibly convey he's uninformed →

CEO's goal is to maximize firm valuation →

Conjecture that CEO follows a threshold strategy →

Show  $\exists$  a partially revealing equilibrium →

$$\begin{aligned} E[x|D] &= E[x|ND] \Leftrightarrow E[x|x = \hat{x}] = E[x|x < \hat{x}] \\ \hat{x} &= \frac{p}{p+(1-p)F(\hat{x})} \mu + \frac{(1-p)}{p+(1-p)F(\hat{x})} \left( \hat{x}F(\hat{x}) - \int_{\underline{x}}^{\hat{x}} F(x)dx \right) \\ \hat{x}(p + (1-p)F(\hat{x})) &= p\mu + (1-p) \left( \hat{x}F(\hat{x}) - \int_{\underline{x}}^{\hat{x}} F(x)dx \right) \\ 0 &= p(\mu - \hat{x}) - (1-p) \int_{\underline{x}}^{\hat{x}} F(x)dx \Rightarrow \text{denote this by } h(\hat{x}) \end{aligned}$$

Higher (FOSD) and/or less-risky (SOSD) payoffs lead to higher threshold values of  $\hat{x}$ , ie: less disclosure.

**Proof by contradiction:**

Let  $F \succeq_{FOSD} G$  and  $F \succeq_{SOSD} G$

Then assuming above isn't true, if thresholds  $\hat{x}_g \geq \hat{x}_f$ :

$$\begin{aligned} 0 &= h(\hat{x}_f, F) - h(\hat{x}_g, G) \\ 0 &= p \underbrace{(\hat{x}_g - \hat{x}_f)}_{>0} + p \underbrace{(\mu_f - \mu_g)}_{\geq 0 \text{ if } F \succeq_{FOSD} G} \\ &+ (1-p) \underbrace{\int_{\underline{x}}^{\hat{x}_f} (G(x) - F(x))dx}_{\geq 0 \text{ if } F \succeq_{SOSD} G} + (1-p) \underbrace{\int_{\underline{x}}^{\hat{x}_g} G(x)dx}_{>0} \end{aligned}$$

→ ←

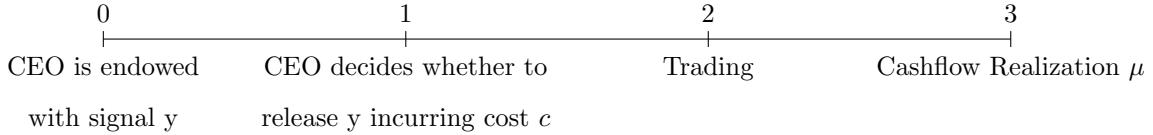
$\exists$  a unique  
 $\hat{x}^*$  s.t.  $\hat{x}^*$   
→ ⇒ partially  
revealing  
equilibrium

# Verrecchia (1983)

## 5. Verifiable Disclosure Models (Partial Signaling Equilibria)

Based on Spring 2020 Accounting Theory class notes on the paper (Prof. Caskey)

This paper shows that under reasonable conditions, a partial disclosure equilibrium will arise: good news disclosed and bad news withheld.



Show  $\exists \hat{x}$  s.t.  $\hat{x}$  results in a Partial Equilibrium

we want to show that  $\exists \hat{x}$  such that:

- (1)  $\forall y \geq \hat{x}$ : CEO discloses  $y$
- (2)  $\forall y < \hat{x}$ : no disclosure of  $y$

For above cases, value of firm =

- (1)  $E(\mu|D) = E(\mu - c | y)$
- (2)  $E(\mu|ND) = E(\mu | y < \hat{x})$

To solve, we consider that  $y$  at threshold  $y = \hat{x}$  will render CEO indifferent between disclosing and not:

$$E(\mu - c | y = \hat{x}) = E(\mu | y < \hat{x})$$

or working with posteriors  $\hat{\mu} = E[\mu|y]$  &  $\hat{\mu}_x = E[\mu|y = x]$ :

$$E(\hat{\mu} - c | \hat{\mu} = \hat{\mu}_x) = E(\hat{\mu} | \hat{\mu} < \hat{\mu}_x)$$

$\exists$  unique  $\hat{x}$  as a function of params resulting in partial signaling equilibrium

Traders are risk neutral  $\rightarrow$

Risky cashflow  $\mu \sim N(y_0, h_0^{-1}) \rightarrow$

Private signal:  $y = \mu + \epsilon \rightarrow$

where  $\epsilon \sim N(0, s^{-1}) \rightarrow$

Disclosure cost:  $c \rightarrow$

Trader's priors  $\Omega$  are diffuse  $\rightarrow$

Let  $D$ = disclosure  $\rightarrow$

$ND$ = no disclosure  $\rightarrow$

Looking for a partial equilibrium  $\rightarrow$

For normal variables  $\mu \sim N(y_0, h_0^{-1})$  &  $\epsilon \sim N(0, s^{-1})$

This gives  $E[u|y] = \hat{u} = y_0 + \frac{1}{\frac{1}{h_0} + \frac{1}{s}} (y - y_0) = \frac{1}{h_0+s} (h_0 y_0 + s y)$

$E[\hat{u}] = \frac{1}{h_0+s} (h_0 y_0 + s E[y]) = y_0$  and  $var(\hat{u}) = \frac{s}{h_0+s} \frac{1}{h_0} = cov(u, \hat{u})$

Let  $\phi(\cdot)$  &  $\Phi(\cdot)$  denote pdf and cdf of  $\hat{u} \sim \mathcal{N}\left(y_0, \frac{s}{h_0+s} \frac{1}{h_0}\right)$

$$E[u | y < \hat{x}] = E[\hat{u} | \hat{u} < \hat{u}_x] = y_0 - \sigma_u \frac{\phi\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)}{\Phi\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)}$$

The disclosure threshold will satisfy:

$$E[\hat{u} | \hat{u} < \hat{u}_x] = \hat{u}_x - c \Leftrightarrow \frac{c}{\sigma_u} = \frac{\hat{u}_x - y_0}{\sigma_u} + \frac{\phi\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)}{\Phi\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)} = f\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)$$

Let's use the implicit function theorem to get  $\frac{d\hat{u}_x}{d\sigma_u}$

$$\text{Let } 0 = F(c, \sigma_u, y_0, \hat{u}_x) = f\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right) - \frac{c}{\sigma_u}$$

$$\frac{d\hat{u}_x}{d\sigma_u} = -\frac{\frac{\partial F}{\partial \sigma_u}}{\frac{\partial F}{\partial \hat{u}_x}} = -\frac{-f'\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right) \frac{\hat{u}_x - y_0}{\sigma_u^2} + \frac{c}{\sigma_u^2}}{f'\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right) \frac{1}{\sigma_u}} = \frac{\hat{u}_x - y_0}{\sigma_u} - \frac{\frac{c}{\sigma_u}}{f'\left(\frac{\hat{u}_x - y_0}{\sigma_u}\right)}$$

# Ottaviani & Sorensen (2006)

## 6. Cheap Talk Models (Partition Equilibria)

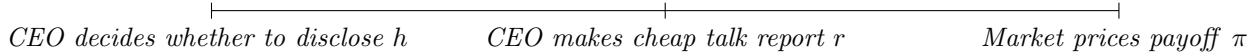
This paper shows that if analysts have incentives to appear talented, then cheap talk distorts their behavior and limits communication. The paper considers impact of various objective functions on analyst forecasts where all parties have common priors.

<i>Analyst of type <math>t</math> releases his forecast <math>m</math></i>	<i>Outcome <math>x</math> is realized</i>	<i>Market evaluates analyst's talent <math>t</math></i>
<p><math>n</math> analysts simultaneously issue forecasts →</p> <p>Common prior of outcome <math>x \sim N(\mu, v^{-1})</math> →</p> <p>Each forecaster <math>i</math> sees <math>s_i = x + \epsilon_i</math> →</p> <p>where <math>s_i x \sim N(x, \tau_i^{-1})</math> and their posterior becomes: <math>x s_i \sim N\left(\frac{\tau_i s_i + v \mu}{\tau_i + v}, \frac{1}{\tau_i + v}\right)</math> →</p> <p>Denote pdf of this posterior by <math>q_i(x s_i)</math> →</p> <p>Denote an honest forecast from <math>i</math> by <math>h_i</math>: →</p> <p><math>h_i(s_i) = E(x s_i) = \left(\frac{\tau_i}{\tau_i + v}\right) s_i + \left(\frac{v}{\tau_i + v}\right) \mu</math> →</p> <p>Let <math>t_i &gt; 0</math> be forecast accuracy talent of <math>i</math> →</p> <p>Common prior: <math>t</math> is distributed by pdf <math>p(t)</math> →</p> <p><math>i</math> issues a forecast <math>m</math> for outcome of <math>x</math> →</p> <p>Ex-post: market uses <math>\{m, x\}</math> to update →</p> <p>its beliefs about the analyst's talent. →</p> <p>Let <math>e = x - m</math> be the forecast error →</p> <p><math>t</math> dictates error <math>e</math>'s distribution by pdf <math>\tilde{g}(e t)</math> →</p> <p><math>\tilde{g}(e t) = \frac{1}{2} t \hat{g}(t e)</math> where <math>\hat{g}</math> is a pdf over <math>[0, \infty)</math> →</p> <p>To ensure <math>e \sim N(0, \tau^{-1})</math> priors must satisfy: →</p> <p>(normal pdf) <math>\sqrt{\frac{\tau}{2\pi}} \exp\left\{-\frac{\tau}{2} e^2\right\} = \int_0^\infty \tilde{g}(e t) dP(t)</math> →</p> <p>Likelihood ratio <math>\frac{\tilde{g}(e t)}{\tilde{g}(e t')} \nearrow</math> in <math> e </math> when <math>t &lt; t'</math> →</p> <p>meaning higher types have lower errors →</p> <p>The analyst payoff is defined by <math>w(m x)</math> →</p> <p><math>w(m x) = \int_0^\infty u(t) dP(t m,x) = E^{mkt}[u(t) m,x]</math> →</p> <p><math>u(t) \nearrow</math> in <math>t</math> &amp; <math>P(t m,x)</math>: mkt's posterior over <math>t</math> →</p> <p>Analyst maximizes <math>u(m s) = E[w(m x) s]</math> →</p>	<p>Analyst minimizes his perceived error <math>\hat{e} s \sim N(\hat{s} - E[x s], var(x s))</math></p> <p>so that <math>E[\hat{e}]</math> is minimized at <math>\hat{s}^* = E[x s] = m_h</math>. This gives the forecast:</p> $\underbrace{\frac{\nu}{\nu + \tau} \mu + \frac{\tau}{\nu + \tau} \hat{s}^*}_{m^*} = \left(1 - \left(\frac{\tau}{\nu + \tau}\right)^2\right) \mu + \left(\frac{\tau}{\nu + \tau}\right)^2 s \neq \underbrace{\left(1 - \frac{\tau}{\nu + \tau}\right) \mu + \frac{\tau}{\nu + \tau} s}_{m_h}$ <p>In other words, the honest forecast <math>m_h</math> minimizes <math>E[(m - x)^2 s]</math></p> <p>while the analyst's optimal forecast minimizes <math>E[(\hat{s} - x)^2 s]</math></p> <p>To infer types <math>t</math>, market has conjecture schedule with distribution <math>\varphi(m x, t)</math></p> <p>Denote by <math>S_m</math> the partition of <math>s</math> that issues <math>m</math> under the conjectured strategy, and <math>\varphi(m x, t) = \int_{S_m} d\tilde{G}(s x, t)</math>. Then posterior distribution over types is:</p> $p(t m, x) = \frac{\varphi(m x, t)}{\int_0^\infty \varphi(m x, t) dP(t m, x)}$	<p><b>Two partition case</b> where <math>m \in \{m_h, m_l\}</math> for <math>S_h = s &gt; \tilde{s}</math>, then:</p> $\begin{aligned} \varphi(m_h x, t) &= \int_{\tilde{s}-x}^\infty \tilde{g}(e t) de = \frac{1}{2} \int_{\tilde{s}-x}^\infty t \hat{g}(t e) de \\ &= \frac{1}{2} \left( \int_{\min\{0, \tilde{s}-x\}}^0 t \hat{g}(t e) de + \int_{\max\{0, \tilde{s}-x\}}^\infty t \hat{g}(t e) de \right) \\ &= \frac{1}{2} \left( \int_0^{\max\{0, x-\tilde{s}\}} \hat{g}(te) de + \int_{\max\{0, \tilde{s}-x\}}^\infty \hat{g}(te) de \right) \\ &= \frac{1}{2} \left( \underbrace{\int_0^{\max\{0, t(x-\tilde{s})\}} \hat{g}(z) dz}_{\text{Change of variable to } z=te} + \int_{\max\{0, t(\tilde{s}-x)\}}^\infty \hat{g}(z) dz \right) \\ &= \frac{1}{2} (1 - \hat{G}(\max\{0, t(\tilde{s}-x)\}) + \hat{G}(\max\{0, t(x-\tilde{s})\})) \end{aligned}$ <p><b>Note that</b> <math>\varphi(m_l 2\tilde{s}-x, t) = 1 - \varphi(m_h 2\tilde{s}-x, t)</math></p> $\begin{aligned} &= 1 - \frac{1}{2} (1 - \hat{G}(\max\{0, t(x-\tilde{s})\}) + \hat{G}(\max\{0, t(\tilde{s}-x)\})) \\ &= \frac{1}{2} (1 - \hat{G}(\max\{0, t(\tilde{s}-x)\}) + \hat{G}(\max\{0, t(x-\tilde{s})\})) = \varphi(m_h x, t) \end{aligned}$ <p>This implies that the payoff <math>w(m_h x) = w(m_l 2\tilde{s}-x)</math>.</p> <p>To compute the difference <math>u(m_h s) - u(m_l s)</math>, first compute:</p> $\begin{aligned} u(m_l s) &= \int_{-\infty}^\infty w(m_l x) q(x s) dx \\ &= \int_{-\infty}^{\tilde{s}} w(m_l x) q(x s) dx + \int_{\tilde{s}}^\infty w(m_l x) q(x s) dx \\ &= \int_{\tilde{s}}^\infty \underbrace{w(m_l 2\tilde{s}-z) q(2\tilde{s}-z s) dz}_{\text{Change variable to } z=2\tilde{s}-x} + \int_{\tilde{s}}^\infty w(m_l x) q(x s) dx \\ &= \int_{\tilde{s}}^\infty (w(m_h x) q(2\tilde{s}-x s) + w(m_l x) q(x s)) dx \end{aligned}$ <p>Similarly: <math>u(m_h s) = \int_{\tilde{s}}^\infty (w(m_l x) q(2\tilde{s}-x s) + w(m_h x) q(x s)) dx</math></p> $u(m_h s) - u(m_l s) = \int_{\tilde{s}}^\infty (w(m_h x) - w(m_l x)) (q(x s) - q(2\tilde{s}-x s)) dx$

## Bertomeu & Marinovic (2016)

### 6. Cheap Talk Model (without Partition Equilibrium)

This paper develops a cheap talk model that has a combination of cheap talk ( $s$  for ‘soft’) and verifiable information ( $h$  for ‘hard’). Authors show that certain soft disclosures may contain as much information as hard disclosures. The paper develops a way of dealing with cheap talk without having partitions.

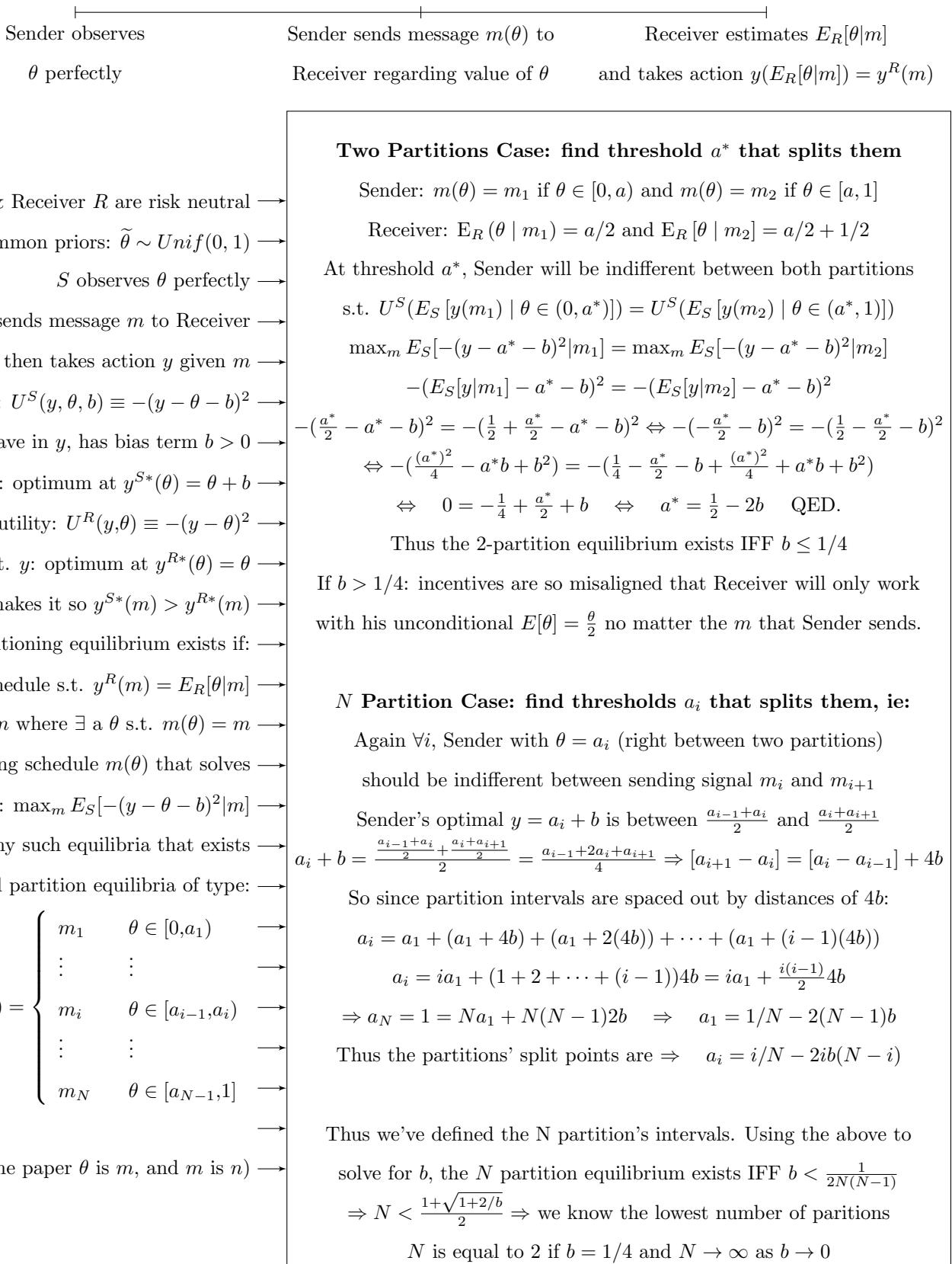


<p>CEO has firm with payoff <math>\pi = h + s \rightarrow</math></p> <p>CEO knows <math>h</math> &amp; <math>s</math>, can convey <math>h</math> at cost <math>c \rightarrow</math></p> <p>CEO can make a cheap-talk report <math>r \rightarrow</math></p> <p>CEO optimizes <math>P(1_h, h, r) = E[\pi   1_h h, r] - 1_h c \rightarrow</math></p> <p>where <math>1_h</math> is an indicator for disclosing <math>h \rightarrow</math></p> <p><math>Prob(CEO \text{ is constrained } [\tau = 1]) = \gamma \rightarrow</math></p> <p><math>Prob(CEO \text{ is unconstrained } [\tau = 0]) = 1 - \gamma \rightarrow</math></p> <p>If CEO is type <math>\tau = 0</math>, <math>r</math> can be anything but if <math>\tau = 1</math>, he can't overreport: <math>r \leq s</math> or <math>r \leq h + s \rightarrow</math></p> <p>Common prior of outcome <math>x \sim N(\mu, v^{-1}) \rightarrow</math></p> <p>CEO has disclosure strategy <math>D(h, s, \tau) \rightarrow</math></p> <p>and reporting strategy <math>R(h, s, \tau) \rightarrow</math></p> <p>Formally his problem is: <math>\max_{d, r} P(d, 1_h h, r) \rightarrow</math></p> <p>s.t. <math>\tau r \leq \tau(1_h s + (1 - 1_h)\pi) \rightarrow</math></p> <p>Market pricing rule is: <math>P(0, 0, r) = \min(r, \lambda_0) \rightarrow</math></p> <p>and <math>P(1, h, r) = h - c + \min(r, z) \rightarrow</math></p> <p>where <math>\lambda_0</math> is max possible price if <math>1_h = 0 \rightarrow</math></p> <p><math>(\Rightarrow \lambda_0 = \max_r P(0, 0, r))</math> and <math>\rightarrow</math></p> <p><math>z</math> is max valuation of soft report if <math>1_h = 1 \rightarrow</math></p> <p><math>(\Rightarrow h - c + z = \max_r P(1, h, r)) \rightarrow</math></p>	<p>Consider the unconstrained CEO's problem of whether to disclose <math>h</math>:</p> <p><math>\tau = 0</math> Unconstrained CEO will disclose <math>h</math> IFF the resulting payoff is <math>\geq</math> some threshold <math>\lambda_0</math>:</p> <p><u>Payoff from issuing <math>h</math></u> <math>\rightarrow h - c + z \geq \lambda_0 \leftarrow</math> <u>Payoff from issuing only <math>s</math></u> <math>\Leftrightarrow h \geq \lambda_0 - z + c \equiv k</math></p> <p>Meanwhile a <math>\tau = 1</math> type Constrained CEO will disclose if <math>h - c + \min(s, z) \geq \min(\lambda_0, h + s)</math></p> <p>Or equivalently,</p> <p>the reporting strategies for disclosing <math>h</math> are:</p> <p><i>Unconstrained</i>: <math>h - c + z \geq \lambda_0</math></p> <p><i>Constrained</i>: <math>h - c + \min\{s, z\} \geq \lambda_0</math></p> <p><b>z solves:</b> <math>z = E[s   \tau = 0 \cup \{\tau = 1, s \geq z\}]</math></p> <p><b>The maximum soft-only value <math>\lambda_0</math> solves:</b></p> <p><math>\lambda_0 = E[\pi   \{\tau = 0, h - c + z &lt; \lambda_0\} \cup</math></p> <p><math>\{\tau = 1, \{\lambda_0 \leq \pi &lt; \lambda_0 + c \text{ or } \underbrace{\lambda_0 + c \leq \pi &lt; \lambda_0 + c + s - z}_{\text{Exists only if } s &gt; z}\}\}]</math></p>
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# Crawford & Sobel (1982)

## 6. Cheap Talk Models (Partition Equilibria)

This paper explores the extent to which information transmission occurs when a sender's report is **not verifiable** and the sender **does not incur a direct cost** associated with a function of the action the receiver takes in response to the report. Full revelation is possible when sender and receiver incentives are perfectly aligned, whereas partial information transmission is possible when they are not too misaligned. An uninformative "babbling equilibrium" occurs when they become too misaligned.



