

Accepted Manuscript

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PII: S0304-3932(18)30457-4

DOI: <https://doi.org/10.1016/j.jmoneco.2018.08.002>

Reference: MONEC 3033

To appear in: *Journal of Monetary Economics*

Received date: 13 February 2017

Revised date: 7 August 2018

Accepted date: 7 August 2018

Please cite this article as: Dan Cao, Guido Lorenzoni, Karl Walentin, Financial Frictions, Investment, and Tobin's q , *Journal of Monetary Economics* (2018), doi: <https://doi.org/10.1016/j.jmoneco.2018.08.002>



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1 Highlights

- 2 • A model of firm-level investment with fully state-contingent financial constraints.
- 3 • A characterization of Tobin's q in presence of financial frictions.
- 4 • Two forces are driving q : the value of invested capital and future quasi-rents.
- 5 • This weakens the correlation between investment and q .
- 6 • Implications for investment regressions depend crucially on the nature of shocks.

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Financial Frictions, Investment, and Tobin's q

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June 2018

Abstract

A model of investment with financial constraints is used to study the relation between investment and Tobin's q . A firm is financed by both inside and outside investors. When insiders' wealth is scarce, the firm's value includes a quasi-rent on invested capital. Therefore, two forces drive q : the value of invested capital and future quasi-rents. Relative to a frictionless benchmark, this weakens the relationship between investment and q , generating more realistic correlations between investment, q , and cash flow. The quantitative implications of the model for investment regressions depend crucially on the nature of the shocks hitting the firm.

Keywords: Financial constraints, optimal financial contracts, investment, Tobin's q , limited enforcement.

JEL codes: E22, E30, E44, G30.

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23 1 Introduction

24 Dynamic models of the firm imply that investment decisions and the value of the firm
25 should both respond to expectations about future profitability of capital. In models with
26 constant returns to scale and convex adjustment costs these relations are especially clean,
27 as investment and the firm's value respond exactly in the same way to new information
28 about future profitability. This is the main prediction of Tobin's q theory, which implies
29 that current investment moves one-for-one with q , the ratio of the firm's financial market
30 value to its capital stock. This prediction, however, is typically rejected in the data, where
31 investment appears to correlate more strongly with current cash flow than with q .

32 In this paper, we investigate the relation between investment, q , and cash flow in a
33 model with financial frictions. The presence of financial frictions introduces quasi-rents
34 in the market valuation of the firm. These quasi-rents break the one-to-one link between
35 investment and q . We study how the presence of these quasi-rents affects the statistical
36 correlations between investment, q , and cash flow, and ask whether a model with financial
37 frictions can match the correlations in the data.

38 Our main conclusion is that the presence of financial frictions can bring the model
39 closer to the data, but that the model's implications depend crucially on the shock struc-
40 ture. In a model with financial frictions it is still true that investment and q respond to fu-
41 ture profitability, but the two variables now respond differently to information at different
42 horizons. Investment is particularly sensitive to current profitability, which determines
43 current internal financing, and to near-term financial profitability, which determines col-
44 lateral values. On the other hand, q is relatively more sensitive to profitability farther
45 in the future, which will determine future growth and thus the size of future quasi-rents.
46 Therefore, to break the link between investment and q , requires the presence of both short-
47 lived shocks—which tend to move investment more and have relatively smaller effects on
48 q —and long-lived shocks—which do the opposite.

49 These points are developed in the context of a stochastic model of investment sub-
50 ject to limited enforcement, with fully state-contingent claims. The ability of borrowers
51 to issue state-contingent claims is limited by the fact that, *ex post*, they can renege on
52 their promises and default. The consequence of default is the loss of a fraction of in-
53 vested assets. We show that this environment is equivalent to an environment with state-
54 contingent collateral constraints, so the model is essentially a stochastic version of Kiy-
55 otaki and Moore (1997) with adjustment costs and state-contingent claims.¹ The model

¹Related recent stochastic models that combine state-contingent claims with some form of collateral constraint include Lorenzoni (2008), He and Krishnamurthy (2013), Rampini and Viswanathan (2013), Cao and Nie (2017), and Di Tella (2017).

56 leads to a wedge between average q —which correspond to the q measured from financial
 57 market values—and marginal q —which captures the marginal incentive to invest and is
 58 related one-to-one to investment.² Two versions of the model are analyzed, looking at
 59 their implications for investment regressions in which the investment rate is regressed on
 60 average q and cash flow.

61 The first version of the model features no adjustment costs and, under some simplify-
 62 ing assumptions, it can be linearized and studied analytically. When a single persistent
 63 shock is introduced, the model has indeterminate predictions regarding investment re-
 64 gression coefficients. This simply follows because in this case q and cash flow are perfectly
 65 collinear. With two shocks—a temporary shock and a persistent shock—the one-to-one
 66 relation between q and investment breaks down because investment is driven by produc-
 67 tivity in periods t and $t + 1$ while q responds to all future values of productivity. Finally,
 68 a “news shock” is introduced, that allows agents to observe the realization of future pro-
 69 ductivity shocks J periods in advance. Increasing the length of the horizon J reduces the
 70 coefficient on q and increases the coefficient on cash flow in investment regressions. This
 71 is due again to the differential responses of investment and q to information on produc-
 72 tivity at different horizons.