# Caskey (2014)

## 7. Earnings management/Costly signaling (Partially Revealing Equilibrium)

This is an example of a **Fischer and Verrecchia (2000)**-type model where litigation is restricted to cases where investors may have actually suffered harm. The paper illustrates the feedback effect of litigation and the effect of unknown manager incentives to bias.

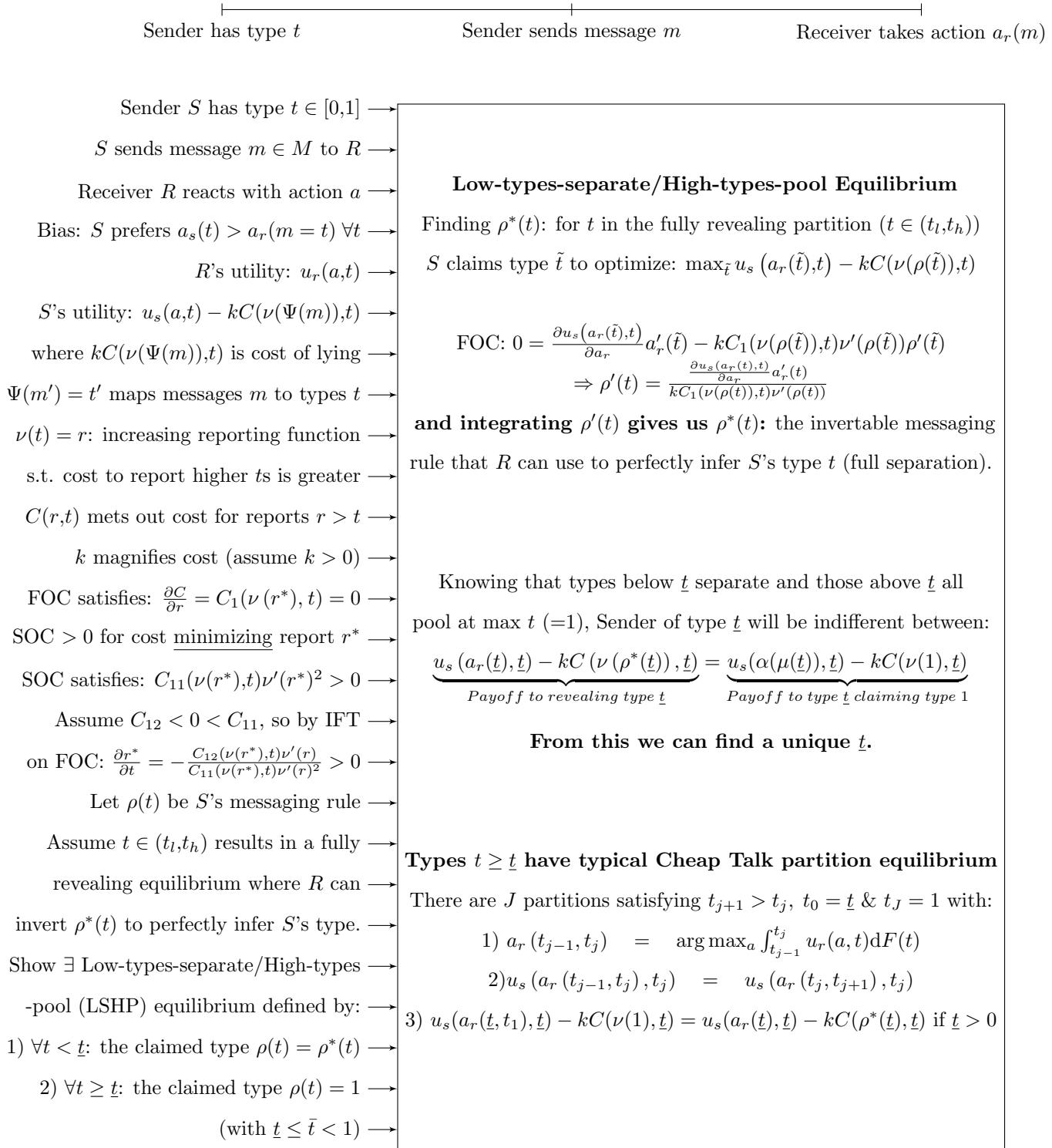
Time 0	Time 1	Time 2	Time 3
Investors observe signal $y$	CEO with bias $b_m$ observes $r$ reports $r_m$ , Time 1 investors buy at price $p_1$	Time 2 information $s$ is released, Time 2 investors buy at price $p_2$	$v$ is realized. If Time 1 investors successfully sue they receive $1 - a \times$ the firm's costs

Firm has terminal cashflow $v \sim N(\mu_v, \tau_v^{-1})$	Solving for $p_2$ :
CEO observes $r$ and makes report $r_m$	$p_2 = E_2[v - \theta \max\{0, p_1 - p_2\}   r_m, y, s] = v_2 - E_2[\theta \max\{0, p_1 - p_2\}   r_m, y, s]$
where $r = v + e_r$ & $e_r \sim N(0, \tau_e^{-1})$	$\Rightarrow p_2 = v_2 - 1_{v_2 < p_1} \frac{E_2[\theta   r_m, y, s]}{1 - E_2[\theta   r_m, y, s]} (p_1 - v_2)$
CEO has private information $b_m$ about his incentives to bias his report $r_m$	(note that $v_2$ is like a valuation of the $v$ at $t = 2$ before litigation is considered)
CEO optimizes:	The price $p_2 = v_2$ whenever $v_2 \geq p_1$ or $1_{p_2 < p_1} E_2[\theta] = 0$
$\max_{r_m} E[bp_1(r_m, y) - \frac{c}{2}(r_m - r)^2   r, b]$	The $\frac{E_2[\theta]}{1 - E_2[\theta]}$ term reflects the feedback effect of litigation: expected lawsuits cause price declines, which increase the expected magnitude of lawsuits.
$c$ is cost of misreporting parameter	<b>Example:</b>
Investors have prior $b \sim N(\mu_b, \sigma_b^2)$	Denote $v_2 = E_2[v] = \frac{\tau_v \mu_v + \tau_e(s_1 + s_2)}{\tau_v + 2\tau_e}$ and $v_1 = E_1[v] = \frac{\tau_v \mu_v + \tau_s \partial_1}{\tau_s + \tau_e}$
$t=1$ investors use $r_m$ & $y$ to price $v$ at $p_1$	Assume: $p_2 = v_2 - 1_{v_2 < p_1} \theta(v_2 - p_1)$ where $\theta \in (0, 1)$
at $t=2$ , public signal $s$ is released	and: $p_1 = E_1[p_2] = v_1 - \theta E_1[1_{v_2 < p_1} (v_2 - p_1)]$
where $y = v + e_y$ , $s = v + e_s$	$var_1(v_2) = \underbrace{\frac{1}{\tau_v + \tau_e}}_{var_1(v)} - \underbrace{\frac{1}{\tau_v + 2\tau_e}}_{var_2(v)} = \frac{1}{\tau_v + \tau_e} \frac{\tau_e}{\tau_v + 2\tau_e}$
$e_y \sim N(0, \tau_y^{-1})$ & $e_s \sim N(0, \tau_s^{-1})$	Using the formulas for expectations of truncated normals, we have:
$t=2$ investors price $v$ at $p_2$ using $s, y, r_m$	$E_1[1_{v_2 < p_1} (v_2 - p_1)] = (v_1 - p_1) \Phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right) - std_1(v_2) \phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right)$
If $p_2 < p_1$ $t=1$ investors can choose to sue	Thus: $p_1 = v_1 - \theta \left( (v_1 - p_1) \Phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right) - std_1(v_2) \phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right) \right)$
thus $p_2$ reflects risk of lawsuit costs:	$\Rightarrow 0 = \theta \left( \frac{p_1 - v_1}{std_1(v_2)} \Phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right) - \phi\left(\frac{p_1 - v_1}{std_1(v_2)}\right) \right) - \frac{p_1 - v_1}{std_1(v_2)} = g\left(\frac{p_1 - v_1}{std_1(v_2)}\right)$
$p_2 = E_2[v - \theta \max\{0, p_1 - p_2\}]$	where $g(x) = \theta(x\Phi(x) - \phi(x)) - x$ , $g'(x) = \theta\Phi(x) - 1 < 0$
$E_2[v] - 1_{p_2 < p_1} E_2[\theta](p_1 - p_2)$ , where $\theta =$	Because $g(x) \rightarrow \infty$ as $x \rightarrow -\infty$ and
$Prob(\text{sue lost}) E[\Delta\% \text{ price} \searrow   \text{sue lost}]$	$g(0) = -\theta\phi(0) < 0$ , there is a unique $x < 0$ that solves $g(x) = 0$ , or,
ie: $\theta$ summarizes expected damages and	equivalently, a $x > 0$ that solves $g(-x) = 0$ , so we can define $p_1$ as:
$E_2[\theta] < 1$ (cannot gain value from suit)	$\frac{p_1 - v_1}{std_1(v_2)} = -\hat{x} \Rightarrow p_1 = v_1 - std_1(v_2) \hat{x}$
$1_{p_2 < p_1}$ is indicator function for $p_2 < p_1$	where $\hat{x}$ depends only on $\theta$ so that $p_1$ is linear in $s_1$ .
note: $v_1 = E_1[v   r_m, y]$ & $v_2 = E_2[v   r_m, y, s]$	<b>Insurance:</b> The model includes the possibility of an insurance policy that covers up to $X$ in damages, in exchange for a premium $x$ :
Investors conjecture CEO reporting rule:	$x = \theta E[1_{v_2 < v_1} 1_{p_2 < p_1} \min\{X, p_1 - p_2\}]$ .
$r_m = \rho_0 + \rho_r r + \rho_b b$	The ex ante firm value is: $E[v] - \theta\alpha \times \text{Expected damages } (D)$
CEO conjectures investor $t=1$ pricing rule:	
$p_1 = E[v   r_m, y] + \pi_0 + \pi_r r_m + \pi_y y$	

## Kartik (2009)

### 7. Earnings management/Costly signaling (Partially Revealing Equilibrium)

This paper introduces lying costs to cheap talk models. Main result is that lying costs make the sender's objective depend directly on his report, which removes some of the cheap talk aspects of the behavior.



[Kartik (2009, S4)] Example with uniform type distribution:  $t \sim Unif(0, 1)$

with  $F(t) = t$ ;  $u_r(a_r, t) = -(a - t)^2$ ;  $u_s = -(a - t - b)^2$ ;  $\nu(t) = t$ ;  $C(x, t) = -(x - t)^2$

Receiver's Problem:  $\max_{a_r} u_r(a_r, t) = \max_{a_r} -(a - t)^2$

FOC:  $\frac{\partial u_r}{\partial a_r} = 0 = -2(a_r - \tilde{t}) \Leftrightarrow a_r = \tilde{t}$

Sender's problem:  $\max_{\tilde{t}} u_s(a_r(\tilde{t}), t) + kC(\nu(\rho(\tilde{t})), t) = \max_{\tilde{t}} -(\tilde{t} - t - b)^2 - k(\rho(\tilde{t}) - t)^2$

FOC (where conjectures are met and  $\tilde{t} = t$ )  $\frac{\partial u_s}{\partial a} = 0 = -2b - 2k(\rho(t) - t)\rho'(t) \Leftrightarrow \rho'(t) = \frac{b}{k(\rho(t) - t)}$

Let  $v(x) = \rho(t) - t$ :

$$\frac{dv}{dt} = \frac{\frac{b}{k} - v}{v} \Leftrightarrow \frac{v}{\frac{b}{k} - v} dv = dt$$

By integrating we get

$$-v - \frac{b}{k} \ln \left( \frac{b}{k} - v \right) = t + \text{constant}$$

Let  $v = \mu - x$ , then

$$-(\rho(t) - t) - \frac{b}{k} \ln \left( \frac{b}{k} - (\rho(t) - t) \right) = t + \text{constant}$$

using  $\rho(0) = 0$  we get  $\text{constant} = -\frac{b}{k} \ln \frac{b}{k}$  and

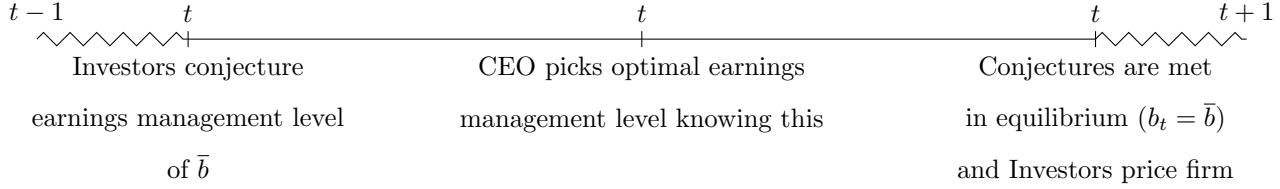
$$\begin{aligned} -\rho(t) - \frac{b}{k} \ln \left( \frac{b}{k} - \rho(t) + t \right) &= -\frac{b}{k} \ln \frac{b}{k} \Leftrightarrow -\frac{k}{b} \rho(t) - \ln \left( \frac{b}{k} - \rho(t) + t \right) = -\ln \frac{b}{k} \\ \Leftrightarrow -\frac{k}{b} \rho(t) + \ln \frac{b}{k} &= \ln \left( \frac{b}{k} - \rho(t) + t \right) \Leftrightarrow \exp \left( -\frac{k}{b} \rho(t) + \ln \frac{b}{k} \right) = \left( \frac{b}{k} - \rho(t) + t \right) \\ \Leftrightarrow \frac{b}{k} \exp \left( -\frac{k}{b} \rho(t) \right) &= \frac{b}{k} - \rho(t) + t \Leftrightarrow \exp \left( -\frac{k}{b} \rho^*(t) \right) = 1 - \frac{k}{b} (\rho^*(t) - t) \end{aligned}$$

Note that this implies inflated language, with  $t = 0$  giving  $\rho(0) = 0$  and  $t > 0$  giving  $\rho(t) > t$  because  $\omega(x) > 0$  for  $x > -e^{-1}$ . The ratio  $\frac{k}{b}$  of lying costs to bias determines the extent of separation. All types  $t < \underline{t} < \bar{t} < 1$  separate, and there is no separation ( $\underline{t} = 0$ ) if  $b \geq k + \frac{1}{4}$ . In other words, when  $b \geq k + \frac{1}{4}$  there is only cheap talk. The cheap-talk partitions look similar to Crawford and Sobel's (1982) uniform/quadratic example (Kartik 2009, p. 1373).

# Stein (1989)

## 7. Costly Signaling (Signal Jamming Equilibrium)

This paper examines the effects of costly lying on manager reporting behavior. It introduces the “myopic” behavior: CEOs always exaggerating their earnings in spite of the fact that investors adjusts for this fact, knowing management always inflates their reports. **Note:** This is a steady-state learning model.



Let  $e_t^n = z_t + v_t$  be natural earnings →

Structural earnings  $z_t$  follow the random walk: →

$$z_t = z_{t-1} + u_t \text{ with } u_t \sim N(0, \sigma_u^2) \rightarrow$$

Thus  $z_t$  has persistent components over time →

vs  $v_t \sim N(0, \sigma_v^2)$  which are time independent →

$z_t$  &  $v_t$  are not observed by CEO or market →

Observable earnings are  $e_t = e_t^n + b_t - c(b_{t-1})$  →

by manipulating  $e_t^n$ , CEO can “borrow”  $b_t$  →

and pay it back as  $c(b_t)$  next period where →

$c(\cdot)$  is a convex (eg: exponential) cost function →

$$c'(0) = (1 + r) \text{ and } r \text{ is firm's discount rate} \rightarrow$$

Firm pays out all earnings  $e_t$  as dividends →

Its market price is  $P_t = E_t^{mkt} \sum_{j=1}^{\infty} \frac{e_{t+j}}{(1+r)^j} j \rightarrow$

Every period CEOs sell  $\pi$  of their shares →

$$\text{CEO utility: } U_t = e_t + \pi P_t + \frac{(1-\pi)e_{t+1}}{(1+r)} \rightarrow$$

In steady state, equilibrium borrowing is  $\bar{b}$  →

Investors have a fixed conjecture on value of  $\bar{b}$  →

So market can reconstruct  $\hat{e}_t^n = e_t + c(\bar{b}) - \bar{b} \rightarrow$

So given past  $\hat{e}_t^n$ , expected future natural →

$$\text{earnings are } E_t(e_{t+k}^n) = \sum_{j=0}^{\infty} \alpha_j \hat{e}_{t-j}^n \quad \forall k > 0 \rightarrow$$

where  $\alpha_j$  sum to one and depend on  $\sigma_u^2$  and  $\sigma_v^2$  →

So coefficient  $\alpha_0$ : impact of  $e_t$  on expectations →

$$\text{can be derived as } \alpha_0 = (k^2/4 + k)^{1/2} - k/2 \rightarrow$$

$$\text{where } k = \frac{\sigma_u^2}{\sigma_v^2} = \frac{\text{permanent earnings drift}}{\text{same-period earnings shock}} \rightarrow$$

$\frac{\sigma_u^2}{\sigma_v^2}$  plays following roles in forming expectations: →

as  $\frac{\sigma_u^2}{\sigma_v^2} = \alpha_0 \rightarrow 0$ : past values more important →

as  $\frac{\sigma_u^2}{\sigma_v^2} = \alpha_1 \rightarrow 0$ : past values less important →

Looking at CEO's problem at time  $t$ , taking

$\bar{b}$  as being fixed by market's conjecture:

$$\max_{b_t} U_t = e_t + \pi P_t + \frac{(1-\pi)e_{t+1}}{(1+r)} \rightarrow$$

$$\text{FOC: } \frac{de_t}{db_t} + \pi \frac{dP_t}{db_t} + \left( \frac{1-\pi}{1+r} \right) \frac{de_{t+1}}{db_t} = 0$$

$$1 + \pi \frac{\alpha_0}{r} + \frac{1-\pi}{1+r} (-c'(b_t))$$

$$c'(b_t) = (1 + \pi \frac{\alpha_0}{r}) \frac{1+r}{1-\pi}$$

This setting  
leads to a  
→ steady-state  
signal jamming  
equilibrium

In equilibrium conjectures are fulfilled, thus  $b_t = \bar{b}$ :

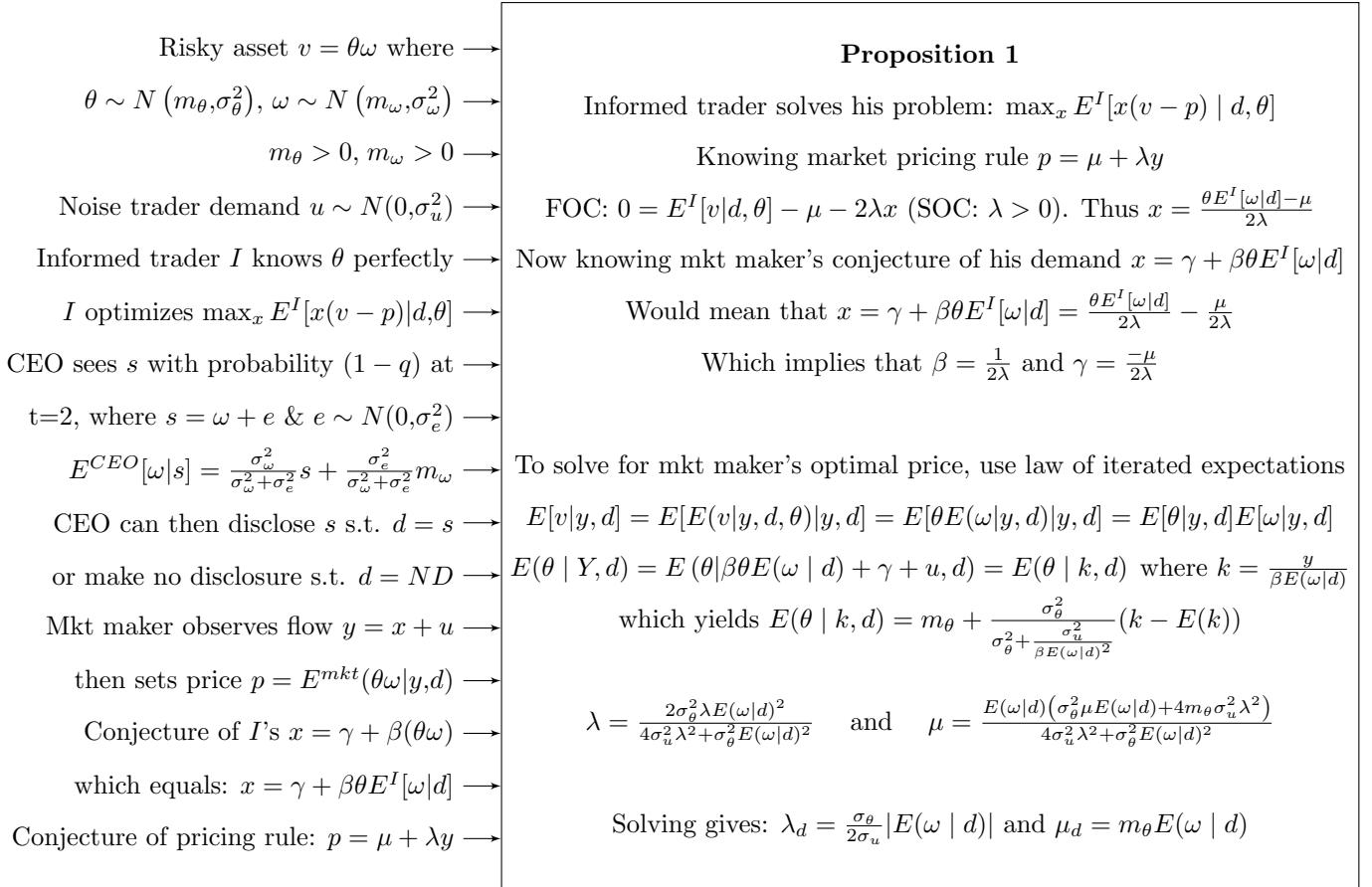
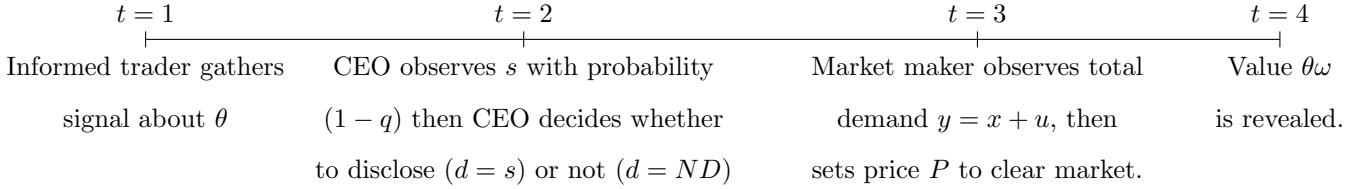
$$c'(\bar{b}) = (1 + \pi \frac{\alpha_0}{r}) \frac{1+r}{1-\pi}$$

Thus in a steady state signal-jamming equilibrium,  
managers will borrow a constant amount  $\bar{b}$  each period  
from next period's earnings  $e_t + 1$  and the market  
will correctly anticipate this borrowing.

## Cheynel & Levine(2020)

### 8. Voluntary + mandatory disclosure (Rational Expectations Model)

This paper examines the interaction between voluntary public disclosure and private information in a setting where the two types of signals interact. The firm's payoff  $v = \theta\omega$  where an investor has private information about  $\theta$  and the firm can release public information about  $\omega$ .



**Law of Iterated Expectations (also known as Law of Total Expectation):**

For a bivariate random variables, X and Y :

$$E[Y] = E[E[Y | X]]$$

For a trivariate random variables, X, Y and Z where  $E[Y | X, Z]$  = conditional mean given X and Z :

$$E[Y] = E[E(Y | X, Z)]$$

$$E[Y | Z] = E[E(Y | X, Z) | Z]$$

$$E[Y | X] = E[E(Y | X, Z) | X]$$

$$E[Y] = E[E[E(Y | X, Z) | Z]]$$

# Einhorn (2018)

## 8. Voluntary + mandatory disclosure (Rational Expectations Model)

The author's Kyle (1985) based model uses competing public & private signals and discretionary manager disclosure as a reason for why firms withhold verifiable information to favor private information collection.

CEO perfectly knows but cannot credibly disclose $v$ .	Public signal $\tilde{s}_2$ . Then Speculator decides whether to additionally buy private signal $\tilde{s}_3$ for cost $c$ .	Trading. Market maker observes demand flow $y = x + z$ and sets price to clear market.	Value $\tilde{v}$ is revealed.
But he decides whether to disclose noisy signal $\tilde{s}_1$ .			

Prior for risky firm value  $\tilde{v}$  is diffuse →

Liquidity trader demand  $\tilde{z} \sim N(0, \sigma_z^2)$  →

Risk **neutral** Speculator has demand  $x$  →

Risk averse CEO has CARA coefficient  $a$  →

CEO perfectly knows but can't disclose  $v$  →

CEO has noisy private signal  $\tilde{s}_1 = \tilde{v} + \tilde{\epsilon}_1$  →

CEO can choose whether to disclose  $\tilde{s}_1$  →

If so then mkt's posterior  $\tilde{v}|s_1 \sim N(s_1, \frac{1}{h_1})$  →

Afterwards public signal  $s_2$  is released →

Then Speculator decides whether to →

additionally buy private signal  $\tilde{s}_3$  for cost  $c$  →

$\tilde{s}_i = \tilde{v} + \tilde{\epsilon}_i$  and  $\tilde{\epsilon}_i \sim N(0, \frac{1}{h_i})$  for  $i = 1, 2, 3$  →

Mkt maker linear price conjecture  $p = \alpha + \beta y$  →

clears mkt after seeing total demand  $y = x + z$  →

CEO maxes mkt price  $p$ :  $E[p] - \frac{a}{2} var(p)$  →

CEO doesn't know  $\tilde{s}_2$  &  $\tilde{s}_3$  beforehand →

We conjecture that Speculator will only find →

$s_3$  to be worth buying if  $s_1$  was not disclosed →

& his demand  $x > 0$  IFF he buys  $s_3$ , else  $x = 0$  →

This implies that  $c_L < c < c_H$ , where  $c_L$  bound →

prevents him from buying when  $s_1$  is disclosed →

and  $c_H$  is the max he'll be willing to pay for  $s_3$  →

He maximized  $E^S[x(\tilde{v} - (\alpha + \beta(x + \tilde{z})))|s_2, s_3]$  →

**Additional Definitions & Notes:** →

Let  $F(x) = \frac{\phi(x)}{\Phi(x)}$  for std normal pdf  $\phi$  & cdf  $\Phi$  →

Note on FOC/SOC: Given  $\max_x U(x)$  →

(all w.r.t.  $x$ )  $FOC : \frac{dU}{dx} = U'(x) = 0$  →

$SOC : \frac{d^2U}{dx^2} = U''(x) < 0$  for maximization →

$SOC : \frac{d^2U}{dx^2} = U''(x) > 0$  for minimization →

**Case 1: No competing information benchmark ( $h_2 = h_3 = 0$ )**

Here signals  $s_2$  and  $s_3$  are uninformative. Speculator never buys  $\tilde{s}_3$ .

“Unraveling”: CEO discloses  $s_1$  and price  $p = E^{mkt}[\tilde{v} | \tilde{s}_1 = s_1] = s_1$

By Grossman (1981) if he did not disclose, off-equilibrium  $p = -\infty$ .

**Case 2: Competing public but not private info ( $h_2 > 0; h_3 = 0$ )**

Here signal  $s_3$  is uninformative, so Speculator never buys  $\tilde{s}_3$ . If CEO discloses  $s_1$ , the market conjectures that it must have exceeded some

threshold  $v - \lambda^0$  for some scalar  $\lambda^0 \in \mathbb{R}$  s.t.  $v \leq s_1 + \lambda^0$ . Thus:

$$E^{mkt}[\hat{v} | \tilde{s}_1 = s_1, \tilde{s}_2 = s_2, \tilde{v} \leq s_1 + \lambda^0]$$

using Inverse Mills Ratio:  $E[X | X < \alpha] = \mu - \sigma \frac{\phi(\frac{\alpha-\mu}{\sigma})}{\Phi(\frac{\alpha-\mu}{\sigma})} = \mu - \sigma F(\frac{\alpha-\mu}{\sigma})$

$$E[X | X < \alpha] = E[\hat{v} | s_1, s_2] | \tilde{v} \leq s_1 + \lambda^0]$$

$$= E[\hat{v} | s_1, s_2] - \sqrt{var(\hat{v} | s_1, s_2)} \cdot F\left(\frac{(s_1 + \lambda^0) - E[\hat{v} | s_1, s_2]}{\sqrt{var(\hat{v} | s_1, s_2)}}\right)$$

$$\text{Mkt price } p = \frac{h_1 s_1 + h_2 s_2}{h_1 + h_2} - \frac{1}{\sqrt{h_1 + h_2}} F\left(s_1 + \lambda^0 - \frac{h_1 s_1 + h_2 s_2}{h_1 + h_2}\right) \sqrt{h_1 + h_2}$$

under disclosure, and under non-disclosure:  $p = E^{mkt}[\tilde{v} | s_2] = s_2$ . Note that

CEO doesn't know the realization of  $\tilde{s}_2$  when deciding whether to disclose  $s_1$ .

By working with CEO's expectations of  $\tilde{s}_2$  and equating his expected risk adjusted utility,  $\exists$  a unique  $\lambda_0$  yielding a partial signaling equilibrium.

**Case 3: Competing public and private info ( $h_2 > 0; h_3 > 0$ )**

Speculator maxes  $E^S[x(\tilde{v} - (\alpha + \beta(x + \tilde{z})))|s_2, s_3] = xE^S[\tilde{v}|s_2, s_3] - \alpha x - \beta x^2$  →

$$= x \frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha x - \beta x^2 \Rightarrow FOC: \frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha - 2\beta x = 0, SOC: -2\beta < 0$$

This yields optimal  $x^* = (2\beta)^{-1} \left( \frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha \right)$ . Plugging this into his

pricing rule, market maker infers  $(2\beta)^{-1} \left( \frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha \right) + \tilde{z} = y$

$$\Leftrightarrow \tilde{s}_3 + 2\beta \frac{h_2 + h_3}{h_3} \tilde{z} = -\frac{h_2}{h_3} s_2 + \alpha \frac{h_2 + h_3}{h_3} + 2\beta \frac{h_2 + h_3}{h_3} y$$

He sets  $p = E^{mkt}[\hat{v} | \tilde{s}_3 + 2\beta \frac{h_2 + h_3}{h_3} \tilde{z} = -\frac{h_2}{h_3} s_2 + \alpha \frac{h_2 + h_3}{h_3} + 2\beta \frac{h_2 + h_3}{h_3} y, \tilde{s}_2 = s_2]$

$$\text{price } p = \frac{\left(\frac{1}{h_3} + \left(2\beta \frac{h_2 + h_3}{h_3} \sigma_z\right)^2\right)^{-1} \left(-\frac{h_2}{h_3} s_2 + \alpha \frac{h_2 + h_3}{h_3} + 2\beta \frac{h_2 + h_3}{h_3} y\right) + h_2 s_2}{h_2 + \left(\frac{1}{h_3} + \left(2\beta \frac{h_2 + h_3}{h_3} \sigma_z\right)^2\right)^{-1}}$$

*continued on next page.*

$$\alpha = \frac{\left(\frac{1}{h_3} + \left(2\beta \frac{h_2+h_3}{h_3} \sigma_z\right)^2\right)^{-1} \left(-\frac{h_2}{h_3} s_2 + \alpha \frac{h_2+h_3}{h_3}\right) + h_2 s_2}{h_2 + \left(\frac{1}{h_3} + \left(2\beta \frac{h_2+h_3}{h_3} \sigma_z\right)^2\right)^{-1}} \quad \text{and} \quad \beta = \frac{\left(\frac{1}{h_3} + \left(2\beta \frac{h_2+h_3}{h_3} \sigma_z\right)^2\right)^{-1} 2\beta \frac{h_2+h_3}{h_3}}{h_2 + \left(\frac{1}{h_3} + \left(2\beta \frac{h_2+h_3}{h_3} \sigma_z\right)^2\right)^{-1}}$$

$$\alpha = s_2 \quad \text{and} \quad \beta = \frac{1}{2\sigma_z} \cdot \sqrt{\frac{h_3}{h_2(h_2+h_3)}}$$

Plugging these into aggregate demand  $y = (2\beta)^{-1} \left(\frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha\right) + \tilde{z}$  and pricing coefficient  $\alpha = s_2$  into the stock pricing rule  $p = \alpha + \beta y$  implies  $p = \frac{1}{2}s_2 + \frac{1}{2} \frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} + \beta z$

$$\Leftrightarrow p = \frac{(h_2 + 1/2h_3)s_2 + 1/2h_3 \left(s_3 + \sigma_z^{-1} \sqrt{h_2^{-1} + h_3^{-1}} \cdot z\right)}{h_2 + h_3}$$

Also Speculator's optimal  $x^* = (2\beta)^{-1} \left(\frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha\right) = \sigma_z \cdot \left(\frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - s_2\right) \cdot \sqrt{\frac{h_2(h_2+h_3)}{h_3}}$

**Find upper bound  $c_H$  knowing Speculator will acquire  $s_3$  IFF  $s_1$  is not disclosed, ie:  $c < c_H$**

The max cost  $c_H$  he is willing to pay is his maximum expected profit (given he's risk neutral), which is:

Given his demand  $x = \frac{E^S[\tilde{v}|s_2, s_3] - \alpha}{2\beta}$ , his expected utility is  $E^S[U^S] = E^S[x(\tilde{v} - (\alpha + \beta(x + \tilde{z})))|s_2, s_3]$

$$E^S[U^S|D] = \frac{E[\tilde{v}|s_2, s_3] - \alpha}{2\beta} \left( E[\tilde{v}|s_2, s_3] - \left( \alpha + \beta \left( \frac{E[\tilde{v}|s_2, s_3] - \alpha}{2\beta} + \underbrace{E^S[\tilde{z}]}_{=0} \right) \right) \right)$$

$$E^S[U^S|D] = \frac{E[\tilde{v}|s_2, s_3] - \alpha}{2\beta} \left( E[\tilde{v}|s_2, s_3] - \alpha - \frac{E[\tilde{v}|s_2, s_3] - \alpha}{2} \right) = \frac{(E[\tilde{v}|s_2, s_3] - \alpha)^2}{4\beta} = \frac{\left(\frac{h_2 s_2 + h_3 s_3}{h_2 + h_3} - \alpha\right)^2}{4\beta}$$

or  $U^S = \frac{\sigma_z}{2} \cdot \frac{h_3}{h_2 + h_3} \sqrt{\frac{h_2 h_3}{h_2 + h_3}} (\tilde{s}_3 - s_2)^2$  after plugging in  $\alpha$  &  $\beta$ . However since he needs to make his decision of whether to buy  $s_3$  before observing it, his expected utility is  $E^S[U^S|D]$  before cost  $c$ .

$$E^S[U^S|D] = \frac{\sigma_z}{2} \cdot \frac{h_3}{h_2 + h_3} \sqrt{\frac{h_2 h_3}{h_2 + h_3}} E[(\tilde{s}_3 - s_2)^2 | s_2] = \frac{\sigma_z}{2} \cdot \frac{h_3}{h_2 + h_3} \sqrt{\frac{h_2 h_3}{h_2 + h_3}} \text{var}(\tilde{s}_3 | s_2) = \frac{\sigma_z}{2} \cdot \sqrt{\frac{h_3}{h_2(h_2 + h_3)}}$$

This implies that the max cost  $c_H$  he's be willing to pay is equal to  $\frac{\sigma_z}{2} \cdot \sqrt{\frac{h_3}{h_2(h_2 + h_3)}}$ .

**Find lower bound  $c_L$  knowing Speculator won't acquire  $s_3$  when  $s_1$  is disclosed, ie:  $c_L < c$**

This implies that his expected profit under disclosure of  $s_1$  is at least 0 if he buys  $s_3$ :

Starting again with his expected profit:  $E[x(\tilde{v} - (\alpha + \beta(x + \tilde{z})))|s_1, s_2, s_3]$  where  $x = \frac{E[\tilde{v}|s_1, s_2, s_3] - \alpha}{2\beta}$

Linear pricing rule conjecture coefficients become:  $\alpha = \frac{h_1 s_1 + h_2 s_2}{h_1 + h_2}$  and  $\beta = \frac{1}{2\sigma_z} \cdot \sqrt{\frac{h_3}{(h_1 + h_2)(h_1 + h_2 + h_3)}}$

$$E^S[U^S|ND \text{ with } s_3] = \frac{\left(\frac{h_1 s_1 + h_2 s_2 + h_3 s_3}{h_1 + h_2 + h_3} - \alpha\right)^2}{4\beta} \text{ or } \frac{\sigma_z}{2} \cdot \frac{h_3}{h_1 + h_2 + h_3} \sqrt{\frac{h_3(h_1 + h_2)}{h_1 + h_2 + h_3}} \cdot \left(s_3 - \frac{h_1 s_1 + h_2 s_2}{h_1 + h_2}\right)^2 =$$

$$\frac{\sigma_z}{2} \cdot \frac{h_3}{h_1 + h_2 + h_3} \sqrt{\frac{h_3(h_1 + h_2)}{h_1 + h_2 + h_3}} \cdot E\left[\left(\tilde{s}_3 - \frac{h_1 s_1 + h_2 s_2}{h_1 + h_2}\right)^2 | s_1, s_2\right] = \frac{\sigma_z}{2} \cdot \frac{h_3}{h_1 + h_2 + h_3} \sqrt{\frac{h_3(h_1 + h_2)}{h_1 + h_2 + h_3}} \cdot \text{var}(\tilde{s}_3 | s_1, s_2)$$

$= \frac{\sigma_z}{2} \cdot \sqrt{\frac{h_3}{(h_1 + h_2)(h_1 + h_2 + h_3)}}$   $\Rightarrow$  his expected profit when he has access to  $s_1$  &  $s_3$ . Thus to respect the conjecture,  $c_L$  must be equivalent to this such that paying  $c_L$  for  $s_3$  is unprofitable when  $s_1$  is disclosed.

# Pastor & Veronesi (2003)

## 10. Continuous time dynamic models

The basic idea of this paper is that firm value is convex in growth, so that uncertainty about growth leads to higher multiples.

- The book value of equity  $b_t$  follows the process:  $db_t = \left( \underbrace{y_t}_{Earnings} - \underbrace{d_t}_{dividends} \right) dt = (\rho_t - c) b_t dt$   
where  $\rho_t$  and  $c$  are the instantaneous earnings and dividend rates.
- Profits follow the mean-reverting process:  $d\rho_t = \phi(\bar{\rho} - \rho_t) dt + \boldsymbol{\sigma}'_{\rho} d\mathbf{w}_t$
- where  $\boldsymbol{\sigma}_{\rho} = \{\sigma_{\rho 1}, \sigma_{\rho 2}\}$  is the two-dimensional vector of volatilities
- $\mathbf{w}_t$  is a two-dimensional standard Brownian motion.
- The first component represents systematic risks that are correlated with the later-specified, exogenous stochastic discount factor, and the second component represents idiosyncratic risks. Note that the drift is zero when  $\rho_t = \bar{\rho}$ , is positive when  $\rho_t < \bar{\rho}$ , and is negative when  $\rho_t > \bar{\rho}$ .
- Solving:  $d\rho_t = \phi(\bar{\rho} - \rho_t) dt + \boldsymbol{\sigma}'_{\rho} d\mathbf{w}_t \Leftrightarrow de^{\phi t} \rho_t = e^{\phi t} \phi(\bar{\rho} - \rho_t) dt + \boldsymbol{\sigma}'_{\rho} e^{\phi t} d\mathbf{w}_t$

– Applying Ito's Lemma to  $dX_t = \mu_t dt + \sigma_t dB_t$

$$\Rightarrow df(t, X_t) = \left( \frac{\partial f}{\partial t} + \mu_t \frac{\partial f}{\partial x} + \frac{\sigma_t^2}{2} \frac{\partial^2 f}{\partial x^2} \right) dt + \sigma_t \frac{\partial f}{\partial x} dB_t$$

– Here:  $X_t = \rho_t$ ;  $\mu_t = \phi(\bar{\rho} - \rho_t)$ ;  $B_t = \mathbf{w}_t$ ;  $f(t, X_t) = f(t, \rho_t) = e^{\phi t} \rho_t$

$$Thus: df(t, \rho_t) = \left( \frac{\partial f}{\partial t} + \phi(\bar{\rho} - \rho_t) \frac{\partial f}{\partial \rho} + \frac{\sigma_t' \sigma_t}{2} \frac{\partial^2 f}{\partial \rho^2} \right) dt + \sigma_t' \frac{\partial f}{\partial p} dw_t$$

$$de^{\phi t} \rho_t = \left( \phi e^{\phi t} \rho_t + \phi(\bar{\rho} - \rho_t) e^{\phi t} + \frac{\sigma_t' \sigma_t}{2} \underbrace{\frac{\partial^2 f}{\partial \rho^2}}_{=0} \right) dt + \sigma_t' e^{\phi t} dw_t = \bar{\rho} \phi e^{\phi t} dt + \sigma_t' e^{\phi t} dw_t$$

$$Integrating over (0, t): e^{\phi t} \rho_t - e^{\phi 0} \rho_0 = \phi \bar{\rho} \int_0^t e^{\phi s} ds + \sigma_t' \int_0^t e^{\phi s} dw_s$$

$$e^{\phi t} \rho_t - \rho_0 = \phi \bar{\rho} \left( \frac{[e^{\phi s}]_0^t}{\phi} \right) + \sigma_t' \int_0^t e^{\phi s} dw_s = \bar{\rho}[e^{\phi t} - 1] + \sigma_t' \int_0^t e^{\phi s} dw_s$$

$$\frac{e^{\phi t}}{e^{\phi t}} \rho_t = \frac{\rho_0}{e^{\phi t}} + \bar{\rho} \frac{[e^{\phi t} - 1]}{e^{\phi t}} + \sigma_t' \int_0^t e^{\phi s} dw_s \Leftrightarrow \rho_t = e^{-\phi t} \rho_0 + (1 - e^{-\phi t}) \bar{\rho} + \int_0^t e^{-\phi(t-s)} \boldsymbol{\sigma}'_{\rho} d\mathbf{w}_s$$

- or, equivalently, for  $T > t$ :  $\rho_T = e^{-\phi(T-t)} \rho_t + (1 - e^{-\phi(T-t)}) \bar{\rho} + \int_t^T e^{-\phi(T-s)} \boldsymbol{\sigma}'_{\rho} d\mathbf{w}_s$

- Note that the expected value drifts towards the long-run mean of  $\bar{\rho}$ , with  $\phi$  denoting the speed of convergence.

- Firm value  $m_t$  is the present value of future expected dividends that, given the exogenous stochastic discount factor  $\pi_t$ , equals:  $m_t = E_t \left[ \int_t^T \frac{\pi_s}{\pi_t} c b_s ds \right] + E_t \left[ \frac{\pi_T}{\pi_t} b_T \right]$  where the second term reflects that the market value  $m_T$  equals book value  $b_T$  at some known time  $T$ .