73 The model with no adjustment costs, while analytically tractable, is quantitatively un-
 74 appealing, as it tends to produce too much short-run volatility and too little persistence
 75 in investment. Therefore, for a more quantitative evaluation of the model we introduce
 76 adjustment costs. The model is calibrated to data moments from Compustat and analyze
 77 its implications both in terms of impulse responses and in terms of investment regres-
 78 sions. The baseline calibration is based on the two shocks structure, with temporary and
 79 persistent shocks. In this calibration q responds relatively more strongly to the persistent
 80 shock while investment responds relatively more strongly to the transitory shock, in line
 81 with the intuition from the no-adjustment-cost case. This leads to investment regressions
 82 with a smaller coefficient on q and a larger coefficient on cash flow, relative to a model
 83 with no financial frictions, thus bringing us closer to empirical coefficients. However, the
 84 q coefficient is still larger than in the data and the cash flow coefficient is smaller than
 85 in the data. When adding the possibility of news shocks, the disconnect between q and
 86 investment increases, leading to further reductions in the q coefficient and increases in the
 87 cash flow coefficient.

88 Fazzari et al. (1988) started a large empirical literature that explores the relation be-
 89 tween investment and q using firm-level data. The typical finding in this literature is a

²The terminology goes back to Hayashi (1982), who shows that the two are equivalent in a canonical model with convex adjustment costs.

90 small coefficient on q and a positive and significant coefficient on cash flow.³ Fazzari et al.
 91 (1988), Gilchrist and Himmelberg (1995) and most of the subsequent literature interpret
 92 these findings as a symptom of financial frictions at work. More recent work by Gomes
 93 (2001) and Cooper and Ejarque (2003) questions this interpretation. The approach taken
 94 in these two papers is to look at the statistical implications of simulated data generated by
 95 a model to understand the empirical correlations between investment, q and cash flow.⁴
 96 In their simulated economies with financial frictions q still explains most of the variability
 97 in investment, and cash flow does not provide additional explanatory power. In this pa-
 98 per, we take a similar approach but reach different conclusions. This is due to two main
 99 differences. First, Gomes (2001) and Cooper and Ejarque (2003) model financial frictions
 100 by introducing a transaction cost which is a function of the flow of outside finance issued
 101 each period, while we introduce a contractual imperfection that imposes an upper bound
 102 on the stock of outside liabilities as a fraction of total assets. Our approach adds a state
 103 variable to the problem, namely the stock of existing liabilities of the firm as a fraction
 104 of assets, thus generating slower dynamics in the gap between internal funds and the
 105 desired level of investment. Second, we explore a variety of shock structures, which, as
 106 argued below, play an important role in our results.

107 A related strand of recent literature has focused on violations of q theory coming from
 108 decreasing returns or market power, leaving aside financial frictions.⁵ Our effort is com-
 109plementary to this literature, since both financial frictions and decreasing returns deter-
 110mine the presence of future rents embedded in the value of the firm. Also in that literature
 111the shock structure plays an important role in the results. For example, Eberly et al. (2008)
 112show that it is easier to obtain realistic implications for investment regressions by assum-
 113ing a Markov process in which the distribution from which persistent productivity shocks
 114are drawn switches occasionally between two regimes. Abel and Eberly (2011) also show
 115that in models with decreasing returns it is possible to obtain interesting dynamics in q
 116with no adjustment costs, similarly to the results presented in Section 3 for a model with
 117constant returns to scale and financial constraints.

118 The simplest shock that breaks the link between q and investment in models with
 119financial constraints is a purely temporary shock to cash flow, which does not affect cap-
 120ital's future productivity. Absent financial frictions this shock should have no effect on
 121current investment. This idea is the basis of a strand of empirical literature that tests for
 122financial constraints by identifying some source of purely temporary shocks to cash flow.

³See Hubbard (1998) for a survey.

⁴An approach that goes back to Sargent (1980).

⁵See Schiantarelli and Georgoutsos (1990), Altı (2003), Moyen (2004), Eberly et al. (2008), Abel and Eberly (2011), Abel and Eberly (2012).

123 This is the approach taken by Blanchard et al. (1994) and Rauh (2006), which provide
124 reliable evidence of the presence of financial constraints. Our paper builds on a similar
125 intuition, by showing that in general shocks affecting profitability at different horizons
126 have differential effects on q and investment and asks whether, given a realistic mix of
127 shocks, a model with financial frictions can produce the unconditional correlations ob-
128 served in the data.

129 This paper uses the simplest possible model with the features needed: an occasionally
130 binding financial constraint; a dynamic, stochastic structure; adjustment costs that can
131 produce realistic investment dynamics. There is a growing literature that builds richer
132 models that are geared more directly to estimation. In particular, Hennessy and Whited
133 (2007) build a rich structural model of firms' investment with financial frictions, which
134 is estimated by simulated method of moments. They find that the financial constraint
135 plays an important role in explaining observed firms' behavior. In their model, due to the
136 complexity of the estimation task, the financial friction is introduced in a reduced form
137 manner, by assuming transaction costs associated to the issuance of new equity or debt, as
138 in Gomes (2001) or Cooper and Ejarque (2003).⁶ This paper takes a complementary route,
139 as it features a more stylized model, but financial constraints coming from an explicitly
140 modeled contractual imperfection.

141 A growing number of papers uses recursive methods to characterize optimal dynamic
142 financial contracts in environments with different forms of contractual frictions (Atkeson
143 and Cole (2005), Clementi and Hopenhayn (2006), DeMarzo and Sannikov (2006), De-
144 Marzo et al. (2012)). The limited enforcement friction in this paper makes it closer to the
145 models in Albuquerque and Hopenhayn (2004) and Cooley et al. (2004). Within this liter-
146 ature Biais et al. (2007) look more closely at the implications of the theory for asset pricing.
147 In particular, they find a set of securities that implements the optimal contract and then
148 study the stochastic behavior of the prices of these securities. Here, our objective is to ex-
149 amine the model's implication for q theory, therefore we simply focus on the total value
150 of the firm, which includes the value of all the claims held by insiders and outsiders.