- The stochastic discount factor follows a geometric Brownian motion, where  $\sigma_\pi = \{\sigma_{\pi 1}, 0\}$  so that the first component of  $w_t$  represents systematic risk. The solution to  $d\pi_t = -r\pi_t dt - \pi_t \sigma'_\pi d\mathbf{w}_t$  is:
$$\pi_t = e^{-rt} \pi_0 \exp \left\{ - \left( \frac{1}{2} \sigma_{\pi 1}^2 t + \int_0^t \sigma'_\pi d\mathbf{w}_s \right) \right\}$$
- or, equivalently, for  $T > t : \pi_T = e^{-r(T-t)} \pi_t \exp \left\{ - \left( \frac{1}{2} \sigma_{\pi 1}^2 (T-t) + \int_t^T \sigma'_\pi d\mathbf{w}_s \right) \right\}$
- Note that  $E_t \left[ \frac{\pi_T}{\pi_t} \right] = e^{-r(T-t)}$  reflects only the risk-free rate  $r$

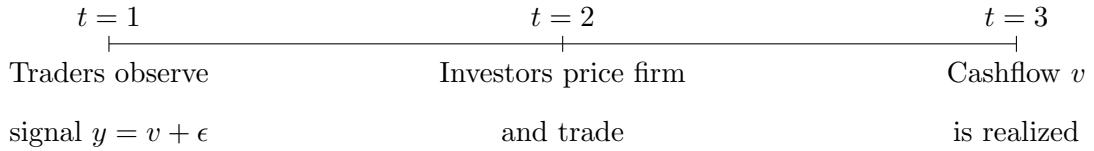
Additional notes:

1.  $\frac{M_t}{B_t} = c \int_0^\tau z(\bar{\rho}, \rho_t, s) ds + z(\bar{\rho}, \rho_t, \tau) \rightarrow$  This expression is just a statement that the market value equals the risk-adjusted present value of dividends. The ‘proposition’ part is obtaining the explicit expression  $z$  in terms of primitives.
2. The volatilities  $\sigma'_\rho \sigma_\rho = \sigma_{\rho 1}^2 + \sigma_{\rho 2}^2$  and  $\sigma'_\pi \sigma_\rho = \sigma_{\pi 1} \sigma_{\rho 1}$  affect  $z$  only via the  $q$  term:  

$$q(s) = \frac{\sigma'_\rho \sigma_\rho}{2\phi^3} \underbrace{\left( \frac{1 - e^{-2\phi s}}{2} + \phi s - 2(1 - e^{-\phi s}) \right)}_{>0} + \frac{\sigma'_\pi \sigma_\rho}{\phi^2} \underbrace{(1 - e^{-\phi s} - \phi s)}_{<0}$$
  
so that  $q(s)$ , and therefore  $z(\bar{\rho}, \rho_t, s)$ , is increasing in the volatility of profitability  $\sigma'_\rho \sigma_\rho$  and decreasing in systematic volatility  $\sigma'_\pi \sigma_\rho$
3. Current profitability  $\rho_t$  clearly increases  $z$ , while average profitability  $\bar{\rho}$  impacts  $z$  by a factor proportional to  $1 - \frac{1 - e^{-\phi s}}{\phi s} > 0$
4. The case with unknown mean profitability  $\bar{\rho}$  is a straightforward application of the Kalman filter.  
In this particular case, the uncertainty is deterministic.
5. Unknown mean profitability amplifies volatility. When  $\bar{\rho}$  is known, a high profit reflects  $\rho_t$  that will increase prices by a small amount because the effect dissipates over time as can be seen in expression:  $\rho_T = e^{-\phi(T-t)} \rho_t + (1 - e^{-\phi(T-t)}) \bar{\rho} + \int_t^T e^{-\phi(T-s)} \sigma'_\rho d\mathbf{w}_s$ . When  $\bar{\rho}$  is unknown, a high profit also increases beliefs about  $\bar{\rho}$ , which has persistent effects.
6. Learning has no effect on expected returns, consistent with the notion that expected returns do not depend on firm-specific information (e.g., Caskey et al. 2015)

# Homework 1

*Accounting Theory with Brett Trueman*



*What will be the market price at date 2?*

**By using properties of normal distributions:**

$$E[v|y = 7] = 5 + \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2} (7 - 5)$$

**Or by scaling signals by their weighted average precisions:**

$$E[v|y = 7] = \frac{1/\sigma^2}{1/\sigma^2 + 1/\sigma_\epsilon^2} 5 + \frac{1/\sigma_\epsilon^2}{1/\sigma^2 + 1/\sigma_\epsilon^2} 7$$

**Which we show to be equivalent here:**

$$\begin{aligned} E[v|y = 7] &= \frac{1/\sigma^2}{1/\sigma^2 + 1/\sigma_\epsilon^2} 5 + \frac{1/\sigma_\epsilon^2}{1/\sigma^2 + 1/\sigma_\epsilon^2} 7 = \frac{1}{1 + \sigma^2/\sigma_\epsilon^2} 5 + \frac{1}{\sigma_\epsilon^2/\sigma^2 + 1} 7 \\ &= \frac{\sigma_\epsilon^2}{\sigma^2 + \sigma_\epsilon^2} 5 + \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2} 7 = (1 - \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2}) 5 + \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2} 7 = 5 + \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2} (7 - 5) \end{aligned}$$

Risky asset prior  $v \sim N(5, \sigma^2) \rightarrow$

Riskless asset yields zero return  $\rightarrow$

Traders are all risk **neutral**  $\rightarrow$

At  $t=1 \rightarrow$

all traders see  $y = v + \epsilon \rightarrow$

where  $\epsilon \sim N(0, \sigma_\epsilon^2)$  and  $\epsilon \perp v \rightarrow$

Assume traders saw  $y = 7 \rightarrow$

*Effect on market price from increase in  $\sigma_2$ :*

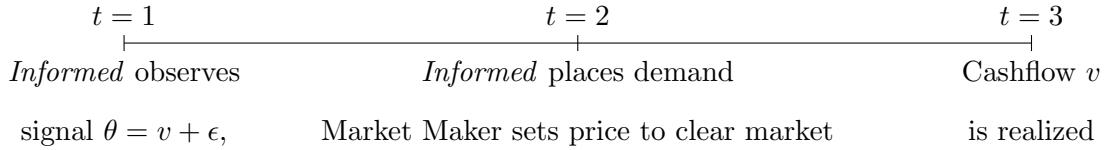
With  $P = E[v|y = 7] = 5 + \frac{\sigma^2}{\sigma^2 + \sigma_\epsilon^2} (2)$

$$\frac{\partial P}{\partial \sigma_2} = \frac{2}{\sigma^2 + \sigma_\epsilon^2} + \frac{-2\sigma^2}{(\sigma^2 + \sigma_\epsilon^2)^2} = \frac{2\sigma_\epsilon^2}{(\sigma^2 + \sigma_\epsilon^2)^2} > 0$$

Price goes up as investors put more weight on the now relatively more precise higher signal ( $y = 7$ ) rather than the now relatively more imprecise lower prior (5).

## Homework 2: Kyle (1985)

*Accounting Theory with Brett Trueman*



**Starting with the *Informed* trader's optimization problem:**

$$\max_x E[W_3] - \frac{a}{2} \text{var}(W_3)$$

$$\max_x E[W_1 + x(v - P)|\theta] - \frac{a}{2} \text{var}(W_1 + x(v - P)|\theta)$$

$$\max_x W_1 + xE[v - P|\theta] - \frac{a}{2} \text{var}(x(v - P)|\theta)$$

$$\max_x W_1 + xE[v - (\mu + \lambda y)|\theta] - \frac{a}{2} x^2 \text{var}(v - (\mu + \lambda y)|\theta)$$

$$\max_x W_1 + xE[v - (\mu + \lambda(x + u))|\theta] - \frac{a}{2} x^2 \text{var}(v - \mu - \lambda(x + u)|\theta)$$

$$\max_x W_1 + xE[v|\theta] - xE[\mu + \lambda x|\theta] - \frac{a}{2} x^2 \text{var}(v|\theta) - \frac{a}{2} x^2 \text{var}(\lambda(x + u)|\theta)$$

$$\max_x W_1 + xE[v|\theta] - x\mu - \lambda xE[x|\theta] - \frac{a}{2} x^2 \text{var}(v|\theta) - \frac{a}{2} x^2 \text{var}(\lambda u|\theta)$$

$$\max_x W_1 + xE[v|\theta] - x\mu - \lambda x^2 - \frac{a}{2} x^2 \text{var}(v|\theta) - \frac{a}{2} x^2 \lambda^2 \sigma_u^2$$

$$\max_x W_1 + x(E[v|\theta] - \mu) - \lambda x^2 - \frac{a}{2} x^2 (\text{var}(v|\theta) + \lambda^2 \sigma_u^2)$$

$$\text{FOC: } 0 = E[v|\theta] - \mu - 2\lambda x - ax (\text{var}(v|\theta) + \lambda^2 \sigma_u^2)$$

$$0 = E[v|\theta] - \mu - x (2\lambda + a (\text{var}(v|\theta) + \lambda^2 \sigma_u^2))$$

$$x = \frac{E[v|\theta] - \mu}{2\lambda + a (\text{var}(v|\theta) + \lambda^2 \sigma_u^2)}$$

$$E[v|\theta] = E[v] + \frac{\text{cov}(v,\theta)}{\text{var}(\theta)} (\theta - E[\theta]) = \bar{v} + \frac{\text{cov}(v,v+\epsilon)}{\text{var}(v+\epsilon)} (\theta - E[v + \epsilon])$$

$$E[v|\theta] = \bar{v} + \frac{\sigma_v^2}{\sigma_v^2 + \sigma_\epsilon^2} (\theta - \bar{v})$$

$$\text{var}(v|\theta) = \text{var}(v) - \frac{\text{cov}^2(v,\theta)}{\text{var}(\theta)} = \sigma_v^2 - \frac{\sigma_v^4}{\sigma_v^2 + \sigma_\epsilon^2} = \frac{\sigma_v^2 \sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2}$$

$$x = \frac{\bar{v} + \frac{\sigma_v^2}{\sigma_v^2 + \sigma_\epsilon^2} (\theta - \bar{v}) - \mu}{2\lambda + a \left( \frac{\sigma_v^2 \sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} + \lambda^2 \sigma_u^2 \right)} = \frac{\frac{\sigma_v^2}{\sigma_v^2 + \sigma_\epsilon^2} \theta + \frac{\sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} \bar{v} - \mu}{2\lambda + a \left( \frac{\sigma_v^2 \sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} + \lambda^2 \sigma_u^2 \right)}$$

$$\beta = \frac{\frac{\sigma_v^2}{\sigma_v^2 + \sigma_\epsilon^2}}{2\lambda + a \left( \frac{\sigma_v^2 \sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} + \lambda^2 \sigma_u^2 \right)} \quad \alpha = \frac{\frac{\sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} \bar{v} - \mu}{2\lambda + a \left( \frac{\sigma_v^2 \sigma_\epsilon^2}{\sigma_v^2 + \sigma_\epsilon^2} + \lambda^2 \sigma_u^2 \right)}$$

Now working with the pricing rule:

$$P = E[v|y] = E[v] + \frac{\text{cov}(v,y)}{\text{var}(y)} = \bar{v} + \frac{\text{cov}(v,x+u)}{\text{var}(x+u)} (y - E[y])$$

$$P = \bar{v} + \frac{\text{cov}(v,x+u)}{\text{var}(x+u)} (y - E[x + u]) = \bar{v} + \frac{\text{cov}(v,x+u)}{\text{var}(x+u)}$$

$$P = \bar{v} + \frac{\text{cov}(v,\alpha + \beta\theta + u)}{\text{var}(\alpha + \beta\theta + u)} (y - E[\alpha + \beta\theta])$$

$$P = \bar{v} + \frac{\text{cov}(v,\beta\theta)}{\text{var}(\beta\theta) + \text{var}(u)} (y - \alpha - \beta E[\theta])$$

$$P = \bar{v} + \frac{\beta \text{cov}(v,v+\epsilon)}{\beta^2 \text{var}(v+\epsilon) + \sigma_u^2} (y - \alpha - \beta E[v + \epsilon]) = \bar{v} + \frac{\beta \sigma_v^2}{\beta^2 (\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2} (y - \alpha - \beta \bar{v})$$

$$P = \bar{v} + \frac{\beta \sigma_v^2}{\beta^2 (\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2} (y - \alpha - \beta \bar{v})$$

$$\text{Thus: } \mu = \bar{v} - \frac{\beta \sigma_v^2}{\beta^2 (\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2} (\alpha + \beta \bar{v}) \quad \lambda = \frac{\beta \sigma_v^2}{\beta^2 (\sigma_v^2 + \sigma_\epsilon^2) + \sigma_u^2}$$

Plugging these coefficients into *Informed*'s  $\alpha$  &  $\beta$  lets

us solve for his demand in terms of raw economic parameters.

# Homework 3 $\simeq$ Scharfstein & Stein (1990)

*Accounting Theory with Brett Trueman*



$$\text{Just confirming: } P(v_L|y_L) = \frac{P(y_L|v_L)P(v_L)}{P(y_L)} = \frac{p/2}{p/2+(1-p)/2} = \frac{p/2}{1/2} = p$$

Thus  $P(v_L|y_L) = P(y_L|v_L)$  and  $P(v_H|y_L) = P(y_L|v_H)$  and so on.

An equilibrium in which a CEO hires the auditor only upon seeing  $y_H$  means the market can infer that the CEO saw  $y_H$  if he hires the auditor, and saw  $y_L$  if he did not hire her. Define  $\hat{y}$  as the signal market infers given hiring decision such that  $\hat{y} = y_L$  if no audit,  $\hat{y} = y_H$  if auditor hired.

**Costly Signaling** →

Firm cashflow  $v \in \{v_L, v_H\}$  →

$Prob(v_L) = Prob(v_H) = 0.5$  →

CEO sees signal  $y \in \{y_L, y_H\}$  →

where  $Prob(y_i|v_i) = p$  for  $i \in \{L, H\}$  →

CEO cannot credibly disclose  $y$  →

CEO sells a fixed  $a\%$  of his shares →

He can choose to hire auditor for  $\$C$  →

who can provide report  $a \in \{a_L, a_H\}$  →

where  $Prob(a_i|v_i) = q$  for  $i \in \{L, H\}$  →

CEO maximizes stock issue proceeds →

**Question:** →

For what range of  $C$  will the →

equilibrium be such that a →

CEO that sees  $y_H$  hires auditor →

but one who sees  $y_L$  doesn't? →

**Conditional on not hiring auditor:**

$$E^{mkt}[v | \text{no audit}] = E^{mkt}[v | \hat{y} = y_L] = P(v_H|y_L)v_H + P(v_L|y_L)v_L$$

$$= (1 - p)v_H + pv_L = (1 - p)v_H + pv_L + (1 - p)v_L - (1 - p)v_L$$

$$\text{And so: } E^{mkt}[v | \text{no audit}] = E^{mkt}[v | \hat{y} = y_L] = v_L + (1 - p)(v_H - v_L)$$

**Conditional only on hiring auditor (before seeing audit result):**

$$E^{mkt}[v | \text{audit}] = E^{mkt}[v | \hat{y} = y_H] = P(v_H|y_H)v_H + P(v_L|y_H)v_L$$

$$E^{mkt}[v | \text{audit}] = pv_H + (1 - p)v_L = v_L + p(v_H - v_L)$$

**Conditional on audit (already knowing that  $\hat{y} = y_H$ )**

$$\text{and } a_H : E^{mkt}[v | a_H, y_H] = P(v_H|a_H, y_H)v_H + P(v_L|a_H, y_H)v_L$$

$$\text{and } a_L : E^{mkt}[v | a_L, y_H] = P(v_H|a_L, y_H)v_H + P(v_L|a_L, y_H)v_L$$

**Knowing that:**

$$E^{mkt}[v | \hat{y} = y_H, y_H] = P(a_H|y_H)E^{mkt}[v | a_H, y_H] + P(a_L|y_H)E^{mkt}[v | a_L, y_H]$$

$$= [P(v_H|a_H, y_H)P(a_H|y_H) + P(v_H|a_L, y_H)P(a_L|y_H)]v_H$$

$$+ [P(v_L|a_H, y_H)P(a_H|y_H) + P(v_L|a_L, y_H)P(a_L|y_H)]v_L$$

$$= \left[ P(v_H|a_H, y_H) \frac{P(a_H, y_H)}{P(y_H)} + P(v_H|a_L, y_H) \frac{P(a_L, y_H)}{P(y_H)} \right] v_H$$

$$+ \left[ P(v_L|a_H, y_H) \frac{P(a_H, y_H)}{P(y_H)} + P(v_L|a_L, y_H) \frac{P(a_L, y_H)}{P(y_H)} \right] v_L$$

$$= \left[ \frac{P(v_H, a_H, y_H)}{P(y_H)} + \frac{P(v_H, a_L, y_H)}{P(y_H)} \right] v_H + \left[ \frac{P(v_L, a_H, y_H)}{P(y_H)} + \frac{P(v_L, a_L, y_H)}{P(y_H)} \right] v_L$$

$$= \frac{P(v_H, y_H)}{P(y_H)} v_H + \frac{P(v_L, y_H)}{P(y_H)} v_L = P(v_H|y_H)v_H + P(v_L|y_H)v_L$$

$$E^{mkt}[v | \text{audit}, y_H] = E^{mkt}[v | \hat{y} = y_H, y_H] = pv_H + (1 - p)v_L = v_L + p(v_H - v_L)$$

$$aE^{mkt}[v \mid audit, \hat{y} = y_H, y_H] - C > aE^{mkt}[v \mid no\ audit, \hat{y} = y_L, y_H]$$

$$aE^{mkt}[v \mid \hat{y} = y_H, y_H] - C > aE^{mkt}[v \mid \hat{y} = y_L, y_H]$$

$$a[v_L + p(v_H - v_L)] - C > a[v_L + (1-p)(v_H - v_L)] \Leftrightarrow a(2p+1)(v_H - v_L) > C$$

Thus  $C$  has to be smaller than  $a(2p+1)(v_H - v_L)$

We now have the upper bound. Now to find the lower bound:

$$aE^{mkt}[v \mid audit, y_L] - C < aE^{mkt}[v \mid no\ audit, y_L]$$

$$aE^{mkt}[v \mid \hat{y} = y_H, y_L] - C < aE^{mkt}[v \mid \hat{y} = y_L, y_L]$$

To calculate this other part, we start with similar structure:

$$E[v|audit, y_L] = E[v|\hat{y} = y_H, y_L] = v_L + P(v_H|a, \hat{y} = y_H, y_L)(v_H - v_L)$$

$$P(v_H|a, \hat{y} = y_H, y_L) = P(a_H|y_L)P(v_H|a_H, y_H) + P(a_L|y_L)P(v_H|a_L, y_H)$$

$$P(v_H|a, \hat{y} = y_H, y_L) = [p(1-q)+(1-p)q]\frac{pq}{pq + (1-p)(1-q)} + [(1-p)(1-q)+pq]\frac{p(1-q)}{p(1-q) + (1-p)q}$$

$$\begin{aligned} & a \left( [p(1-q)+(1-p)q]\frac{pq}{pq + (1-p)(1-q)} + [(1-p)(1-q)+pq]\frac{p(1-q)}{p(1-q) + (1-p)q} \right) - C \\ & < aE^{mkt}[v \mid \hat{y} = y_L, y_H] = a[v_L + (1-p)(v_H - v_L)] \end{aligned}$$

Thus the lower bound is:

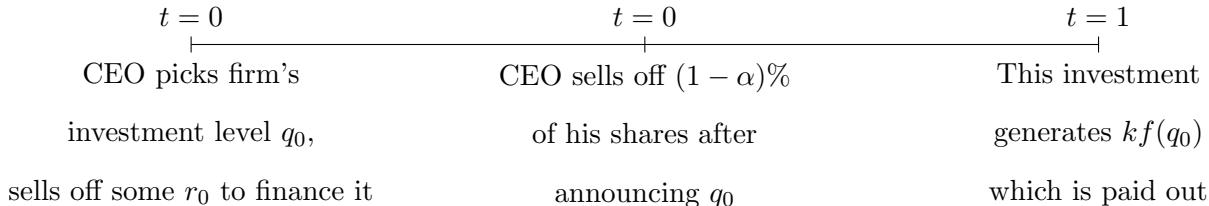
$$\begin{aligned} & a \left( [p(1-q)+(1-p)q]\frac{pq}{pq + (1-p)(1-q)} + [(1-p)(1-q)+pq]\frac{p(1-q)}{p(1-q) + (1-p)q} \right) \\ & - a[v_L + (1-p)(v_H - v_L)] < C \end{aligned}$$

and the upper bound is:

$$a(2p+1)(v_H - v_L) > C$$

# Homework 4 $\simeq$ Leland & Pyle (1977)

*Accounting Theory with Brett Trueman*



We know that in equilibrium, market infers  $\hat{k} = k(q_0) = k$  perfectly

**Solving CEO's problem:**  $E^{CEO}[U] = E^{CEO}[W_1] - \frac{\alpha}{2} var(W_1)$

$$E^{CEO}[W_1] = (1 - \alpha)E^{mkt}[kf(q_0)|q_0] + \alpha kf(q_0) + r_0 - q_0$$

Denote  $t=0$  investment level  $q_0 \rightarrow$

at  $t=0$  CEO picks firm's  $q_0 \rightarrow$

at  $t=1$ , outlay generates  $kf(q_0) \rightarrow$

Function  $f(q_0)$  is publicly known  $\rightarrow$

$$f'(q_0) > 0 \quad f''(q_0) < 0 \rightarrow$$

CEO perfectly knows  $k$  while market  $\rightarrow$

$$var^{CEO}(W_1) = var^{CEO}((1 - \alpha)E^{mkt}[kf(q_0)|q_0] + \alpha kf(q_0) + r_0 - q_0)$$

$$var^{CEO}(W_1) = var^{CEO}((1 - \alpha)f(q_0)E^{mkt}[k|q_0])$$

$$var^{CEO}(W_1) = (1 - \alpha)^2 f(q_0)^2 var^{CEO}(E^{mkt}[k|q_0])$$

**because CEO knows:**

$r_0 - q_0$ : His leftover riskfree asset holding after investing

$kf(q_0)$ : Outcome of production since he knows  $q_0$  and  $k$  perfectly

$\alpha$  &  $1 - \alpha$ : are exogenously fixed thus deterministic

Riskfree asset has zero return  $\rightarrow$

**Thus formally:**

$$\max_{q_0} E^{CEO}[W_1] - \frac{\alpha}{2} var(W_1)$$

$$= \max_{q_0} (1 - \alpha)f(q_0)E^{mkt}[k|q_0] + \alpha kf(q_0) + r_0 - q_0$$

$$- \frac{\alpha}{2} var(1 - \alpha)^2 f(q_0)^2 var^{CEO}(E^{mkt}[k|q_0])$$

CEO is **risk-averse**:  $\rightarrow$

FOC (conjectures are fulfilled in equilibrium so  $E^{mkt}[k|q_0] = k$ ):

$$0 = (1 - \alpha)f'(q_0)k + \alpha kf'(q_0) - 1 - \frac{\alpha}{2} var(1 - \alpha)^2 f(q_0)^2 var^{CEO}(k)$$

$$1 = (1 - \alpha)f'(q_0)k + \alpha kf'(q_0) \Leftrightarrow f'(q_0)k = 1$$

This is the differential equation that solves  $k$  as a function of  $q_0$ .

$E^{mkt}[kf(q_0)|q_0]$  : equals  $k$  since in this fully revealing

signaling equilibrium market, market perfectly infers CEO's

knowledge of  $k$  through his investment level  $q_0$ . The CEO,

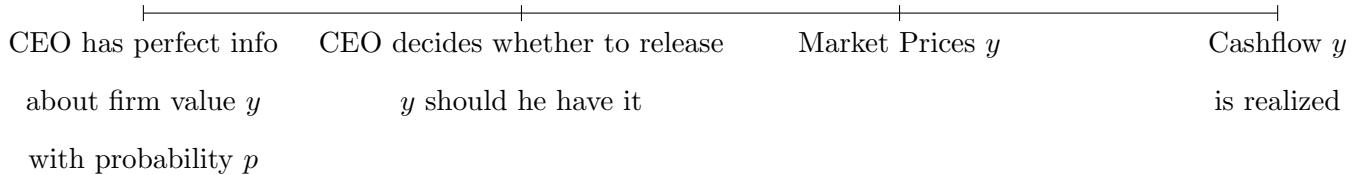
knowing this, can perfectly predict the market price the

shares will sell for, eliminating any risk or volatility.

This equilibrium would continue to exist if the CEO was risk neutral since his aversion played no role in the current equilibrium solution given that there was no riskiness in his wealth outcome.

## Homework 5: Dye (1985) and Jung & Kwon (1988)

*Accounting Theory with Brett Trueman*



Firm has risky cashflow  $y \sim Unif(0,t) \rightarrow$

With probability  $p$  CEO knows  $y$  perfectly  $\rightarrow$

Releasing  $y$  is costless if he has it  $\rightarrow$

He'll only do it if it increases price  $\rightarrow$

**Find cutoff  $\hat{y}$  above which he will release  $y$  and below which he will not.  $\rightarrow$**

**release  $y$  and below which he will not.  $\rightarrow$**

**At the cutoff, a CEO with signal  $\hat{y}$  will be indifferent between (D) disclosing  $\hat{y}$  and (ND) not disclosing it:**

$$E^{mkt}[y|\hat{y}, D] = E^{mkt}[y|\hat{y}, ND]$$

$$\hat{y} = P(\text{no info})E^{mkt}[y] + P(\text{info})P(\text{withholding})E^{mkt}[y|\text{withholding}]$$

$$\hat{y} = \frac{1-p}{(1-p)+pP(y<\hat{y})}\frac{t}{2} + \frac{pP(y<\hat{y})}{(1-p)+pP(y<\hat{y})}E^{mkt}[y|y < \hat{y}]$$

$$\hat{y} = \frac{1-p}{(1-p)+p\frac{\hat{y}}{t}}\frac{t}{2} + \frac{p\frac{\hat{y}}{t}}{(1-p)+p\frac{\hat{y}}{t}}\frac{\hat{y}}{2}$$

$$\hat{y}[(1-p) + p\frac{\hat{y}}{t}] = (1-p)\frac{t}{2} + p\frac{\hat{y}}{t}\frac{\hat{y}}{2}$$

$$\hat{y}[(1-p) + p\frac{\hat{y}}{t}] = (1-p)\frac{t}{2} + p\frac{\hat{y}^2}{2t}$$

$$2(1-p)\hat{y} + 2p\frac{\hat{y}^2}{t} = (1-p)t + p\frac{\hat{y}^2}{t}$$

$$2(1-p)\hat{y} + 2p\frac{\hat{y}^2}{t} - (1-p)t - p\frac{\hat{y}^2}{t} = 0$$

$$\frac{p}{t}\hat{y}^2 + 2(1-p)\hat{y} - (1-p)t = 0$$

$$\text{Using quadratic formula: } \hat{y} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

(keeping only positive term):

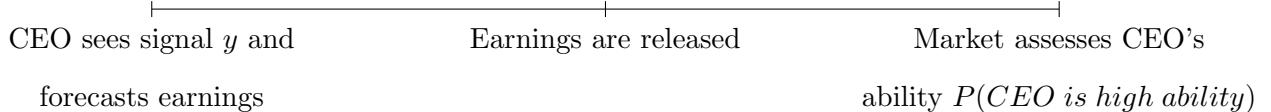
$$\hat{y} = \frac{-2(1-p) + \sqrt{4(1-p)^2 + 4\frac{p}{t}(1-p)t}}{2\frac{p}{t}}$$

$$\hat{y} = \frac{-2(1-p) + 2\sqrt{1-2p+p^2+(p-p^2)}}{2\frac{p}{t}}$$

$$\hat{y} = \frac{t}{p} \left[ -(1-p) + \sqrt{1-p} \right]$$

## Homework 6: Trueman (1994)

*Accounting Theory with Brett Trueman*



**Denote**  $y_L = 1 ; y_H = 2 ; f_L = 1 ; f_H = 2 ; x_L = 1 ; x_H = 2$

**Denote**  $W(f, x)$  the compensation a CEO can expect by investors after seeing his forecast  $f$  and the realization of earnings  $x$ , such that  $W(f, x) = C \cdot P(CEO_H|f, x) - \mathbf{1}_{f \neq x} K$

- Earnings  $x$  can be high (2) or low (1) →
- Ex-ante probability  $P(x = 1) = 0.7$  →
- Ex-ante probability  $P(x = 2) = 0.3$  →
- CEO gets signal  $y$  about level of  $x$  →
- CEO can be high ability ( $CEO_H$ ) →  
or low ability ( $CEO_L$ ) →
- Ex-ante  $P(CEO_H) = P(CEO_L) = \frac{1}{2}$  →
- $P(y = 2|x = 2, CEO_H) = g$  →
- $P(y = 1|x = 2, CEO_H) = 1 - g$  →
- $P(y = 2|x = 2, CEO_L) = b$  →
- $P(y = 1|x = 2, CEO_L) = 1 - b$  →
- CEO forecasts level of  $x$ :  $f \in \{1, 2\}$  →
- CEO receives wage equal to →  
 $C * P^{mkt}(CEO_H|f, x) - \mathbf{1}_{x \neq f} K$  →
- If realized  $x \neq f$ , he gets  $-\$K$  penalty →
- Market conjectures: →
- $CEO_H$  always reports his  $y$  truthfully →
- $CEO_L$  reports a  $y = 1$  truthfully →
- But  $CEO_L$  will lie at a rate  $\alpha$  if  $y = 2$  →  
ie:  $\alpha\%$  of the time  $CEO_L$  →  
reports  $f = 1$  after seeing  $y = 2$  →
- Find the equilibrium level of  $\alpha$**  →

We know  $CEO_H$  will always give his private signal as his forecast.

We know  $CEO_L$  will as well if his signal is  $y = 1$ .

$CEO_L$  needs to optimize his lying rate  $\alpha$  for when he sees  $y = 2$ .

Note:  $P(CEO_L) = P(CEO_H) = \frac{1}{2}$  so they're omitted in calculations below.

**$CEO_L$ 's expected pay when seeing  $y_H$  and lying ( $f = f_L \neq y_H$ ):**

$$E[W(f_L, x)|y_H] = P(x_H|y_H)W(f_L, x_H) + P(x_L|y_H)W(f_L, x_L)$$

$$= P(x_H|y_H)[CP(CEO_H|f_L, x_H) - K] + P(x_L|y_H)[CP(CEO_H|f_L, x_L)]$$

$$E[W(f_L, x)|y_H] = P(x_H|y_H)\left[\frac{C \cdot (1-g)}{(1-g)+b\alpha+(1-b)} - K\right] + P(x_L|y_H)\left[\frac{C \cdot g}{g+b+(1-b)\alpha}\right]$$

**$CEO_L$ 's expected pay when seeing  $y_H$  & not lying ( $f = f_H = y_H$ ):**

$$E[W(f_H, x)|y_H] = P(x_H|y_H)W(f_H, x_H) + P(x_L|y_H)W(f_H, x_L)$$

$$= P(x_H|y_H)[CP(CEO_H|f_H, x_H)] + P(x_L|y_H)[CP(CEO_H|f_H, x_L) - K]$$

$$E[W(f_H, x)|y_H] = P(x_H|y_H)\left[\frac{C \cdot g}{g+b(1-\alpha)}\right] + P(x_L|y_H)\left[\frac{C \cdot (1-g)}{(1-g)+(1-b)(1-\alpha)} - K\right]$$

Get  $P(x_H|y_H)$  &  $P(x_L|y_H)$  using Bayes' Theorem ( $CEO_L$ 's perspective):

$$P(x_H|y_H) = \frac{P(y_H|x_H)P(x_H)}{P(y_H)} = \frac{b(0.3)}{b(0.3)+(1-b)(0.7)}$$

$$P(x_L|y_H) = \frac{P(y_H|x_L)P(x_L)}{P(y_H)} = \frac{(1-b)(0.7)}{b(0.3)+(1-b)(0.7)}$$

**At equilibrium**  $\alpha^*$  will be such that  $CEO_L$  will be indifferent between

the above two, thus equating  $E[W(f_L, x)|y_H] = E[W(f_H, x)|y_H] \Leftrightarrow$

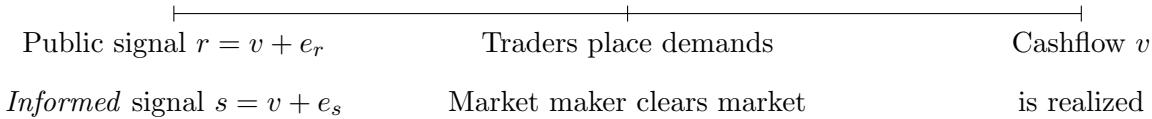
$$\frac{b(0.3)}{b(0.3)+(1-b)(0.7)}\left[\frac{C \cdot (1-g)}{(1-g)+b\alpha+(1-b)} - K\right] + \frac{(1-b)(0.7)}{b(0.3)+(1-b)(0.7)}\left[\frac{C \cdot g}{g+b+(1-b)\alpha}\right] =$$

$$\frac{b(0.3)}{b(0.3)+(1-b)(0.7)}\left[\frac{C \cdot g}{g+b(1-\alpha)}\right] + \frac{(1-b)(0.7)}{b(0.3)+(1-b)(0.7)}\left[\frac{C \cdot (1-g)}{(1-g)+(1-b)(1-\alpha)} - K\right]$$

will give an expression for  $\alpha^*$  in equilibrium.

# Assignment 1: Problem 1: Grossman & Stiglitz (1980)

Accounting Theory with Judson Caskey



## 1) Starting with the *Informed* trader's optimization problem:

$$\max_{q_I} q_I(E^I[v|r,s] - p) - \frac{1}{2\rho} q_I^2 \text{var}(v|s,r) \Rightarrow \text{FOC: } q_I = \frac{E^I[v|r,s] - p}{\frac{1}{\rho} \cdot \text{var}(v|s,r)}$$

Note:  $\text{var}(v|r) = \tau_v^{-1} - \frac{\tau_v^{-2}}{\tau_v^{-1} + \tau_r^{-1}} = \frac{\tau_v^{-1} \tau_r^{-1}}{\tau_v^{-1} + \tau_r^{-1}} = \frac{1}{\frac{\tau_v^{-1}}{\tau_v^{-1} + \tau_r^{-1}} + \frac{\tau_r^{-1}}{\tau_v^{-1} + \tau_r^{-1}}} = \frac{1}{\tau_v + \tau_r}$

And similarly  $\text{var}(v|s,r) = \frac{1}{\tau_v + \tau_r + \tau_s}$

Whereas we know that  $E^I[v|r,s] = \frac{\tau_v \mu_v + \tau_r r + \tau_s s}{\tau_v + \tau_r + \tau_s}$

$$\text{Thus: } q_I = \rho(\tau_v + \tau_r + \tau_s)(\frac{\tau_v \mu_v + \tau_r r + \tau_s s}{\tau_v + \tau_r + \tau_s} - p)$$

Model is based on →

Grossman & Stiglitz (1980) →

Continuum of risk averse traders →

With risk tolerance coefficient  $\rho$  →

Thus CARA coefficient  $1/\rho$  →

Prior of payoff:  $v \sim N(\mu_v, \tau_v^{-1})$  →

All investors see  $r = v + e_r$  →

$\lambda\%$  of traders are informed so →

they see private signal  $s = v + e_s$  →

$e_r \sim N(0, \tau_r^{-1}) \perp e_s \sim N(0, \tau_s^{-1})$  →

Noisy supply  $x \sim N(\mu_x, \sigma_x^2)$  →

Prices determined by mkt clearing →

Assume linear strategies all around →

*Uninformed's Conjectures:* →

$$p = p_0 + p_r r + p_s s - p_x x \rightarrow$$

$\hat{s} = s + k \times (x - \mu_x)$  & constant  $k$  →

But his expected value  $E[\hat{s}] = E[\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s} r + \frac{p_x}{p_s} x] = \frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s} r + \frac{p_x}{p_s} \mu_x$

This is because he knows  $\tau_s$  but not the realization of  $s$ . Also note:  $k = \frac{p_x}{p_s}$ .

Question: →

1) What is the equilibrium price? →

2) Compute ex-ante utility if  $\lambda = 0$  →

3) How does it vary with  $\tau_r$ ? →

$$E[v|r, \hat{s}] = \frac{\tau_v \mu_v + \tau_r r + \hat{s}_s (\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s} r + \frac{p_x}{p_s} \mu_x)}{\tau_v + \tau_r + \hat{s}_s}$$

$$\text{and } \text{var}(v|\hat{s}) = \tau_v^{-1} - \frac{\tau_v^{-2}}{\tau_v^{-1} + \tau_s^{-1} + k^2 \sigma_x^2} = \frac{\tau_v^{-1} \hat{s}_s^{-1}}{\tau_v^{-1} + \hat{s}_s^{-1}} = \frac{1}{\tau_v + \hat{s}_s}$$

similarly:  $\text{var}(v|r, \hat{s}) = \frac{1}{\tau_v + \tau_r + \hat{s}_s}$  where  $\hat{s}_s = \frac{\tau_s}{1 + (\frac{p_x}{p_s})^2 \tau_s \sigma_x^2}$

$$\text{Thus: } q_U = \frac{E^U[v|r, \hat{s}] - p}{\frac{1}{\rho} \cdot \text{var}(v|r, \hat{s})} = \rho(\tau_v + \tau_r + \hat{s}_s)(\frac{\tau_v \mu_v + \tau_r r + \hat{s}_s (\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s} r + \frac{p_x}{p_s} \mu_x)}{\tau_v + \tau_r + \hat{s}_s} - p)$$

We can now calculate aggregate demand using  $\lambda q_I + (1 - \lambda) q_U$

$$= \lambda \rho (\tau_v \mu_v + \tau_r r + \tau_s s - p(\tau_v + \tau_r + \tau_s)) +$$

$$(1 - \lambda) \rho \left( \tau_v \mu_v + \tau_r r + \hat{s}_s (\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s} r + \frac{p_x}{p_s} \mu_x) - p(\tau_v + \tau_r + \hat{s}_s) \right)$$

continued on next page

2

**Now Plugging into Supply = Demand:**  $x = \lambda q_I + (1 - \lambda)q_U$

$$x = \lambda\rho(\tau_v\mu_v + \tau_r r + \tau_s s - p(\tau_v + \tau_r + \tau_s)) + (1 - \lambda)\rho(\tau_v\mu_v + \tau_r r + \hat{\tau}_s(\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s}r + \frac{p_x}{p_s}\mu_x) - p(\tau_v + \tau_r + \hat{\tau}_s))$$

$$\frac{x}{\rho} = \tau_v\mu_v + \tau_r r + \lambda\tau_s s - p(\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s) + (1 - \lambda)\hat{\tau}_s(\frac{p}{p_s} - \frac{p_0}{p_s} - \frac{p_r}{p_s}r + \frac{p_x}{p_s}\mu_x)$$

$$p(\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s) = -\frac{x}{\rho} + \left[ \tau_v\mu_v + (1 - \lambda)\hat{\tau}_s(\frac{p_x}{p_s}\mu_x - \frac{p_0}{p_s}) \right] + \tau_r r + \lambda\tau_s s + (1 - \lambda)(\hat{\tau}_s \frac{p}{p_s} - \hat{\tau}_s \frac{p_r}{p_s}r)$$

$$p(\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s) - (1 - \lambda)\hat{\tau}_s \frac{p}{p_s} = -\frac{1}{\rho}x + \left[ \tau_v\mu_v + (1 - \lambda)\hat{\tau}_s(\frac{p_x}{p_s}\mu_x - \frac{p_0}{p_s}) \right] + \left[ \tau_r - (1 - \lambda)\hat{\tau}_s \frac{p_r}{p_s} \right] r + \lambda\tau_s s$$

$$p \left( \tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(1 - \frac{1}{p_s}) \right) = \left[ \tau_v\mu_v + (1 - \lambda)\hat{\tau}_s(\frac{p_x\mu_x - p_0}{p_s}) \right] + \left[ \tau_r - (1 - \lambda)\hat{\tau}_s \frac{p_r}{p_s} \right] r + \lambda\tau_s s - \frac{1}{\rho}x$$

$$p = \underbrace{\frac{\tau_v\mu_v + (1 - \lambda)\hat{\tau}_s(\frac{p_x\mu_x - p_0}{p_s})}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(\frac{p_s-1}{p_s})}}_{p_0} + \underbrace{\frac{\tau_r - (1 - \lambda)\hat{\tau}_s \frac{p_r}{p_s}}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(\frac{p_s-1}{p_s})}}_r r$$

$$+ \underbrace{\frac{\lambda\tau_s}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(\frac{p_s-1}{p_s})}}_{p_s} s - \underbrace{\frac{\frac{1}{\rho}}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(\frac{p_s-1}{p_s})}}_{p_x} x$$

Thus  $\frac{p_x}{p_s} = \frac{1}{\rho\lambda\tau_s}$  so now we know that  $\hat{\tau}_s = \frac{\tau_s}{1 + (\frac{p_x}{p_s})^2\tau_s\sigma_x^2} = \frac{\tau_s}{1 + (\rho\lambda\tau_s)^{-2}\tau_s\sigma_x^2} = \frac{\tau_s}{1 + \frac{\sigma_x^2}{\rho^2\lambda^2\tau_s}}$

$$p_s = \frac{\lambda\tau_s}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(\frac{p_s-1}{p_s})} \Leftrightarrow \left( \tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s(1 - \frac{1}{p_s}) \right) p_s = \lambda\tau_s$$

$$\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s - \frac{(1 - \lambda)\hat{\tau}_s}{p_s} = \frac{\lambda\tau_s}{p_s} \Leftrightarrow \tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s = \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{p_s}$$

$$\Rightarrow p_s = \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s} \Rightarrow p_x = \frac{p_x}{p_s} p_s = \frac{p_s}{\rho\lambda\tau_s} = \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{(\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s)\rho\lambda\tau_s}$$

$$\Rightarrow \frac{p_s - 1}{p_s} = 1 - \frac{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s} = \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s - \tau_v - \tau_r - \lambda\tau_s - (1 - \lambda)\hat{\tau}_s}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s} = \frac{-\tau_v - \tau_r}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}$$

$$\frac{p_r}{p_s} = \frac{\tau_r - (1 - \lambda)\hat{\tau}_s \frac{p_r}{p_s}}{\lambda\tau_s} \Leftrightarrow \frac{p_r}{p_s} \lambda\tau_s = \tau_r - (1 - \lambda)\hat{\tau}_s \frac{p_r}{p_s} \Leftrightarrow \frac{p_r}{p_s} [\lambda\tau_s + (1 - \lambda)\hat{\tau}_s] = \tau_r \Leftrightarrow \frac{p_r}{p_s} = \frac{\tau_r}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}$$

$$p_r = \frac{\tau_r}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s} p_s = \frac{\tau_r}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s} \times \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s} = \frac{\tau_r}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s}$$

$$\frac{p_0}{p_s} = \frac{\tau_v\mu_v + (1 - \lambda)\hat{\tau}_s(\frac{p_x\mu_x - p_0}{p_s})}{\lambda\tau_s} \Leftrightarrow \frac{p_0}{p_s} \lambda\tau_s = \tau_v\mu_v + (1 - \lambda)\hat{\tau}_s \left( \frac{p_x\mu_x}{p_s} - \frac{p_0}{p_s} \right) = \tau_v\mu_v + (1 - \lambda)\hat{\tau}_s \left( \frac{\mu_x}{\rho\lambda\tau_s} - \frac{p_0}{p_s} \right)$$

$$\Leftrightarrow \frac{p_0}{p_s} (\lambda\tau_s + (1 - \lambda)\hat{\tau}_s) = \tau_v\mu_v + \frac{(1 - \lambda)\hat{\tau}_s\mu_x}{\rho\lambda\tau_s} \Leftrightarrow \frac{p_0}{p_s} = \frac{\tau_v\mu_v + \frac{(1 - \lambda)\hat{\tau}_s\mu_x}{\rho\lambda\tau_s}}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}$$

$$\Rightarrow p_0 = \frac{p_0}{p_s} p_s = \frac{\tau_v\mu_v + \frac{(1 - \lambda)\hat{\tau}_s\mu_x}{\rho\lambda\tau_s}}{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s} \frac{\lambda\tau_s + (1 - \lambda)\hat{\tau}_s}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s} = \frac{\tau_v\mu_v + \frac{(1 - \lambda)\hat{\tau}_s\mu_x}{\rho\lambda\tau_s}}{\tau_v + \tau_r + \lambda\tau_s + (1 - \lambda)\hat{\tau}_s}$$

Thus we now have the equilibrium price in terms of raw parameters:  $p = p_0 + p_r r + p_s s - p_x x$