151 Section 2 presents the model. Section 3 contains the case of no adjustment costs. Sec-
152 tion 4 contains the model with adjustment costs.

⁶The difference in results, relative to these papers, appears due to the fact that Hennessy and Whited (2007) also match the behavior of a number of financial variables.

2 The model

Consider an infinite horizon economy, in discrete time, populated by a continuum of entrepreneurs who invest in physical capital and raise funds from risk neutral investors.

The entrepreneurs' technology is linear: K_{it} units of capital, installed at time $t - 1$ by entrepreneur i , yield profits $A_{it}K_{it}$ at time t . We can think of the linear profit function $A_{it}K_{it}$ as coming from a constant returns to scale production function in capital and other variable inputs which can be costlessly adjusted. Therefore, changes in A_{it} capture both changes in technology and changes in input and output prices. For brevity, we just call A_{it} "productivity". Productivity is a function of the state s_{it} , $A(s_{it})$, where s_{it} is a Markov process with a finite state space \mathbf{S} and transition probability $\pi(s_{it}|s_{it-1})$. There are no aggregate shocks, so the cross sectional distribution of s_{it} across entrepreneurs is constant.

Investment is subject to convex adjustment costs. The cost of changing the installed capital stock from K_{it} to K_{it+1} is $G(K_{it+1}, K_{it}; s_{it})$ units of consumption goods at date t . The function G includes both the cost of purchasing capital goods and the installation cost. G is increasing and convex in its first argument, decreasing in the second argument, and displays constant returns to scale. For numerical results, we use the quadratic functional form

$$G(K_{it+1}, K_{it}; s_{it}) = \phi(s_{it}) (K_{it+1} - (1 - \delta(s_{it})) K_{it}) + \frac{\zeta}{2} \frac{(K_{it+1} - K_{it})^2}{K_{it}}, \quad (1)$$

in which the state s_{it} can affect both the depreciation rate $\delta(s_{it})$ and the price of capital goods $\phi(s_{it})$.

All agents in the model are risk neutral. The entrepreneurs' discount factor is β and the investors' discount factor is $\hat{\beta}$ and entrepreneurs are more impatient: $\beta < \hat{\beta}$. Investors have a large enough endowment of the consumption good each period so that in equilibrium the interest rate is $1 + r_t = 1/\hat{\beta}$. Each period an entrepreneur retires with probability γ and is replaced by a new entrepreneur with an endowment of 1 unit of capital. When an entrepreneur retires, productivity A_{it} is zero from next period on. The retirement shock is embedded in the process s_{it} by assuming that there is an absorbing state s^r with $A(s^r) = 0$ and the probability of transitioning to s^r from any other state is γ .

Each period, entrepreneur i issues one-period state contingent liabilities, subject to limited enforcement. The entrepreneur controls the firm's capital K_{it} and, at the beginning of each period, can default on his liabilities and divert a fraction $1 - \theta$ of the firm's capital. If he does so, he re-enters the financial market as a new entrepreneur, with capital $(1 - \theta) K_{it}$ and no liabilities. That is, the punishment for a defaulting entrepreneur is the loss of a fraction θ of the firm's assets.

186 2.1 Optimal investment

187 Let us formulate the optimization problem of the individual entrepreneur in recursive
 188 form, dropping the subscripts i and t . Let $V(K, B, s)$ be the expected utility of an en-
 189 trepreneur in state s , who enters the period with capital stock K and current liabilities B .
 190 For now, simply assume that the problem's parameters are such that the entrepreneur's
 191 optimization problem is well defined. In the following sections, we provide conditions
 192 that ensure that this is the case.⁷ The function V satisfies the Bellman equation

$$V(K, B, s) = \max_{C \geq 0, K' \geq 0, \{B'(s')\}} C + \beta \mathbb{E} [V(K', B'(s'), s') | s], \quad (2)$$

subject to

$$C + G(K', K; s) \leq A(s)K - B + \hat{\beta} \mathbb{E} [B'(s') | s], \quad (3)$$

$$V(K', B'(s'), s') \geq V((1 - \theta)K', 0, s'), \forall s', \quad (4)$$

193 where C is current consumption, K' is next period's capital stock, and $B'(s')$ are next pe-
 194 riod's liabilities contingent on s' . Constraint (3) is the budget constraint and $\hat{\beta} \mathbb{E} [B'(s') | s]$
 195 are the funds raised by selling the state contingent claims $\{B'(s')\}$ to the investors. Con-
 196 straint (4) is the enforcement constraint that requires the continuation value under repay-
 197 ment to be greater than or equal to the continuation value under default.

198 The assumption of constant returns to scale implies that the value function takes the
 199 form $V(K, B, s) = v(b, s)K$ for some function v , where $b = B/K$ is the ratio of current
 200 liabilities to the capital stock. The Bellman equation then becomes, using the notation
 201 $c = C/K$ and $k' = K'/K$,

$$v(b, s) = \max_{c \geq 0, k' \geq 0, \{b'(s')\}} c + \beta \mathbb{E} [v(b'(s'), s') | s] k', \quad (5)$$

subject to

$$c + G(k', 1; s) \leq A(s) - b + \hat{\beta} \mathbb{E} [b'(s') | s] k', \quad (6)$$

$$v(b'(s'), s') \geq (1 - \theta) v(0, s'), \forall s'. \quad (7)$$

202 It is easy to show that v is strictly decreasing in b . We can then find state-contingent

⁷In the online appendix we provide a general existence result.