2) [incorrect] **Compute ex-ante utility if  $\lambda = 0$**  (the answer on this page is ‘wrong’, he meant ex-ante before seeing  $r$ , but here’s the ‘ex-ante’ after seeing  $r$  but before trading actually takes place and  $x \in P$  are realized) : this would mean that all traders are uninformed and do not see signal  $s$  nor can they infer it from the price signal (ie: no  $\hat{s}$ ). The uninformed trader’s conjecture then becomes:  $p = p_0 + p_r r - p_x x$

$$\max_{q_U} E^U[U^U] = \max_{q_U} q_U(E^U[v|r] - p) - \frac{1}{2\rho} q_U^2 \text{var}(v|r) \Rightarrow \text{FOC: } q_U = \rho \frac{E^U[v|r] - p}{\text{var}(v|r)}$$

$$E^U[v|r] = E^U[v] + \frac{\text{cov}(v, r)}{\text{var}(r)}(r - E[r]) = \mu_v + \frac{\tau_v^{-1}}{\tau_v^{-1} + \tau_r^{-1}}(r - \mu_v) = \frac{\tau_v \mu_v + \tau_r r}{\tau_v + \tau_r} \leftarrow \text{or just begin with this}$$

$$\text{var}(v|r) = \tau_v^{-1} - \frac{\tau_v^{-2}}{\tau_v^{-1} + \tau_r^{-1}} = \frac{\tau_v^{-1} \tau_r^{-1}}{\tau_v^{-1} + \tau_r^{-1}} = \frac{1}{\frac{\tau_v^{-1}}{\tau_v^{-1} + \tau_r^{-1}} + \frac{\tau_r^{-1}}{\tau_v^{-1} + \tau_r^{-1}}} = \frac{1}{\tau_v + \tau_r}$$

$$q_U = \rho \frac{E^U[v|r] - p}{\text{var}(v|r)} = \rho(\tau_v + \tau_r) \left( \frac{\tau_v \mu_v + \tau_r r}{\tau_v + \tau_r} - p \right) = \rho(\tau_v \mu_v + \tau_r r - p(\tau_v + \tau_r))$$

In equilibrium: **Demand = Supply**

$$q_U = x \Leftrightarrow \tau_v \mu_v + \tau_r r - p(\tau_v + \tau_r) = x/\rho \Leftrightarrow p(\tau_v + \tau_r) = \tau_v \mu_v + \tau_r r - x/\rho$$

$$p = \frac{\tau_v \mu_v + \tau_r r - \frac{x}{\rho}}{\tau_v + \tau_r} = \underbrace{\frac{\tau_v \mu_v}{\tau_v + \tau_r}}_{p_0} + \underbrace{\frac{\tau_r}{\tau_v + \tau_r} r}_{p_r} - \underbrace{\frac{1/\rho}{\tau_v + \tau_r} x}_{p_x}$$

$$q_U = \rho(\tau_v + \tau_r) \left( \frac{\tau_v \mu_v + \tau_r r}{\tau_v + \tau_r} - \left[ \underbrace{\frac{\tau_v \mu_v}{\tau_v + \tau_r} + \frac{\tau_r}{\tau_v + \tau_r} r}_{p=p_0+p_r r + p_s s} - \underbrace{\frac{1/\rho}{\tau_v + \tau_r} x}_{p_x} \right] \right) = \rho(\tau_v \mu_v + \tau_r r - \tau_v - \tau_r r + \frac{x}{\rho})$$

$$q_U = \rho(\tau_v \mu_v - \tau_v) + x = \rho \tau_v (\mu_v - 1) + x$$

$$E^U[U^U] = q_U(E^U[v|r] - p) - \frac{1}{2\rho} q_U^2 \text{var}(v|r)$$

2) [correct] **Compute ex-ante utility if  $\lambda = 0$** :

**Notes beforehand:** Because competitive traders in a CARA/Normal setting with risk **aversion**  $\gamma$  have ex-ante quantities  $q = \frac{E[V]-p}{\gamma \text{var}(V)}$  their certainty equivalent of buying  $q$  units is:

**C.E. of  $q$  units:**  $q(E[V]-p) - \frac{\gamma}{2} \text{var}(V)q^2 \Rightarrow$  from FOC:  $0 = E[V]-p-\gamma \text{var}(V)q \Leftrightarrow p = E[V]-\gamma \text{var}(V)q$

**Plugging this  $p$  into C.E.:**  $\Rightarrow q(E[V]-p) - \frac{\gamma}{2} \text{var}(V)q^2 = q(E[V]-(E[V]-\gamma \text{var}(V)q)) - \frac{\gamma}{2} \text{var}(V)q^2$   
 $= \gamma \text{var}(V)q^2 - \frac{\gamma}{2} \text{var}(V)q^2 = \frac{\gamma}{2} \text{var}(V)q^2 \rightarrow$  this is the certainty equivalent of  $q$  risky units

In this setting we have a risk **tolerance** parameter so  $\gamma = \rho^{-1}$

#### ALSO NOTE:

Here ex-ante price is uncertain so add it to the mean and variance operators as well:  $E[V] = E[v-p]$  &  $\text{var}(V) = \text{var}(v-p)$

$$\frac{\gamma}{2} \text{var}(V)q^2 = \frac{1}{2\rho} \text{var}(v-p) = q^2 \frac{1}{2\rho} \text{var}(v-p) \left( \frac{\rho E[v-p]}{\text{var}(v-p)} \right)^2 = \frac{\rho}{2} \frac{E[v-p]^2}{\text{var}(v-p)}$$

However this does not take into consideration the fact that while price is unknown *ex-ante*, it will be known at the time of trade. In calculating the *ex-ante* expectation at time 0, an adjustment has to be made for the fact that the trader will know certain information (signal  $r$  here) which will lower his risk level when trading at time 1:

**Using the law of total variance:**  $\text{var}_0(V-P) = \text{var}(V-P) + \text{var}_0(E[V-P])$

In this setting it would be:  $\text{var}(v-p) = \text{var}(v-p|r) + \text{var}(E[v-p|r])$

**Note:**  $\text{var}(v-p|r)$  is the expectation once  $r$  is received and  $p$  is then also in the information set at that point (deterministic), so in this setting  $\text{var}(v-p|r) = \text{var}(v|r) = \frac{1}{\tau_v \tau_r}$ .

$$E[u(c)] = -e^{-\gamma \mu_c + \frac{1}{2} \sigma_c^2 \gamma^2} = -e^{-\gamma(\mu_c - \frac{\gamma}{2} \sigma_c^2)} = u\left(\mu_c - \frac{\gamma}{2} \sigma_c^2\right)$$

$$\text{We set } \gamma = \frac{1}{\rho} \quad \& \quad X = \frac{E[v-p]}{\sqrt{\frac{2}{\rho} \text{var}(v-p)}} \quad \Rightarrow \quad E[e^{-\gamma X^2}] = E\left[\exp\left\{-\frac{1}{\rho} \frac{E[v-p]^2}{2 \text{var}(v-p)}\right\}\right]$$

Then we would get the following ex-ante certainty equivalent payment where the 0 subscripts denote moments with respect to ex ante information:

$$\begin{aligned} \frac{1}{2\gamma} \frac{E_0[V-P]^2}{\text{var}_0(V-P)} + \frac{1}{\gamma} \log \sqrt{\frac{\text{var}_0(V-P)}{\text{var}(V-P)}} &= \frac{\rho}{2} \frac{E[v-p]^2}{\text{var}(v-p)} + \rho \log \sqrt{\frac{\text{var}(v-p)}{\text{var}(v-p|r)}} \\ &= \frac{\rho}{2} \frac{E[v-p]^2}{\text{var}(v-p)} + \rho \log \sqrt{\frac{\text{var}(v-p)}{\frac{1}{\tau_v \tau_r}}} \end{aligned}$$

$$E[v - p] = \mu_v - \frac{\tau_v \mu_v + \tau_r E[r] - \frac{E[x]}{\rho}}{\tau_v + \tau_r} = \frac{\tau_v \mu_v + \tau_r \mu_v - \tau_v \mu_v - \tau_r \mu_v + \frac{\mu_x}{\rho}}{\tau_v + \tau_r} = \frac{1}{\rho} \frac{1}{\tau_v + \tau_r} \mu_x$$

$$\text{var}[v - p] = \text{var} \left( v - \frac{\tau_v \mu_v + \tau_r r - \frac{x}{\rho}}{\tau_v + \tau_r} \right) = \text{var} \left( \frac{v(\tau_v + \tau_r) - \tau_v \mu_v - \tau_r(v + e_r) + \frac{x}{\rho}}{\tau_v + \tau_r} \right)$$

$$\text{var}[v - p] = \frac{\text{var} \left( v\tau_v - \tau_r e_r + \frac{x}{\rho} \right)}{(\tau_v + \tau_r)^2} = \frac{\frac{\tau_v^2}{\tau_v} + \frac{\tau_r^2}{\tau_r} + \frac{\sigma_x^2}{\rho^2}}{(\tau_v + \tau_r)^2} = \frac{\tau_v + \tau_r + \frac{\sigma_x^2}{\rho^2}}{(\tau_v + \tau_r)^2}$$

Then:

$$\frac{\rho}{2} \frac{\text{E}[v - p]^2}{\text{var}(v - p)} + \rho \log \sqrt{\frac{\text{var}(v - p)}{\frac{1}{\tau_v + \tau_r}}} = \frac{\rho}{2} \frac{\left[ \frac{1}{\rho} \frac{1}{\tau_v + \tau_r} \mu_x \right]^2}{\frac{\tau_v + \tau_r + \frac{\sigma_x^2}{\rho^2}}{(\tau_v + \tau_r)^2}} + \rho \log \sqrt{\frac{\tau_v + \tau_r + \sigma_x^2/\rho^2}{(\tau_v + \tau_r)^2} (\tau_v + \tau_r)}$$

$$E_0[U] = \frac{1}{2\rho} \frac{\mu_x^2}{\tau_v + \tau_r + \frac{\sigma_x^2}{\rho^2}} + \rho \log \sqrt{1 + \frac{\sigma_x^2/\rho^2}{\tau_v + \tau_r}}$$

- 3) Ex-ante utility falls as the precision of public information increases ( $\tau_r \uparrow$ ). Investors earn profit from risk premium, and information lowers that.

# Assignment 1: Problem 4

*Accounting Theory with Judson Caskey*

*Based on material seen in Accounting Theory (Prof. Caskey Spring 2020)*

**Grossman & Stiglitz (1980) →**

and **Guttman (2010)** elements →

Investors trade over  $t \in [0, T]$  →

They have CARA coefficient  $\gamma$  →

Asset payoff is  $\pi \sim \mathcal{N}(\mu_{\pi 0}, \sigma_{\pi 0}^2)$  →

Public information at time  $t$  has precision: →

$f(t)$  with  $f(0) = \frac{1}{\sigma_{\pi 0}^2} \geq 1$  (i.e.  $\sigma_{\pi 0}^2 \leq 1$ ) →

$f(t) \rightarrow \infty$  as  $t \rightarrow T$  and  $f'(t) > 0$  →

Investors can buy signal  $\psi_i(t) = \pi + e_i(t)$  →

from **analyst**  $i$  where  $\text{var}(e_i(t)) = \frac{1}{f_i(t)}$  →

Imputing the cost of information  $c$  from →

the informed/uninformed indifference†: →

† from Grossman & Stiglitz (1985) →

$$c = \frac{1}{\gamma} \log \sqrt{\frac{\text{var}_u(v-p)}{\text{var}_i(v-p)}} \rightarrow$$

$$\text{where } \text{var}_u(\pi - p) = \frac{1}{f(t)} \rightarrow$$

$$\text{and } \text{var}_i(\pi - p) = \frac{1}{f(t) + f_i(t)} \text{ and so:} \rightarrow$$

$$U^i(f(t)) = C(t) = \frac{1}{\gamma} \log \sqrt{\frac{\text{var}_u(v-p)}{\text{var}_{(v-p)}}} = \rightarrow$$

$$\frac{1}{2\gamma} \log \frac{f(t) + f_i(t)}{f(t)} \Leftarrow \text{analyst } i \text{ maximizes this} \rightarrow$$

$$\text{By assumption } f_i(t) = F_i + \alpha_i \log f(t) \rightarrow$$

$$\text{where } \alpha_i \text{ is speed of analyst learning} \rightarrow$$

$$\text{and } F_i \text{ is his initial info advantage} \rightarrow$$

$$\text{for some constants } F_i \geq 0 \text{ and } \alpha_i > 0 \rightarrow$$

$$\text{Both } f(t) \text{ & } f_i(t) \text{ increase over time} \rightarrow$$

$$f'_i(t) = \frac{\alpha_i}{f(t)} f'(t): \text{analyst precision grows} \rightarrow$$

$$\text{faster than public precision if } f(t) < \alpha_i \rightarrow$$

**Question's additional assumptions:** →

$$\cdot \text{Noisy supply } x \sim \mathcal{N}(\mu_x, \sigma_x^2) \rightarrow$$

$$\cdot \lambda\% \text{ of investors receive the forecast} \rightarrow$$

$$\text{All investors know the forecast timing} \rightarrow$$

$$\text{parameter } F_{0i} < \alpha_i \text{ and slope } \alpha_i < e \text{ in } f_i \rightarrow$$

**Derive optimal forecast timing** →

From *uninformed* in Problem 1 we had that:

$$\text{var}(v | \hat{s}) = \tau_v^{-1} - \frac{\tau_v^{-2}}{\tau_v^{-1} + \tau_s^{-1} + k^2 \sigma_x^2} = \frac{\tau_v^{-1} \hat{\tau}_s^{-1}}{\tau_v^{-1} + \hat{\tau}_s^{-1}} = \frac{1}{\tau_v + \hat{\tau}_s}$$

$$\text{and } \hat{\tau}_s = \frac{\tau_s}{1 + (\frac{p_x}{p_s})^2 \tau_s \sigma_x^2} = \frac{\tau_s}{1 + (\rho \lambda \tau_s)^{-2} \tau_s \sigma_x^2} = \frac{\tau_s}{1 + \frac{\sigma_x^2}{\rho^2 \lambda^2 \tau_s}}$$

For *uninformed* in this setting simply replace

$$\tau_s = f_i(t), \hat{\tau}_s = \hat{f}_i(t) \text{ and } \tau_v = f(t) :$$

$$\text{var}(\pi | \hat{f}_i(t)) = \frac{1}{f(t) + \hat{f}_i(t)} = \left( f(t) + \frac{f_i(t)}{1 + \frac{\gamma^2 \sigma_x^2}{\lambda^2 f_i(t)}} \right)^{-1}$$

$$\text{Now precision: } \text{var}(\pi | \hat{f}_i(t))^{-1} = f(t) + \frac{f_i(t)}{1 + \frac{\gamma^2 \sigma_x^2}{\lambda^2 f_i(t)}}$$

$$= f(t) + \frac{f_i(t)}{\frac{\lambda^2 f_i(t) + \gamma^2 \sigma_x^2}{\lambda^2 f_i(t)}} = f(t) + \frac{f_i(t)^2 \lambda^2}{\rho^2 \lambda^2 f_i(t) + \gamma^2 \sigma_x^2} = f(t) + \frac{\lambda^2}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i(t)$$

Then plug this into analyst's optimization problem

$$\begin{aligned} \max_f \frac{1}{2\gamma} \log \frac{\text{var}^U(\pi | \hat{f}_i(t))}{\text{var}^I(\pi | f_i(t))} &= \max_f \frac{1}{2\gamma} \log \frac{\text{var}^I(\pi | f_i(t))^{-1}}{\text{var}^U(\pi | \hat{f}_i(t))^{-1}} \\ &= \max_f \frac{1}{2\gamma} \log \frac{f(t) + f_i(t)}{f(t) + \frac{\lambda^2}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i(t)} \end{aligned}$$

Analyst revenue is maximized when

the term inside the log is maximized:

$$\begin{aligned} 0 &= \frac{d}{df} \left( \frac{f + f_i}{f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i} \right) = \frac{\partial}{\partial f} \left( \frac{f + f_i}{f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i} \right) + \frac{\partial}{\partial f_i} \left( \frac{f + f_i}{f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i} \right) \frac{df_i}{df} \\ 0 &= -\frac{\frac{\gamma^2 \sigma_x^2}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i}}}{{\left( f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i \right)}^2} + \frac{\frac{\gamma^2 \sigma_x^2}{f_i}}{{\left( f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i \right)}^2} \frac{\frac{\gamma^2 \sigma_x^2}{f_i} f - \lambda^2 f_i}{\left( \lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i} \right)^2} \frac{\alpha_i}{f} \\ 0 &= \frac{\frac{\gamma^4 \sigma_x^2 \frac{\sigma_x^2}{f_i}}{f_i}}{{\left( f + \frac{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}}{\lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i(t)}} f_i \right)}^2} \frac{\frac{\alpha_i}{f_i} - 1 - \lambda^2 \frac{f_i}{\gamma^2 \sigma_x^2} \left( \frac{\alpha_i}{f} + 1 \right)}{\left( \lambda^2 + \gamma^2 \frac{\sigma_x^2}{f_i} \right)^2} \end{aligned}$$

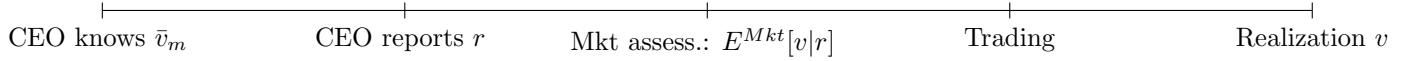
Note that putting  $\lambda = 0$  implies that the expression is maximized at

$$\frac{\alpha_i}{f_i} = 1, \text{ giving } f(t^*) = e^{1-F_0/\alpha_i} \text{ as in Guttman (2010).}$$

## Assignment 2: Problem 3: Part 1

Stein (1989): Signal Jamming

*Accounting Theory with Judson Caskey*



CEO is risk **neutral** →

Investors believe  $\tilde{v} \sim N(\mu_v, \sigma_v^2)$  →

CEO privately believes  $\tilde{v} \sim N(\bar{v}_m, (1 - \delta)\sigma_v^2)$  →

where  $cov(v, \bar{v}_m) = var(\bar{v}_m) = \delta\sigma_v^2$  →

CEO reports  $r$  to investors to optimize →

$\max_r bE^{Mkt}[v|r] - \frac{c}{2}(r - \bar{v}_m)^2$  →

$b$  &  $c$  are publicly known parameters →

Mkt assess:  $P(r) = E^{Mkt}[v | r]$  →

CEO choice of report is  $r(\bar{v}_m)$  →

Mkt's conjecture of CEO's reporting →

strategy:  $\hat{r}(\bar{v}_m) = r_0 + r_v \bar{v}_m$  →

$\hat{p}$ : CEO's conject. of mkt price assessment →

has linear form  $\hat{E}^{Mkt}[v|r,v] = \hat{p}(r) = p_0 + p_{rr}r$  →

→

In equilibrium:  $\hat{r}(\bar{v}_m) = r(\bar{v}_m)$  →

In equilib.:  $\hat{p}(r) = p(r) = p_0 + p_{rr}r$  →

→

Note that given this setup Investors' →

previous prior is completely dominated →

by  $\bar{v}_m$  as though they were diffuse →

More specifically, to get  $E[v|r]$  investors →

invert their conjectured  $r^{-1}(\bar{v}_m)$  to get  $v_m$  →

**1) Solve for equilibrium policies  $r(\bar{v}_m)$  &  $p(r)$**

Starting with CEO's problem:  $\max_r bE^{Mkt}[v|r] - \frac{c}{2}(r - \bar{v}_m)^2$

Plugging mkt's pricing rule:  $\max_r b(p_0 + p_{rr}) - \frac{c}{2}(r - \bar{v}_m)^2$

FOC:  $0 = bp_r - c(r - \bar{v}_m) \Leftrightarrow r = \frac{b}{c}p_r + \bar{v}_m$

Thus market infers that  $r_0 = \frac{b}{c}p_r$  and  $r_v = 1$

To get the expected value of  $v$  it simply inverts

their conjectured reporting function:  $E[v|r] = r - \frac{b}{c}p_r = \bar{v}_m$

The CEO can estimate the coefficients in his conjecture

of the market's pricing strategy:  $p_r = 1$  and then  $p_0 = -\frac{b}{c}$ .

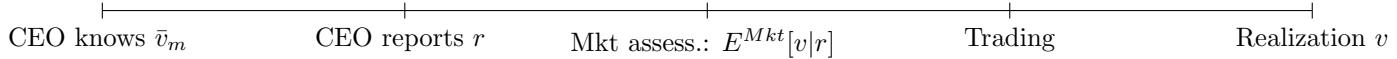
Thus the strategies are  $r(\bar{v}_m) = r_0 + r_v \bar{v}_m = \frac{b}{c} + \bar{v}_m$

$\hat{p}(r) = p(r) = p_0 + p_{rr}r = -\frac{b}{c} + r = \bar{v}_m$

## Assignment 2: Problem 3: Part 2a

Crawford & Sobel (1982): Cheap Talk

*Accounting Theory with Judson Caskey*



CEO is risk **neutral** →  
 Investors believe  $\tilde{v} \sim N(\mu_v, \sigma_v^2)$  →  
 CEO privately believes  $\tilde{v} \sim N(\bar{v}_m, (1 - \delta)\sigma_v^2)$  →  
 where  $cov(v, \bar{v}_m) = var(\bar{v}_m) = \delta\sigma_v^2$  →  
 CEO reports  $r$  to investors to optimize →  
 $\max_r bE^{Mkt}[v|r] - \frac{c}{2}(E^{Mkt}[v|r] - \bar{v}_m)^2$  →  
 $b$  &  $c$  are publicly known parameters →  
 Mkt assess:  $P(r) = E^{Mkt}[v | r]$  →  
 CEO choice of report is  $r(\bar{v}_m)$  →  
 Mkt's conjecture of CEO's reporting →  
 strategy:  $\hat{r}(\bar{v}_m) = r_0 + r_v \bar{v}_m$  →  
 $\hat{p}$ : CEO's conject. of mkt price assessment →  
 has linear form  $\hat{E}^{Mkt}[v|r,v] = \hat{p}(r) = p_0 + p_r r$  →  
 In equilibrium:  $\hat{r}(\bar{v}_m) = r(\bar{v}_m)$  →  
 In equilib.:  $\hat{p}(r) = p(r) = p_0 + p_r r$  →  
**Questions:** →

1) For market's utility  $u_r(p, \bar{v}_m) = -(p - \bar{v}_m)^2$   
 and CEO's utility  $u_s(p, \bar{v}_m, b) = bp - \frac{c}{2}(p - \bar{v}_m)^2$   
 We can check the following conditions to see  
 if this is a Cheap Talk equilibrium:  
 $\frac{\partial u_r}{\partial p} = -(p - \bar{v}_m) \quad \frac{\partial u_s}{\partial p} = b - c(p - \bar{v}_m)$   
 (↑ This means each agent's optimal action depends on the other's,  
 which is a requirement in Cheap Talk models)  
 $\frac{\partial^2 u_r}{\partial p^2} = -1 < 0 \quad \frac{\partial^2 u_s}{\partial p^2} = -c < 0$   
 $\frac{\partial^2 u_r}{\partial p \partial \bar{v}_m} = 1 > 0 \quad \frac{\partial^2 u_s}{\partial p \partial \bar{v}_m} = c > 0$   
 Note that part 1 did not satisfy these as  
 $bp(r) - \frac{c}{2}(r - \bar{v}_m)^2$  so that  $\frac{\partial u_s}{\partial p} = b$  and  $\frac{\partial^2 u_s}{\partial p^2} = \frac{\partial^2 u_s}{\partial p \partial \bar{v}_m} = 0$   
 To prove there isn't a separating equilibrium that fully reveals  $\hat{v} = \bar{v}_m$   
 for a range of types  $\bar{v}_m \in (v_1, v_2)$  by contradiction let's assume there is.  
 Then the equilibrium report is equivalent to claiming a type  $\hat{v}$ .  
 This gives the objective:  $\max_{\hat{v} \in (v_1, v_2)} b\hat{v} - \frac{c}{2}(\hat{v} - \bar{v}_m)^2$   
 But this has a unique optimum at  $\hat{v} = \bar{v}_m + \frac{b}{c}$ , implying that  
 there cannot be any set of types that separate in equilibrium.