203 borrowing limits $\bar{b}(s')$ such that the enforcement constraint is equivalent to

$$b'(s') \leq \bar{b}(s'), \forall s'. \quad (8)$$

204 So the enforcement constraint is equivalent to a state contingent upper bound on the
 205 ratio of the firm's liabilities to capital. Relative to existing models with collateral con-
 206 straints, two distinguishing features of this model are the presence of state-contingent
 207 claims and the fact that state-contingent bounds are derived endogenously from limited
 208 enforcement.⁸

209 2.2 Average and marginal q

210 To characterize the solution to the entrepreneur's problem let us start from the first order
 211 condition for k' :

$$\lambda G_1(k', 1; s) = \lambda \hat{\beta} \mathbb{E}[b'|s] + \beta \mathbb{E}[v'|s], \quad (9)$$

212 where λ is the Lagrange multiplier on the budget constraint (6), or the marginal value
 213 of wealth for the entrepreneur. The expressions $\mathbb{E}[b'|s]$ and $\mathbb{E}[v'|s]$ are shorthand for
 214 $\mathbb{E}[b'(s')|s]$ and $\mathbb{E}[v(b'(s'), s')|s]$. Optimality for consumption implies that $\lambda \geq 1$ and
 215 the non-negativity constraint on consumption is binding if $\lambda > 1$.

216 To interpret condition (9) rewrite it as:

$$\lambda = \frac{\beta \mathbb{E}[v'|s]}{G_1(k', 1; s) - \hat{\beta} \mathbb{E}[b'|s]} \geq 1. \quad (10)$$

217 When the inequality is strict the entrepreneur strictly prefers reducing current consump-
 218 tion to invest in new units of capital. If C was positive the entrepreneur could reduce it
 219 and use the additional funds to increase the capital stock. The marginal cost of an extra
 220 unit of capital is $G_1(k', 1; s)$ but the extra unit of capital increases collateral and allows
 221 the entrepreneur to borrow $\hat{\beta} \mathbb{E}[b'|s]$ more from the consumers. So a unit reduction in
 222 consumption leads to a levered increase in capital invested of $1/(G_1 - \hat{\beta} \mathbb{E}[b'|s])$. Since
 223 capital tomorrow increases future utility by $\beta \mathbb{E}[v'|s]$, we obtain (10).

224 Condition (9) can be used to derive our main result on average and marginal q . The
 225 value of all the claims on the firm's future earnings, held by investors and by the en-

⁸Other recent models that allow for state-contingent claims include He and Krishnamurthy (2013) and Rampini and Viswanathan (2013). Cao (2018) develops a general model with an explicit stochastic structure that studies collateral constraints with non-state-contingent debt.

226 entrepreneur at the end of the period, is

$$\hat{\beta}\mathbb{E} [B' (s') |s] + \beta\mathbb{E} [V (K', B' (s'), s') |s].$$

227 Dividing by total capital invested gives us average q :

$$q^a \equiv \hat{\beta}\mathbb{E} [b'|s] + \beta\mathbb{E} [v'|s].$$

228 Marginal q , on the other hand, is just the marginal cost of one unit of new capital, $q^m \equiv$
229 $G_1 (k', 1; s)$. Rearrange equation (9) and express it in terms of q^a and q^m to get:

$$q^a = q^m + \frac{\lambda - 1}{\lambda}\beta\mathbb{E} [v'|s]. \quad (11)$$

230 Since $\lambda > 1$ if and only if the non-negativity constraint on consumption is binding, we
231 have proved the following result.

232 **Proposition 1.** *Average q is greater than or equal to marginal q , with strict equality if and only*
233 *if the non-negativity constraint on consumption is binding.*

234 Equation (11) also shows that the difference between average and marginal q is in-
235 creasing in the Lagrange multiplier λ and in the future value of entrepreneurial equity
236 $\mathbb{E} [v'|s]$ (if $\lambda > 1$). As we shall see in the numerical part of the paper, an increase in
237 indebtedness b increases λ but reduces the future value of entrepreneurial equity, so in
238 general the relation between b and $q^a - q^m$ can be non-monotone. There is a cutoff for
239 b such that $\lambda = 1$ below the cutoff and $\lambda > 1$ above the cutoff, so the relation must be
240 increasing in some region.

241 The first order condition for b' can be written as

$$\hat{\beta}\lambda + \beta v_b (b' (s'), s') = \mu(s'),$$

242 where $\pi(s'|s)\mu(s')k'$ is the Lagrange multiplier on the debt constraint (8). Using the en-
243 velope condition for b to substitute for v_b and using time subscripts, write

$$\lambda_t = \frac{\beta}{\hat{\beta}}\lambda_{t+1} + \frac{1}{\hat{\beta}}\mu_{t+1}. \quad (12)$$

244 This condition shows that λ_t is a forward looking variable determined by current and
245 future values of μ_{t+1} . Positive values of this Lagrange multiplier in the future induce
246 the entrepreneur to reduce consumption today to increase internal funds available. The

247 forward looking nature of λ_t will be useful to interpret some of our numerical results
 248 about news shocks.

249 Now one can see the role of our assumption $\beta < \hat{\beta}$. If we had $\hat{\beta} = \beta$, condition
 250 (12) would imply that if, at some date t , the entrepreneur's consumption is positive and
 251 $\lambda_t = 1$, then the non-negativity constraint and the collateral constraint can not be binding
 252 at any future date. In other words, once the entrepreneur is unconstrained he can never
 253 go back to being constrained. This is due to the assumption of complete state contingent
 254 markets. Assuming $\beta < \hat{\beta}$ ensures that entrepreneurs alternate between periods in which
 255 they are constrained and periods in which they are unconstrained.

256 To conclude this section, let us introduce some asset pricing relations that charac-
 257 terize the equilibrium. The notation $G_{1,t}$ and $G_{2,t}$ is shorthand for $G_1(K_{t+1}, K_t; s_t)$ and
 258 $G_2(K_{t+1}, K_t; s_t)$.

259 **Proposition 2.** *The following conditions hold in equilibrium*

$$\lambda_t = \beta \mathbb{E}_t \left[\lambda_{t+1} \frac{A_{t+1} - G_{2,t+1} - b_{t+1}}{G_{1,t} - \hat{\beta} \mathbb{E}_t b_{t+1}} \right], \quad (13)$$

260 and

$$\hat{\beta} \mathbb{E}_t \left[\frac{A_{t+1} - G_{2,t+1}}{G_{1,t}} \right] \geq 1 \geq \mathbb{E}_t \left[\frac{\beta \lambda_{t+1} A_{t+1} - G_{2,t+1}}{\lambda_t G_{1,t}} \right]. \quad (14)$$

261 *The last two conditions hold with strict inequality if the collateral constraint is binding with*
 262 *positive probability.*

263 The ratio

$$\frac{A_{t+1} - G_{2,t+1} - b_{t+1}}{G_{1,t} - \hat{\beta} \mathbb{E}_t b_{t+1}}$$

264 represents the levered rate of return on capital. Condition (13) further illustrates the
 265 forward-looking nature of λ_t . In particular, it shows that λ_t is a geometric cumulate of
 266 all future levered returns on capital. Condition (13) can also be interpreted as a stan-
 267 dard asset pricing condition, dividing both sides by λ_t and observing that $\beta \lambda_{t+1} / \lambda_t$ is the
 268 stochastic discount factor of the entrepreneur.

269 The expression

$$\frac{A_{t+1} - G_{2,t+1}}{G_{1,t}}$$

270 is the unlevered return on capital. When the collateral constraint is binding the first in-
 271 equality in (14) is strict and this implies that the expected rate of return on capital is
 272 higher than the interest rate $1 + r$. This implies that the levered return on capital is higher
 273 than the unlevered return. The entrepreneurs will borrow up to the point at which the

274 discounted levered rate of return is 1, by condition (13). At that point the discounted
 275 unlevered return will be smaller than 1, by the second inequality in (14).

276 Define the finance premium as the difference between the expected return on en-
 277 trepreneurial capital and the interest rate:

$$fp_t \equiv \mathbb{E}_t \left[\frac{A_{t+1} - G_{2,t+1}}{G_{1,t}} \right] - (1 + r). \quad (15)$$

278 The first inequality in (14) shows that the finance premium is positive whenever the collat-
 279 eral constraint is binding. This definition of the finance premium is used in an alternative
 280 calibration in the online appendix.

281 3 No adjustment costs

282 This section considers the case of zero adjustment costs, that is $\zeta = 0$ in equation (1). In
 283 this case, analytical results can be derived that map directly the shock structure into the
 284 coefficients of the investment regression.

285 With zero adjustment costs, the value function is linear

$$V(K, B, s) = \Lambda(s) [R(s)K - B], \quad (16)$$

286 where R is the gross return on capital defined by

$$R(s) \equiv A(s) + \phi(s)(1 - \delta(s)).$$

287 With a linear value function the borrowing limits are simply

$$\bar{b}(s) = \theta R(s), \quad (17)$$

288 and they have a natural interpretation: the entrepreneur can pledge a fraction θ of the
 289 firm's gross returns.

We now make assumptions that ensure that the problem is well defined and that the collateral constraint is always binding in equilibrium. Assume the following three in-

equalities hold for all s :

$$\beta \mathbb{E} [R (s') | s] > 1, \quad (18)$$

$$\theta \hat{\beta} \mathbb{E} [R (s') | s] < 1, \quad (19)$$

$$\frac{(1 - \gamma) (1 - \theta) \beta \mathbb{E} [R (s') | s, s' \neq s^r]}{\phi(s) - \theta \hat{\beta} \mathbb{E} [R (s') | s]} \leq \zeta, \quad (20)$$

290 for some scalar $\zeta < 1$. Condition (18) implies that the expected rate of return on capi-
 291 tal, discounted using entrepreneur's discount factor, is greater than 1, so entrepreneurs
 292 prefer investment to consumption. Condition (19) implies that pledgeable returns are in-
 293 sufficient to finance the purchase of one unit of capital, i.e., investment cannot be fully
 294 financed with outside funds. This condition ensures that investment is finite. Finally,
 295 condition (20) ensures that the entrepreneurs' utility is bounded. The last condition al-
 296 lows us to use the contraction mapping theorem to characterize the equilibrium marginal
 297 value of wealth $\Lambda (s)$ in the following proposition. The proof of this lemma and of the
 298 following results in this section are in the online appendix.

299 **Lemma 1.** *If conditions (18)-(20) hold there is a unique function $\Lambda : \mathbf{S} \rightarrow [1, \infty)$ that satisfies*
 300 *the recursion*

$$\Lambda (s) = \frac{\beta (1 - \theta) \mathbb{E} [\Lambda (s') R (s') | s]}{\phi(s) - \theta \hat{\beta} \mathbb{E} [R (s') | s]}, \text{ for all } s \neq s^r, \quad (21)$$

301 and $\Lambda (s) = 1$ for $s = s^r$.

302 Condition (21) is a special case of condition (13), in which the constraint is always
 303 binding. The following proposition characterizes an equilibrium.

304 **Proposition 3.** *If conditions (18)-(20) hold and $\Lambda (s)$ satisfies*

$$\Lambda (s) > \frac{\beta}{\hat{\beta}} \Lambda (s'), \quad (22)$$

305 for all $s, s' \in \mathbf{S}$, then the collateral constraint is binding in all states, consumption is zero until
 306 the retirement shock, investment in all periods before retirement is given by

$$\frac{K' - (1 - \delta(s)) K}{K} = \frac{(1 - \theta) R(s)}{\phi(s) - \theta \hat{\beta} \sum_{s'} \pi(s'|s) R (s')} - (1 - \delta(s)), \quad (23)$$

307 and average q is

$$q^a = \mathbb{E} [((1 - \theta) \beta \Lambda (s') + \theta \hat{\beta}) R (s') | s]. \quad (24)$$

308 Condition (22) ensures that entrepreneurs never delay investment. Namely, it implies

309 that they always prefer to invest in physical capital today rather than buying a state-
310 contingent security that pays in some future state.