**Show that this is a cheap talk equilibrium** →

and that sender's and receiver's implied utility  $u_r$  →

satisfy conditions in Crawford and Sobel (1982) →

where  $u_r = (p, \bar{v}_m) = -(p - \bar{v}_m)^2$  →

and that there is no separating region  $(v_1, v_2)$  such →

that truthful reporting  $\hat{v} = \bar{v}_m$  optimizes: →

$\max_{\hat{v} \in (v_1, v_2)} b\hat{v} - \frac{c}{2}(\hat{v} - \bar{v}_m)^2$  →

→

*Note that this form is the same* →

*as in p.228 of Stocken (2012)* →

**Caskey HW2 Q3 2b) The partitioning equilibrium we are looking for exists if:**

- i)  $\exists$  a price schedule such that  $p(r) = E^{mkt}[v|r]$
- ii)  $\exists$  a reporting schedule  $r(\bar{v}_m)$  that solves:  $\max_r bE^{mkt}[v|r] - \frac{c}{2}(E^{mkt}[v|r] - \bar{v}_m)^2$

Ex-ante, market's optimal price from FOC would be  $p = \bar{v}_m$

For a claimed separating region  $(v_1, v_2)$  the equilibrium would be:

CEO reports  $r(\bar{v}_m) = v_1$  if  $\bar{v}_m < \hat{v}$       and      CEO reports  $r(\bar{v}_m) = v_2$  if  $\bar{v}_m > \hat{v}$

**At threshold  $\hat{v}$  CEO would be indifferent between both partitions:**

$$E^{CEO}[U^{CEO}(p(r = v_1))|\bar{v}_m < \hat{v}] = E^{CEO}[U^{CEO}(p(r = v_2))|\bar{v}_m > \hat{v}]$$

$$E^{CEO}[bE^{mkt}[v|r = v_1] - \frac{c}{2}(E^{mkt}[v|r = v_1] - \hat{v})^2] = E^{CEO}[bE^{mkt}[v|r = v_2] - \frac{c}{2}(E^{mkt}[v|r = v_2] - \hat{v})^2]$$

$$bE^{CEO}[p|v_1] - \frac{c}{2}(E^{CEO}[p|v_1] - \hat{v})^2 = bE^{CEO}[p|v_2] - \frac{c}{2}(E^{CEO}[p|v_2] - \hat{v})^2$$

$$bE[p|\bar{v}_m < \hat{v}] - (E[p|\bar{v}_m < \hat{v}] - \hat{v})^2 = bE[p|\bar{v}_m > \hat{v}] - (\frac{c}{2}E[p|\bar{v}_m > \hat{v}] - \hat{v})^2$$

**Inverse Mills Ratio:**  $E[\bar{v}_m | \bar{v}_m \geq \hat{v}] = \mu_v + \sigma_v \sqrt{\delta} \frac{\phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}{1 - \Phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}$  &  $E[\bar{v}_m | \bar{v}_m < \hat{v}] = \mu_v - \sigma_v \sqrt{\delta} \frac{\phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}{\Phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}$

$$b \left( \mu_v - \frac{\sqrt{\delta} \sigma_v \phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})}{\Phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})} \right) - \frac{c}{2} \left( \mu_v - \frac{\sqrt{\delta} \sigma_v \phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})}{\Phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})} - \hat{v} \right)^2 = b \left( \mu_v + \frac{\sqrt{\delta} \sigma_v \phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})}{1 - \Phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})} \right) - \frac{c}{2} \left( \mu_v + \frac{\sqrt{\delta} \sigma_v \phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})}{1 - \Phi(\frac{\hat{v}-\mu_v}{\sqrt{\delta} \sigma_v})} - \hat{v} \right)^2$$

Then by completing the square:  $bE[v|v_h] - \frac{c}{2}(E[v|v_h] - \hat{v})^2 = bE[v|v_\ell] - \frac{c}{2}(E[v|v_\ell] - \hat{v})^2$

$$\Leftrightarrow 0 = (b+c\hat{v})(E[v|v_h] - E[v|v_\ell]) - \frac{c}{2} \underbrace{(E[v|v_h]^2 - E[v|v_\ell]^2)}_{=(E[v|v_h] - E[v|v_\ell])(E[v|v_h] + E[v|v_\ell])} \Leftrightarrow 0 = b+c\hat{v} - \frac{c}{2}(E[v|v_h] + E[v|v_\ell])$$

$$\hat{v} = \frac{E[v|v > \hat{v}] + E[v|v < \hat{v}]}{2} - \frac{b}{c}$$

$$\hat{v} = \frac{\mu_v + \sigma_v \sqrt{\delta} \frac{\phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}{1 - \Phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})} + \mu_v - \sigma_v \sqrt{\delta} \frac{\phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}{\Phi(\frac{\hat{v}-\mu_v}{\sigma_v \sqrt{\delta}})}}{2} - \frac{b}{c} = f(\xi)$$

## Theory 1 - Comp 2019: Myers & Majluf (1984)

### Problem 1)

CEO's problem:  $\max_{E \in \{0, I\}} V_0^{old} = V(a, E[\tilde{B}], E)$

If CEO forfeits investment:  $V^{old} = a$  with certainty

If CEO issues:  $V^{old} = \frac{P'}{P'+I}(I + a + E[\tilde{B}])$

CEO issues if  $a < \frac{P'}{P'+I}(I + a + E[\tilde{B}]) \Leftrightarrow \frac{I}{P'+I}a < \frac{P'}{P'+I}(I + E[\tilde{B}])$

Firm has project requiring funding of  $I$  that  $\rightarrow$

would need an equity issue  $E = I$  to go through  $\rightarrow$

Firm assets  $\tilde{A} \sim Unif(0, A)$  with  $E_0^{mkt}[\tilde{A}] = \frac{A}{2} \rightarrow$

Project NPV  $\tilde{B} \sim Unif(0, B)$  with  $E[\tilde{B}] = \frac{B}{2} \rightarrow$

CEO has private info on realized value of  $a$  but  $\rightarrow$

**CEO has no private info on realization**  $b \rightarrow$

CEO acts in interest of *old* ( $t=-1$ ) shareholders  $\rightarrow$

Thus CEO maximizes  $V_0^{old} = V(a, E[\tilde{B}], E) \rightarrow$

$P = E_0^{mkt}[\tilde{A}]$  market value if stock is not issued  $\rightarrow$

$P'$  : mkt value of *old* shares if stock issued  $\rightarrow$

$$P' = E_0^{mkt}[\tilde{A}|E] + E_0^{mkt}[\tilde{B}] \rightarrow$$

If firm forfeits investment:  $V^{old} = a \rightarrow$

If it issues:  $V^{old} = \frac{P'}{P'+I}V = \frac{P'}{P'+I}(I + a + E[\tilde{B}]) \rightarrow$

*old* SHs stand to gain if  $a < \frac{P'}{P'+I}(I + a + E[\tilde{B}]) \rightarrow$

$\Leftrightarrow \frac{I}{P'+I}(a) < \frac{P'}{P'+I}(I + E[\tilde{B}]) \Leftrightarrow \frac{I}{P'}a < I + E[\tilde{B}] \rightarrow$

if  $\frac{I}{P'}a > I + E[\tilde{B}]$ : firm does nothing  $\rightarrow E = 0 \rightarrow$

if  $\frac{I}{P'}a < I + E[\tilde{B}]$ : it issues & invests  $\rightarrow E = I \rightarrow$

Firm is indifferent at  $\frac{I}{P'}(a) = I + E[\tilde{B}] \rightarrow$

Assume all agents are risk **neutral**  $\rightarrow$

What is the range of  $a$  for which CEO will issue?  $\rightarrow$

Logic: *old* stockholders only stand to benefit if the expected incremental value they gain (taking into consideration dilution)  $\frac{P'}{P'+I}(I + E[\tilde{B}])$  is greater than the share of existing assets going to *new* stockholders:  $\frac{I}{P'+I}a$

Thus if:  $\frac{I}{P'+I}a < \frac{P'}{P'+I}(I + E[\tilde{B}]) \Leftrightarrow a < \frac{P'}{I}(I + E[\tilde{B}]) = P'(1 + \frac{E[\tilde{B}]}{I}) \Leftrightarrow a < (E^{mkt}[\tilde{A}|E] + E[\tilde{B}])(1 + \frac{E[\tilde{B}]}{I}) \Leftrightarrow a < (E^{mkt}[\tilde{A}|E] + \frac{B}{2})(1 + \frac{B/2}{I})$

Thus market knows that if firm is issuing, it must be because  $a$  is below a certain level where the CEO is willing to dilute its ownership for the sake of the investment. To figure out the value of  $E^{mkt}[\tilde{A}|E]$  we consider the region of  $E^{mkt}[\tilde{A}|E]$  for which the CEO will refuse to dilute the old shareholders because the market valuation of firm assets is too low. I consider a value of  $a^*$  which will render the CEO indifferent between issuing  $E=I$  and not, which would leave old shareholders equally well off:

$$\begin{aligned} E[V^{old}|E=0] &= E[V^{old}|E=I] \\ E[\tilde{A}|E=0] &= \frac{P'}{P'+I}(I + a^* + E[\tilde{B}]) \\ E[a|a=a^*] &= \frac{E[\tilde{A}|E]+E[\tilde{B}]}{E[\tilde{A}|E]+E[\tilde{B}]+I}(I + a^* + E[\tilde{B}]) \\ a^* &= \frac{E[a|a < a^*] + \frac{B}{2}}{E[a|a < a^*] + \frac{B}{2} + I}(I + a^* + \frac{B}{2}) \\ a^* &= \frac{\frac{a^*}{2} + \frac{B}{2}}{\frac{a^*}{2} + \frac{B}{2} + I}(a^* + \frac{B}{2} + I) \\ (\frac{a^*}{2} + \frac{B}{2} + I)a^* &= (\frac{a^*}{2} + \frac{B}{2})(a^* + \frac{B}{2} + I) \\ \frac{(a^*)^2}{2} + \frac{B}{2}a^* + Ia^* &= (\frac{(a^*)^2}{2} + \frac{B}{4}a^* + \frac{a^*}{2}I) + (\frac{B}{2}a^* + \frac{B^2}{4} + \frac{BI}{2}) \\ \frac{a^*}{2}I &= \frac{B}{4}a^* + \frac{B^2}{4} + \frac{B}{2}I \Leftrightarrow a^*I = \frac{B}{2}a^* + \frac{B^2}{2} + BI \\ \Leftrightarrow (I - \frac{B}{2})a^* &= (I + \frac{B}{2})B \Leftrightarrow a^* = \frac{I + \frac{B}{2}}{I - \frac{B}{2}}B \\ a^* &= \frac{2I+B}{2I-B}B \end{aligned}$$

Thus **CEO will issue & take on investment for**  $a < \frac{2I+B}{2I-B}B$ .

2

**Problem 2)**

**Case 1: Expected  $V^{old}$  when not paying fee  $F$  ex-ante**

Firm has project requiring funding of  $I$  that →  
would need an equity issue  $E = I$  to go through →

Firm assets  $\tilde{A} \sim Unif(0, A)$  with  $E_0^{mkt}[\tilde{A}] = \frac{A}{2}$  →

Project NPV  $\tilde{B} \sim Unif(0, B)$  with  $E[\tilde{B}] = \frac{B}{2}$  →

CEO has private info on realized value of  $a$  but →

**CEO has no private info on realization  $b$**  →

CEO acts in interest of *old* ( $t=-1$ ) shareholders →

Thus CEO maximizes  $V_0^{old} = V(a, E[\tilde{B}], E)$  →

$P = E_0^{mkt}[\tilde{A}]$  market value if stock is not issued →

$P'$ : mkt value of *old* shares if stock issued →

$$P' = E_0^{mkt}[\tilde{A}|E] + E_0^{mkt}[\tilde{B}] \rightarrow$$

If firm forfeits investment:  $V^{old} = a$  →

If it issues:  $V^{old} = \frac{P'}{P'+I}V = \frac{P'}{P'+I}(I + a + E[\tilde{B}])$  →

*old* SHs stand to gain if  $a < \frac{P'}{P'+I}(I + a + E[\tilde{B}])$  →

$\Leftrightarrow \frac{I}{P'+I}(a) < \frac{P'}{P'+I}(I + E[\tilde{B}]) \Leftrightarrow \frac{I}{P'}a < I + E[\tilde{B}]$  →

if  $\frac{I}{P'}a > I + E[\tilde{B}]$ : firm does nothing →  $E = 0$  →

if  $\frac{I}{P'}a < I + E[\tilde{B}]$ : it issues & invests →  $E = I$  →

Firm is indifferent at  $\frac{I}{P'}(a) = I + E[\tilde{B}]$  →

Assume all agents are risk **neutral** →

Assume CEO can pay auditor \$ $F$  to reveal →

$a$ , but must decide to do so before seeing  $a$  →

What's the max  $F$  he would be willing to pay? →

Let us consider the case where CEO did not pay F.

Since his decision to pay F is made at date  $-1$  before seeing the realization of  $a$ , the market cannot infer anything about  $a$  from this decision as both the investors and CEO have the same priors for it.

So in the case of forgoing the audit ex-ante, knowing that  $a^* = \frac{2I+B}{2I-B}B$  the expected outcome is:  $E_{-1}[V^{old}|F \text{ not paid}] = P(0| \text{no } F)E_{-1}[V^{old}|E = 0, \text{ no } F] + P(I| \text{no } F)E_{-1}[V^{old}|E = I, \text{ no } F]$

Taking it one step at a time:

$$P(E = 0| \text{no } F) = P(a > a^*) = \frac{A - a^*}{A} = 1 - \frac{a^*}{A}$$

$$E_{-1}[V^{old}|E = 0, \text{ no } F] = E_{-1}[a|a > a^*] = \frac{A}{2} + \frac{a^*}{2}$$

$$P(I| \text{no } F) = P(a < a^*) = \frac{a^*}{A}$$

$$E_{-1}[V^{old}|E = I, \text{ no } F] = \frac{E_{-1}[P'|E]}{E_{-1}[P'|E]+I}E_{-1}[V]$$

$$= \frac{E[\tilde{A}|E] + E[\tilde{B}]}{E[\tilde{A}|E] + E[\tilde{B}] + I}(I + E[\tilde{A}|E] + E[\tilde{B}])$$

$$= \frac{E[\tilde{A}|a < a^*] + \frac{B}{2}}{E[\tilde{A}|a < a^*] + \frac{B}{2} + I}(I + E[\tilde{A}|a < a^*] + \frac{B}{2})$$

$$= \frac{\frac{a^*}{2} + \frac{B}{2}}{\frac{a^*}{2} + \frac{B}{2} + I}(I + \frac{A}{2} + \frac{B}{2}) = \frac{\frac{a^*}{2} + \frac{B}{2}}{\frac{a^*}{2} + \frac{B}{2} + I}(I + \frac{a^*}{2} + \frac{B}{2})$$

$$E_{-1}[V^{old}|E = I, \text{ no } F] = \frac{a^*}{2} + \frac{B}{2}$$

$$\text{Thus } E_{-1}[V^{old}| \text{ no } F] = \left(1 - \frac{a^*}{A}\right) \left[\frac{A}{2} + \frac{a^*}{2}\right] + \frac{a^*}{A} \left[\frac{a^*}{2} + \frac{B}{2}\right]$$

$$E_{-1}[V^{old}| \text{ no } F] = \left(\frac{A}{2} + \frac{a^*}{2} - \frac{a^*A}{2A} - \frac{(a^*)^2}{2A}\right) + \left[\frac{(a^*)^2}{2A} + \frac{a^*B}{2A}\right]$$

$$E_{-1}[V^{old}| \text{ no } F] = \frac{A}{2} + \frac{a^*B}{2A}$$

**Case 2: Expected  $V^{old}$  when paying fee  $F$  ex-ante**

Again since the decision to pay F is made at date  $-1$ , the audit would reveal the value of the value of  $\tilde{A}$  which would result in the following expected value for old shareholders in case of investment:

$$\begin{aligned} E_{-1}[V^{old}|E = I, F \text{ paid}] &= E_{-1}\left[\frac{P'}{P'+I}(I + E_{-1}[\tilde{A}] + E[\tilde{B}])\right] - F \\ &= E_{-1}\left[\frac{E_{-1}[\tilde{A}|E,F] + E_{-1}[\tilde{B}]}{E_{-1}[A|E,F] + E_{-1}[B] + I}(I + E_{-1}[\tilde{A}] + E[\tilde{B}])\right] - F \\ E_{-1}[V^{old}|E = I, F \text{ paid}] &= \frac{\frac{A}{2} + \frac{B}{2}}{\frac{A}{2} + \frac{B}{2} + I}(I + \frac{A}{2} + \frac{B}{2}) - F \\ E_{-1}[V^{old}|E = I, F \text{ paid}] &= \frac{A}{2} + \frac{B}{2} - F \end{aligned}$$

We also know that in case CEO paid the fee but forgoes the investment:

$$E_{-1}[V^{old}|E = 0, F \text{ paid}] = E[\tilde{A}] - F = \frac{A}{2} - F$$

Since this is always strictly inferior to the case in which

CEO issues after paying the fee it's not considered. Thus:

$$E_{-1}[V^{old}|F \text{ paid}] = E_{-1}[V^{old}|E = I, F \text{ paid}] = \frac{A}{2} + \frac{B}{2} - F$$

And so comparing

$$\begin{aligned} \text{Case 1: } E_{-1}[V^{old}| \text{ no } F] &= \left(1 - \frac{a^*}{A}\right) \left[\frac{A}{2} + \frac{a^*}{2}\right] + \frac{a^*}{A} \left[\frac{a^*}{2} + \frac{B}{2}\right] = \frac{A}{2} + \frac{a^* B}{2A} \\ \text{to} \end{aligned}$$

$$\text{Case 2: } E_{-1}[V^{old}|F \text{ paid}] = \frac{A}{2} + \frac{B}{2} - F$$

Implies that the maximum fee CEO is will to pay is

$$F = \left(\frac{A}{2} + \frac{B}{2}\right) - \left(\frac{A}{2} + \frac{a^* B}{2A}\right) = \frac{B}{2} - \frac{a^* B}{2A} = \frac{B}{2} \left(1 - \frac{a^*}{A}\right) = \frac{B}{2} \cdot \text{Prob}(a > a^*)$$

which (for clarity) is :

$$F = \frac{A}{2} + \frac{B}{2} - E_{-1}[V^{old}| \text{ no } F \text{ paid}]$$

$$F = \frac{A}{2} + \frac{B}{2} - [P(E = 0| \text{ no } F)E_{-1}[V^{old}|E = 0, \text{ no } F] + P(E = I| \text{ no } F)E_{-1}[V^{old}|E = I, \text{ no } F]]$$

$$F = \frac{A}{2} + \frac{B}{2} - \left[P(a > a^*)E_{-1}[a|a > a^*] + P(a < a^*)\frac{E[\tilde{A}|a < a^*] + E[\tilde{B}]}{E[\tilde{A}|a < a^*] + E[\tilde{B}] + I}(E[\tilde{A}|a < a^*] + E[\tilde{B}] + I)\right]$$

$$F = \frac{A}{2} + \frac{B}{2} - \left[P(a > a^*)E_{-1}[a|a > a^*] + P(a < a^*)\frac{\frac{a^*}{2} + \frac{B}{2}}{\frac{a^*}{2} + \frac{B}{2} + I}(I + \frac{a^*}{2} + \frac{B}{2})\right]$$

$$F = \frac{A}{2} + \frac{B}{2} - \left[\left(1 - \frac{a^*}{A}\right) \left[\frac{A}{2} + \frac{a^*}{2}\right] + \frac{a^*}{A} \left[\frac{a^*}{2} + \frac{B}{2}\right]\right]$$

# Theory 1 - 2020

by Dimitri Zafirov

## Economic Setting

- Risk neutral investors
  - Manager knows the value of  $x$  precisely but cannot credibly convey it without an auditor charging fee  $F$ .
  - $x \sim Unif(0, 1)$
- (a) If manager pays auditor fee to certify the value of  $x$ , his payoff is  $x - F$ . If he opts not to hire the auditor, investors' expectations will place the value of  $x$  below that of a certain threshold  $x^*$ . To determine this threshold,  $x^*$ , the manager is indifferent between disclosing and not disclosing. Thus his payoffs would be equal at  $x^*$ :

$$x^* - F = E_{mkt}[x|x < x^*]$$

$$x^* - F = E_{mkt}[Unif(0, x^*)]$$

$$x^* - F = x^*/2$$

$$x^* = 2F$$

Thus  $F \geq 1/2$  would result in the threshold being  $x^* \geq 1$ . Given that the upper bound of  $x$  is 1, this would mean that managers would never hire an auditor charging a fee  $F \geq 1/2$ . Conversely, given the extent of the values  $x$  can take ( $[0,1]$ ), for  $F < 1/2$ , there exists a manager with cashflow  $x$  greater or equal to  $2F$  that would pay the fee.