311 The entrepreneur's problem can be analyzed under weaker versions of (18)-(22), but
312 then the constraint will be non-binding in some states. It is useful to remark that we
313 could embed our model in a general equilibrium environment with a constant returns to
314 scale production function in capital and labor and a fixed supply of labor. In this general
315 equilibrium model $A(s)$ is replaced by the endogenous value of the marginal product of
316 capital. It is then possible to derive conditions (18)-(22) endogenously if shocks are small
317 and the non-stochastic steady state features a binding collateral constraint.

318 Assume now that conditions (18)-(22) hold and let us analyze the model by lineariz-
319 ing the equilibrium conditions (21), (23) and (24) around the non-stochastic steady state.
320 Steady state values are denoted by a bar. A tilde denotes deviations from the steady state,
321 in levels or logs depending on the variable. Namely, level deviations are used for the
322 following variables that are already expressed as ratios: q_t^a , A_t (profits to assets), and the
323 investment rate, or investment to assets ratio,

$$IK_t = \frac{K_{t+1} - (1 - \delta_t) K_t}{K_t}.$$

So, for example, $\tilde{q}_t^a = q_t^a - \bar{q}^a$. Log deviations are used for the variables $\Lambda_t, \delta_t, \phi_t$. So for
example, $\tilde{\Lambda}_t = \log \Lambda_t - \log \bar{\Lambda}$. Finally, for R_t , the approximation is

$$\tilde{R}_t = \tilde{A}_t + \tilde{\phi}_t(1 - \bar{\delta}) - \bar{\delta}\tilde{\delta}_t.$$

324 The steady state price of capital is normalized to $\bar{\phi} = 1$.

325 The following proposition characterizes the dynamics of investment and average q
326 around the steady state.

Proposition 4. *If the economy satisfies (18)-(22) a linear approximation gives the following ex-
pressions for investment, average q and the marginal utility of entrepreneurial wealth Λ_t :*

$$\tilde{IK}_t = \frac{1 - \theta}{1 - \theta\hat{\beta}\bar{R}} \tilde{R}_t + \frac{(1 - \theta)\bar{R}}{1 - \theta\hat{\beta}\bar{R}} \frac{\mathbb{E}_t[\theta\hat{\beta}\tilde{R}_{t+1}] - \tilde{\phi}_t}{1 - \theta\hat{\beta}\bar{R}} + \bar{\delta}\tilde{\delta}_t, \quad (25)$$

$$\tilde{q}_t^a = \mathbb{E}_t[(1 - \theta)\beta\tilde{\Lambda}_{t+1}\bar{\Lambda}\bar{R} + (1 - \theta)\beta((1 - \gamma)\bar{\Lambda} + \gamma)\tilde{R}_{t+1} + \theta\hat{\beta}\tilde{R}_{t+1}], \quad (26)$$

$$\tilde{\Lambda}_t = \frac{1}{1 - \theta\hat{\beta}\bar{R}} \sum_{j=0}^{\infty} \left(\frac{(1-\gamma)\bar{\Lambda}}{\gamma + (1-\gamma)\bar{\Lambda}} \right)^j \mathbb{E}_t \left[\tilde{R}_{t+j+1}/\bar{R} - \tilde{\phi}_{t+j} \right]. \quad (27)$$

327 Equations (25)-(26) express investment and average q in terms of current and future
 328 expected values of productivity. Given assumptions about the exogenous processes for
 329 A_t, ϕ_t, δ_t , equations (25) and (26) give us all the information about the variance-covariance
 330 matrix of $(\tilde{I}K_t, \tilde{q}_t^a, \tilde{A}_t)$ and thus about investment regression coefficients. In particular, we
 331 are interested in the implications of the model for the investment regression

$$IK_{it} = a_{i0} + a_1 q_{it}^a + a_2 CFK_{it} + e_{it}, \quad (28)$$

332 where CFK_{it} is the ratio of cash flow to assets, which is identified with A_{it} in our model.

333 We now turn to a battery of examples that show how different shock structures lead
 334 to different implications for the variance-covariance matrix of investment, average q and
 335 cash flow and thus for investment regressions.

336 3.1 Examples: productivity shocks

337 We begin with examples that only include productivity shocks.

338 **Example 1.** Productivity \tilde{A}_t follows the AR(1) process:

$$\tilde{A}_t = \rho \tilde{A}_{t-1} + \varepsilon_t,$$

339 where ε_t is an i.i.d. shock. There are no shocks to the price of capital and depreciation.

340 In this example, $\mathbb{E}_t [\tilde{A}_{t+j}] = \rho^j \tilde{A}_t$ so all future expected values of \tilde{A}_t are proportional
 341 to the current value. Substituting in (25)-(26), it is easy to show that both \tilde{q}_t^a and $\tilde{I}K_t$ are
 342 linear functions of \tilde{A}_t . Therefore, in this case cash flow and average q are both, separately,
 343 sufficient statistics for investment. This is true even though there is a financial constraint
 344 always binding, simply due to the fact that a single shock is driving both variables.