- (b) Auditor's expected profits are:  $Fee \cdot Prob(Fee)$

Auditor's objective function:  $\max_F F \cdot P(x > x^*) = \max_F F \cdot [1 - F(x^*)] = \max_F F \cdot (1 - x^*)$

Knowing the entrepreneur's above optimization strategy, auditor conjectures that entrepreneur's optimal  $x^* = 2F$  and plugs this into this own objective function:  $\max_F F \cdot (1 - 2F)$

FOC w.r.t.  $F$ :  $0 = 1 - 4F \Leftrightarrow F = 1/4$

---

(c) The equilibrium where all CEOs hire the auditor when  $F < 1/2$  can exist under the following circumstances:

- Since everyone pays the fee, its signal is uninformative, therefore  $E^{mkt}[x|F \text{ paid}] = E^{mkt}[x] = 1/2$
- The off-equilibrium belief  $\mu(F)$  that supports this is  $E^{mkt}[x| F \text{ not paid }] = 0$ .
- These beliefs are realized in this pooling equilibrium since no CEO has any benefit in deviating from hiring the auditor, given that  $E^{mkt}[x] = 1/2 - F > 0$  for  $F < 1/2$ .

(d) The equilibrium where no CEOs hire the auditor when  $F > 1/2$  can exist under the following circumstances:

- Since no one pays the fee, its signal is uninformative, therefore  $E^{mkt}[x|F \text{ not paid}] = E^{mkt}[x] = 1/2$
- The off-equilibrium belief  $\mu(F)$  that supports this is  $E^{mkt}[x| F \text{ paid }] = 1$ .
- These beliefs are realized in this pooling equilibrium since no CEO has any benefit in deviating to hire the auditor, given that  $E^{mkt}[x] = 1/2 > 1 - F$  for  $F > 1/2$ .

## Theory 3 Comp 2019:

Grossman & Stiglitz (1980)

Informed see private signal      Trading      Realization  $v$

$$\begin{pmatrix} v \\ \hat{v} \\ x \end{pmatrix} \sim N \begin{pmatrix} \bar{v} \\ \bar{v} \\ \bar{x} \end{pmatrix}, \begin{pmatrix} \sigma_v^2 & \delta\sigma_v^2 & \rho_{vx}\sigma_v\sigma_x \\ \delta\sigma_v^2 & \delta\sigma_v^2 & 0 \\ \rho_{vx}\sigma_y\sigma_x & 0 & \sigma_x^2 \end{pmatrix}$$

Demands:  $q_i = \frac{E_i[v|\hat{v},x] - p}{r \cdot var_i(v|\hat{v},x)}$ ,  $q_u = \frac{E_u[v|s_u] - p}{r \cdot var_u(v|s_u,x)}$

Note:  $x$  is in the conditional because in this setting it contains information since it's correlated with  $v$ .

Firm has value  $v \sim N(\bar{v}, \sigma_v^2) \rightarrow$

What follows uses these Week 3 relations for 3 joint-normals  $\mathbf{X}, \mathbf{y}_1, \mathbf{y}_2$

Investors have CARA risk-aversion  $r \rightarrow$

$\alpha\%$  of investors are *Informed*  $\rightarrow$

*Informed* posterior is the following:  $\rightarrow$

$v|Info \sim N(\hat{v}, (1-\delta)\sigma_v^2)$  with  $\delta \in (0,1) \rightarrow$

Noisy supply of shares  $x \sim N(\bar{x}, \sigma_x^2) \rightarrow$

$x$  is correlated with  $v \rightarrow$

$$E[\mathbf{X} | \mathbf{y}_1, \mathbf{y}_2] = E[\mathbf{X} | \mathbf{y}_1] + cov(\mathbf{X}, \mathbf{y}_2 | \mathbf{y}_1) var(\mathbf{y}_2 | \mathbf{y}_1)^{-1} (\mathbf{y}_2 - E[\mathbf{y}_2 | \mathbf{y}_1])$$

$$var(\mathbf{x} | \mathbf{y}_1, \mathbf{y}_2) = var(\mathbf{x} | \mathbf{y}_1) - \underbrace{cov(\mathbf{x}, \mathbf{y}_2 | \mathbf{y}_1) var(\mathbf{y}_2 | \mathbf{y}_1)^{-1} cov(\mathbf{y}_2, \mathbf{x} | \mathbf{y}_1)}_{var(E[\mathbf{x} | \mathbf{y}_1, \mathbf{y}_2] | \mathbf{y}_1)}$$

$$cov(v, x | \hat{v}) = cov(v, x)$$

$$var(x | \hat{v}) = var(x)$$

$$E[x | \hat{v}] = [x]$$

$$var_i(v) = var(v | \hat{v}, x) = (1-\delta)\sigma_v^2 - \frac{\rho_{vx}\sigma_v^2\sigma_x^2}{\sigma_x^2} = (1-\delta - \rho_{vx}^2)\sigma_v^2$$

But  $x$  is not correlated with private info  $\rightarrow$

Distributional assumptions are above  $\rightarrow$

Conjecture of market pricing rule:  $p = \pi_0 + \pi_v (E[v|\hat{v},x] - \bar{v}) + \pi_x (x - \bar{x})$

Note that  $\rho_{vx} \in (0, \sqrt{1-\delta}) \rightarrow$

*Uninformed* infer  $\hat{v}$  from the price signal:

$$s_u = \bar{v} + \frac{1}{\pi_v} (p - \pi_0) = E_i[v] + \frac{\pi_x}{\pi_v} (x - \bar{x}) = E[v|\hat{v},x] + b(x - \bar{x})$$

$$E[v|s_u] = E[v] + \frac{cov(v, s_u)}{var(s_u)} (s_u - E[s_u]) = \bar{v} + \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + b^2\sigma_x^2} (s_u - \bar{v})$$

$$E[v|s_u] = \bar{v} + \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2\sigma_x^2} \frac{1}{\pi_v} (p - \pi_0)$$

$$var(v|s_u) = \sigma_v^2 - \frac{(\delta + \rho_{vx}^2)^2 \sigma_v^4}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2\sigma_x^2} = \sigma_v^2 \left( 1 - \frac{(\delta + \rho_{vx}^2)^2 \sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2\sigma_x^2} \right)$$

**By Market Clearing:** Supply = Demand  $\rightarrow x = \alpha q_i + (1-\alpha)q_u$

$$x = \alpha \frac{[\hat{v} + \frac{\rho_{vx}\sigma_v}{\sigma_x}(x - \bar{x})] - p}{r \cdot (1-\delta - \rho_{vx}^2)\sigma_v^2} + (1-\alpha) \frac{\bar{v} + \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2\sigma_x^2} \frac{1}{\pi_v} (p - \pi_0) - p}{r \cdot \sigma_v^2 \left( 1 - \frac{(\delta + \rho_{vx}^2)^2 \sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2\sigma_x^2} \right)}$$

continued on next page

**By Market Clearing:** Supply = Demand  $\rightarrow x = \alpha q_i + (1 - \alpha)q_u$

$$\begin{aligned}
x &= \alpha \frac{[\hat{v} + \frac{\rho_{vx}\sigma_v}{\sigma_x}(x - \bar{x})] - p}{r \cdot (1 - \delta - \rho_{vx}^2) \sigma_v^2} + (1 - \alpha) \frac{\bar{v} + \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2 \sigma_x^2} \frac{1}{\pi_v} (p - \pi_0) - p}{r \cdot \sigma_v^2 \left( 1 - \frac{(\delta + \rho_{vx}^2)^2 \sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2 \sigma_x^2} \right)} \\
p &= \underbrace{\frac{\alpha}{r \cdot var_i(v)} \bar{v} + (1 - \alpha) \frac{\bar{v} - \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2 \sigma_x^2} \frac{1}{\pi_v} \pi_0}{r \cdot var_u(v)} - \bar{x}}_{\pi_0} + \underbrace{\frac{\frac{\alpha}{r \cdot var_i(v)}}{1 - \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2 \sigma_x^2} \frac{1}{\pi_v}} (\mathbb{E}_i[v] - \bar{v})}_{\pi_v} \\
&\quad - \underbrace{\frac{1}{\frac{\alpha}{r \cdot var_i(v)} + (1 - \alpha) \frac{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{\pi_x}{\pi_v})^2 \sigma_x^2 \pi_v}{r \cdot var_u(v)}}}_{\pi_x} (x - \bar{x})
\end{aligned}$$

1. To solve coefficients, first use:

$$\frac{\pi_x}{\pi_v} = -\frac{r \cdot var_i(v)}{\alpha}$$

which gives  $\frac{\pi_x}{\pi_v}$  and, hence,  $var_u(v)$ , in terms of model parameters.

2. Next solve for  $\pi_v$  :

$$\pi_V = \frac{\frac{\alpha}{r \cdot var_i(v)}}{1 - \frac{(\delta + \rho_{vx}^2)\sigma_v^2}{(\delta + \rho_{vx}^2)\sigma_v^2 + (\frac{r \cdot var_i(v)}{\alpha})^2 \sigma_x^2} \frac{1}{\pi_v} \pi_v} r \cdot var_u(v)$$

which is in terms of model parameters.

3. Multiply  $\pi_v$  by  $\frac{\pi_x}{\pi_v}$  to get  $\pi_x$

4. Lastly, solve for  $\pi_0$  :

$$\pi_0 = \bar{v} - \frac{1}{\frac{\alpha}{r \cdot var_i(v)} + \frac{1 - \alpha}{r \cdot var_u(v)}} \bar{x}$$

**Q2) Why do uninformed traders participate in the market?**

- They can still some of the informed traders' information from the price signal. Though it will be made noisier from the random supply  $x$ , uninformed traders can still make a profit from noise traders given their conjecture.

**Q3) Without necessarily solving the model, which parameters would you expect to increase the expected value of being privately informed?**

- $\uparrow \sigma_x^2$  would make it easier for informed traders to camouflage their flow.
- $\uparrow (1-\alpha)$  the smaller the share of informed traders, the easier it will be for them to camouflage their flow and benefit from their private information.
- $\uparrow r^{-1}$  higher risk tolerance means informed traders will trade more aggressively on their information.
- $\uparrow \delta$  the more precise the private information, the more useful it is in forming accurate expectations of the realized value of the asset. A more precise forecast will make knowing which trades will be profitable easier.
- $\uparrow \hat{v}$  (holding  $\bar{v}$  constant) higher posterior average means higher prices thus more profit from trading with uninformed investors forced to work with the prior mean.

*The value of being informed will be higher when there are fewer informed (low  $\alpha$ ). The effects of precision ( $\delta$ ), risk-aversion ( $r$ ) and noise trade ( $\sigma_x$ ) are tricky. For example, high precision increases both the informed and uninformed value functions. When information gets extremely precise, informed traders face such low risk that they reveal most of their information and earn little compared to uninformed traders. Risk aversion may seem to diminish the value of being informed because it leads to smaller positions, but with price-taking traders it also limits aggressive trade and can lead to higher expected utility.*

**Q4) Suppose that instead of  $\alpha$  being given exogenously, there was an initial round in which investors choose simultaneously to acquire the private information at some cost. What condition would have to hold in any equilibrium with an interior fraction of informed investors?**

- For an internal equilibrium where  $0 < \alpha < 1$ , the value of  $\alpha^e$  will be such that the net benefit of acquiring information will equal the benefit of remaining uninformed:

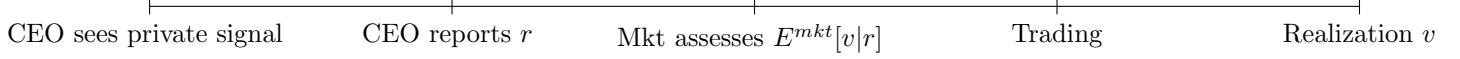
$$EV(W_I^\alpha) = EV(W_U^\alpha)$$

If the cost was, say,  $c$ , the equilibrium would satisfy:

$$\underbrace{E \left[ \frac{1}{2} \frac{(E_i[v] - p)^2}{r \cdot var_i(v)} \right] - C}_{\text{Net value of being informed}} = \underbrace{E \left[ \frac{1}{2} \frac{(E_u[v] - p)^2}{r \cdot var_u(v)} \right]}_{\text{Value of being uninformed}}$$

## Theory 3 Comp 2020 Q1:

Fischer & Verrecchia (2000) and Grossman & Stiglitz (1980)



- Public priors of asset:  $v \sim N(\bar{v}, \sigma_v^2)$  →
- Traders have CARA risk aversion  $\rho$  →
- CEO sees private signal which gives →  
posterior belief  $\bar{v}_m \sim N(\bar{v}, \delta\sigma_v^2)$  →
- where  $\text{cov}(v, \bar{v}_m) = \delta\sigma_v^2$  with  $\delta \in (0, 1)$  →
- CEO produces report  $r$  before trade →
- CEO maximizes  $\max_r bp - \frac{c}{2} (r - \bar{v}_m)^2$  →
- $b$  is a parameter known only to CEO** →
- Investors have prior  $b \sim N(\bar{b}, \sigma_b^2)$  →
- Equilibrium price sets investor demand →  
equal to the **known** supply  $x$  →
- Add  $x$  to the pricing rule conjecture** →
- CEO's report function is  $r(\bar{v}_m, b)$  →
- $\hat{r}(\bar{v}_m, b)$ : Mkt conjecture of CEO's  $r(\bar{v}_m, b)$  →
- has linear form  $\hat{r}(\bar{v}_m, b) = \hat{r}_0 + \hat{r}_v \bar{v}_m + \hat{r}_b b$  →
- In equilibrium:  $\hat{r}(\bar{v}_m, b) = r(\bar{v}_m, b)$  →
- Market asseses  $P(r) = E^{mkt}[v | r]$  →
- $\hat{P}$ : CEO's conjecture of mkt pricing rule →  
has linear form  $\hat{P}(r) = \hat{p}_0 + \hat{p}_r r - \hat{p}_x x$  →
- In equilib.:  $\hat{P}(r) = P(r) = p_0 + p_r r - p_x x$  →

### Q1) Solve for the equilibrium price and report

Starting with the CEO's problem given his conjecture of the pricing rule

$$\max_r bp - \frac{c}{2} (r - \bar{v}_m)^2 = \max_r b(p_0 + p_r r - p_x x) - \frac{c}{2} (r - \bar{v}_m)^2$$

FOC:  $0 = bp_r - c(r - \bar{v}_m) \Rightarrow r = \bar{v}_m + \frac{p_r}{c} b$

Then the market's expected value of  $v$  given its conjecture  $\hat{r}(\bar{v}_m, b)$

and taking into consideration from above that  $r_0 = 0$ ;  $r_v = 1$ ;  $r_b = \frac{p_r}{c}$

$$E[v|r] = E[v] + \frac{\text{cov}(v, r_0 + r_v \bar{v}_m + r_b b)}{\text{var}(r_0 + r_v \bar{v}_m + r_b b)}(r - E[r]) = \bar{v} + \frac{\text{cov}(v, \bar{v}_m + \frac{p_r}{c} b)}{\text{var}(\bar{v}_m + \frac{p_r}{c} b)}(r - E[\bar{v}_m + \frac{p_r}{c} b])$$

$$E^{mkt}[v|r] = \bar{v} + \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}(r - \bar{v} - \frac{p_r}{c}\bar{b}) \quad \text{and} \quad \text{var}(v|r) = \sigma_v^2 - \frac{\delta^2\sigma_v^4}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}$$

$$\text{Thus demand } q = \frac{E^{mkt}[v|r] - p}{\rho \cdot \text{var}(v|r)} = \frac{\bar{v} + \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}(r - \bar{v} - \frac{p_r}{c}\bar{b}) - p}{\rho \left[ \sigma_v^2 - \frac{\delta^2\sigma_v^4}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2} \right]} = x \text{ supply}$$

$$\text{Isolate } p: \rho \text{var}(v|r)x = E^{mkt}[v|r] - p \Leftrightarrow p = E^{mkt}[v|r] - \rho \text{var}(v|r)x$$

$$p = \bar{v} - \underbrace{\frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}(\bar{v} + \frac{p_r}{c}\bar{b})}_{p_0} + \underbrace{\frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}r}_{p_r} - \underbrace{\rho \left[ \sigma_v^2 - \frac{\delta^2\sigma_v^4}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2} \right] x}_{p_x}$$

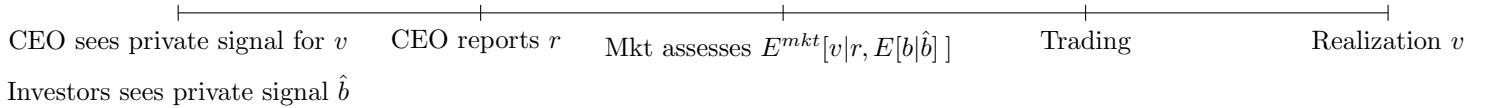
$$p_r = \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2} = \frac{1}{1 + (\frac{p_r}{c})^2 \frac{\sigma_b^2}{\delta\sigma_v^2}} \Rightarrow 0 = p_r^3 + c^2 \frac{\delta\sigma_v^2}{\sigma_b^2} p_r - c^2 \frac{\delta\sigma_v^2}{\sigma_b^2}$$

This has a unique solution  $p_r \in (0, 1)$  per Descartes's rule of signs and that the

right-hand-side of the inequality is negative (positive) at  $p_r = 0$  ( $p_r = 1$ ).

Since  $p_r = \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2\sigma_b^2}$ , price rule is  $p = \bar{v} + p_r(r - \bar{v} - \frac{p_r}{c}\bar{b}) - \rho\sigma_v^2(1 - p_r\delta)x$

### Theory 3 Comp 2020 Q2:



Public priors of asset: $v \sim N(\bar{v}, \sigma_v^2)$ →	
Traders have CARA risk aversion $\rho$ →	
CEO sees private signal which gives →	<b>Q2) how does <math>E[(r - \bar{v}_m)^2]</math> vary with <math>\gamma</math>?</b>
posterior belief $\bar{v}_m \sim N(\bar{v}, \delta\sigma_v^2)$ →	
where $cov(v, \bar{v}_m) = \delta\sigma_v^2$ with $\delta \in (0, 1)$ →	$E[b   \hat{b}] = \hat{b}$ and $cov(b, \hat{b}) = var(\hat{b}) = \gamma\sigma_b^2$ with $\gamma \in (0, 1)$
CEO produces report $r$ before trade →	
CEO maximizes $\max_r bp - \frac{c}{2}(r - \bar{v}_m)^2$ →	$var(b \hat{b}) = var(b) - \frac{cov^2(b, \hat{b})}{var(b)} = \sigma_b^2 - \frac{\gamma^2\sigma_b^4}{\gamma\sigma_b^2} = \sigma_b^2 - \gamma\sigma_b^2 = (1 - \gamma)\sigma_b^2$
<b>b is a parameter known only to CEO</b> →	
Investors have prior $b \sim N(\bar{b}, \sigma_b^2)$ →	
Equilibrium price sets investor demand →	$E^{mkt}[v r] = \bar{v} + \frac{cov(v, \bar{v}_m + \frac{p_r}{c}\hat{b})}{var(\bar{v}_m + \frac{p_r}{c}\hat{b})}(r - E[\bar{v}_m + \frac{p_r}{c}\hat{b}]) = \bar{v} + \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2(1-\gamma)\sigma_b^2}(r - \bar{v} - \frac{p_r}{c}\hat{b})$
equal to the <b>known</b> supply $x$ →	$var(v r) = \sigma_v^2 - \frac{\delta^2\sigma_v^4}{\delta\sigma_v^2 + (\frac{p_r}{c})^2(1-\gamma)\sigma_b^2}$
<b>Add <math>x</math> to the pricing rule conjecture</b> →	
CEO's report function is $r(\bar{v}_m, b)$ →	In a similar way to before we can find $p_r = \frac{\delta\sigma_v^2}{\delta\sigma_v^2 + (\frac{p_r}{c})^2(1-\gamma)\sigma_b^2}$
$\hat{r}(\bar{v}_m, b)$ : Mkt conjecture of CEO's $r(\bar{v}_m, b)$ →	
has linear form $\hat{r}(\bar{v}_m, b) = \hat{r}_0 + \hat{r}_v \bar{v}_m + \hat{r}_b b$ →	$Then E[(r - \bar{v}_m)^2] = (\frac{p_r}{c})^2 E[b^2] = (\frac{p_r}{c})^2 (\sigma_b^2 + \bar{b}^2)$
In equilibrium: $\hat{r}(\bar{v}_m, b) = r(\bar{v}_m, b)$ →	
Market asseses $P(r) = E^{mkt}[v   r]$ →	$\frac{dE[(r - \bar{v}_m)^2]}{d\sigma_b^2} = \underbrace{\left(\frac{p_r}{c}\right)^2}_{\frac{\partial E[(r - \bar{v}_m)^2]}{\partial \sigma_b^2}} + 2\underbrace{\frac{p_r}{c^2}(\sigma_b^2 + \bar{b}^2)}_{\frac{\partial E[(r - \bar{v}_m)^2]}{\partial p_r}} \underbrace{\left(-\frac{c^2 \delta \sigma_v^2 (1 - p_r)}{3p_r^2 + c^2 \frac{\delta \sigma_v^2}{\sigma_b^2}}\right)}_{\frac{dp_r}{d\sigma_b^2}}$
$\hat{P}$ : CEO's conjecture of mkt pricing rule →	
has linear form $\hat{P}(r) = \hat{p}_0 + \hat{p}_r r - \hat{p}_x x$ →	
In equilib.: $\hat{P}(r) = P(r) = p_0 + p_r r - p_x x$ →	
<b>Question 2 additions:</b> →	

Prior to trade, investors receive a signal  $\hat{b}$  →  
 where  $cov(b, \hat{b}) = var(\hat{b}) = \gamma\sigma_b^2$  &  $\gamma \in (0, 1)$  →  
 $E[b | \hat{b}] = \hat{b}$  and  $cov(\hat{b}, v) = cov(\hat{b}, x) = 0$  →