Example 2. Productivity \tilde{A}_t has a persistent component x_t and a temporary component
 η_t :

$$\begin{aligned} \tilde{A}_t &= x_t + \eta_t, \\ x_t &= \rho x_{t-1} + \varepsilon_t. \end{aligned}$$

345 There are no shocks to the price of capital and to depreciation.

346 In this example, we have $\mathbb{E}_t [\tilde{A}_{t+j}] = \rho^j x_t$, and substituting in (25)-(26), after some
347 algebra, we obtain

$$\tilde{I}\tilde{K}_t = \frac{(1-\theta)(1-(1-\rho)\bar{R}\hat{\beta})}{(1-\theta\hat{\beta}\bar{R})^2} x_t + \frac{1-\theta}{1-\theta\hat{\beta}\bar{R}} \eta_t,$$

348 and

$$\tilde{q}_t^a = \left[\beta(1-\theta)(\gamma + (1-\gamma)\bar{\Lambda}) + \theta\hat{\beta} + \frac{\beta(1-\theta)(1-\gamma)(\gamma + (1-\gamma)\bar{\Lambda})}{(1-\theta\hat{\beta}\bar{R})(\gamma + (1-\gamma)(1-\rho)\bar{\Lambda})} \bar{\Lambda}\rho \right] \rho x_t.$$

349 If we run a regression of investment on average q and cash flow, cash flow is the only
350 variable that can capture variations in η_t , so the coefficient on cash flow will be positive
351 and equal to

$$\frac{1-\theta}{1-\theta\hat{\beta}\bar{R}'}$$

352 and cash flow improves the explanatory power of the investment regression. The cru-
353 cial observation is that average q is affected by the marginal value of entrepreneurial net
354 worth, which is a forward looking variable that reflects expectations about all future ex-
355 cess returns on entrepreneurial capital.⁹ Through this channel, average q responds to
356 information about future values of A_t at all horizons. At the same time, investment is
357 only driven by the current and next period value of A_t . The current value determines
358 internal funds, the next period value determines collateral values. Putting these facts to-
359 gether implies that shocks that affect profitability differentially at different horizons can
360 break the link between average q and investment.

361 To get a quantitative sense of the model implications, let us use the parameter values
362 in the calibrated model of next section, summarized in the first two lines of Table 1 below.
363 However, unlike in that parametrization, let us set the parameter $\zeta = 0$ to zero (no ad-
364 justment costs) and calibrate the parameters $\bar{\delta}$ and γ to target the average values of q and
365 of the investment rate specified in the next section, which requires setting

$$\bar{\delta} = 0.092 \quad \text{and} \quad \gamma = 0.095.$$

366 The linearization above yields the following coefficients on Q and cash flow in the
367 investment regression:

$$a_1 = 0.0561 \quad \text{and} \quad a_2 = 1.0273.$$

⁹See the discussion following Proposition 2.

368 If we use as references the coefficients in Gilchrist and Himmelberg (1995) (0.033 and
 369 0.24), the coefficient on q is close to the empirical counter-part while the coefficient on
 370 cash flow is too high. With two shocks and two regressors, the R^2 of the regression is
 371 exactly 1.

372 Notice that in this example, investment, q and cash flow are fully determined by the
 373 two random variables x_t and η_t and the coefficients are independent of the variance pa-
 374 rameters. This implies that, given all the other parameters, the coefficients of the invest-
 375 ment regression are independent of the values of the variances σ_ε^2 and σ_η^2 , as long as both
 376 are positive. As we shall see, this result does not extend to the general model with adjust-
 377 ment costs.

378 As an aside, notice that in this example, the coefficient on cash flow is higher for firms
 379 with larger values of θ , i.e., for firms that can finance a larger fraction of investment with
 380 external funds. These firms respond more because they can lever more any temporary
 381 increase in internal funds. This is reminiscent of the observation in Kaplan and Zingales
 382 (1997) that the coefficient on cash flow in an investment regression should not be used as
 383 measure of the tightness of the financial constraint.

384 It is useful to remark that our microfoundation of the financial constraint matters for
 385 the results derived. In particular, equation (25) makes it clear that investment in our
 386 framework depends only on the future value of the firms' asset values at short horizons
 387 (R_{t+1}) and not on their value further in the future. This comes from the way we have
 388 formulated the participation constraint in (4), which allows the entrepreneur to re-enter
 389 financial markets after a default event. Other formulations may make future values of R_t
 390 enter in richer ways in current investment, through essentially a "franchise value" effect.
 391 It is possible that these forces could increase the correlation between investment and q ,
 392 but we leave this investigation to future research.

393 3.2 Examples: additional shocks

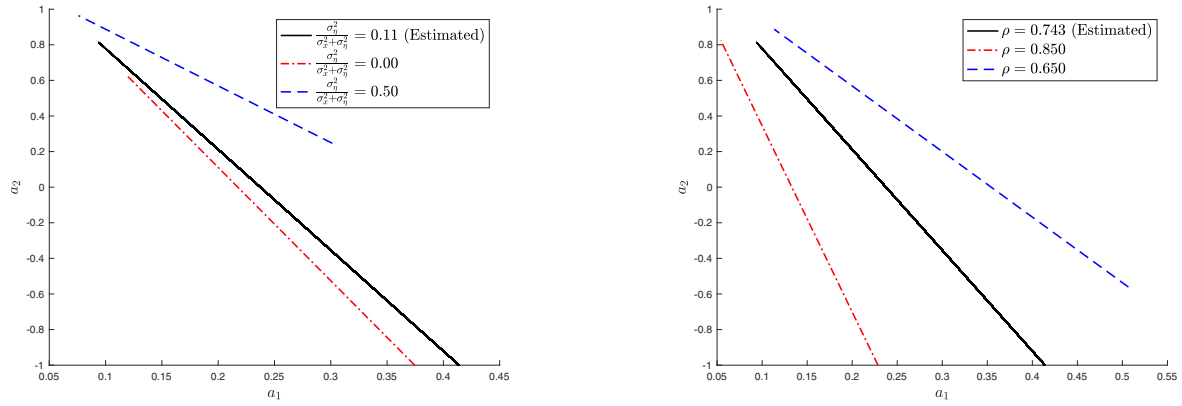
394 We now add shocks to the price of capital ϕ_t and to the depreciation rate δ_t and look at
 395 their quantitative implications for investment regressions. Let us begin with ϕ_t .

396 **Example 3.** The productivity process is as in example 2. The price of capital follows the
 397 AR(1) process

$$\tilde{\phi}_{t+1} = \rho_\phi \tilde{\phi}_t + v_{t+1}.$$

398 To choose a reasonable parametrization for this process, we borrow from the litera-
 399 ture on investment-specific technology shocks and with parameters taken from Justiniano

Figure 1: Effect of other shocks on investment regression



Note: Investment regression coefficients as the persistence and variance of shocks vary. The linear relations are described in Proposition 5.

400 et al. (2010):

$$\rho_\phi = 0.72 \quad \text{and} \quad \sigma_\nu = 0.063.$$

401 The coefficients on q and cash flow in the investment regression are now

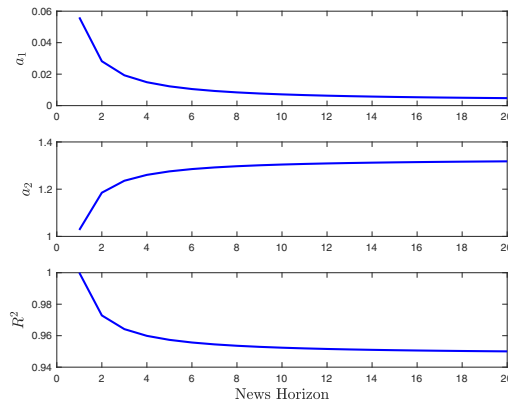
$$a_1 = 0.1238 \quad \text{and} \quad a_2 = 0.6434.$$

402 and R^2 is now 0.9525. So adding the price of capital shock reduces the coefficient on cash
 403 flow, but increases the coefficient on q . The intuition for this result is that the ϕ_t shock
 404 affects investment and q but does not affect cash flow. Therefore it tends to increase the
 405 coefficient on q and decrease the coefficient on cash flow in the investment regression.

406 We have experimented with adding an AR(1) process for the depreciation variable,
 407 alone and in combination with ϕ_t shocks, obtaining analogous results. The investment
 408 regression coefficients for a variety of parametrizations of the processes of productivity,
 409 ϕ_t and δ_t are reported in Figure 3.2. Notice that one fixes the productivity process, there is
 410 a linear relation between a_1 and a_2 . The processes for ϕ_t and δ_t determine a point in that
 411 linear relation, but not the position of the line. The latter is determined by the relative
 412 variance of the two productivity shocks (see the right panel) and by the persistence of the
 413 persistent component (see the left panel). This is an analytical result which relies on the
 414 fact that cash flow is not affected by the other shocks. The result is stated in the following
 415 proposition and proved in the online appendix.

416 **Proposition 5.** Suppose productivity follows the process in Example 2. Then the coefficients a_1, a_2

Figure 2: Effect of news shocks on investment regression



Note: Regression coefficients and R^2 as the news horizon increases

417 in the investment regression satisfy

$$\alpha_{q1}\sigma_x^2 a_1 + (\sigma_x^2 + \sigma_\eta^2)a_2 = \alpha_{i1}\sigma_x^2 + \alpha_{i2}\sigma_\eta^2, \quad (29)$$

418 for some coefficients α_{i1}, α_{i2} and α_{q1} that do not depend on the shock processes for ϕ_t and δ_t .

419 3.3 Examples: news shocks

420 We observed above that the presence of productivity shocks at different horizon alters the
 421 relation between q and investment. Building on this observation, we now introduce news
 422 shocks, that is shocks that reveal information about future profitability.

423 **Example 4.** The productivity process is as in Example 2 but the value of the permanent
 424 component x_t is known J periods in advance, with $J \geq 1$.

425 In the online appendix, we provide derivations for the dynamics of q and investment
 426 in this example and prove the following result.

427 **Proposition 6.** In the economy of Example 4, all else equal, increasing the horizon J at which
 428 shocks are anticipated decreases the coefficient on average q , increases the coefficient on cash flow,
 429 and reduces the R^2 of the investment regression.

430 The proof of this result is in the online appendix. Investment, as in the previous ex-
 431 ample, is just a linear function of productivity at times t and $t + 1$, which fully determine
 432 current cash flow and collateral values. On the other hand, q is a function of all future val-
 433 ues of A_t and, given the presence of news, these values are driven by anticipated future

434 shocks which have no effect on investment. This weakens the relation between q and in-
 435 vestment. Moreover, since q is the only source of information about x_{t+1} , and, with news
 436 shocks, it becomes a noisier source of information, this also reduces the joint explanatory
 437 power of q and cash flow.

438 Notice that news shocks here are acting very much like measurement error in q , by
 439 adding a shock to it that is unrelated to the shocks driving investment. However, financial
 440 frictions are essential in introducing this source of error. Absent financial frictions future
 441 values of productivity should not affect q , and it is only because q includes future quasi-
 442 rents that the relation arises.

443 To get a sense for the quantitative implications of new shocks, Figure 2 shows how the
 444 regression coefficients and R^2 change with the news horizon. Consistently with Propo-
 445 sition 6, as J increases, the coefficient on Q decreases and the coefficient on cash flow
 446 increases (starting from a relatively high value when there is no news), and R^2 decreases.

447 4 Adjustment costs

448 Let us now turn to the full model with adjustment costs and analyze its implications using
 449 numerical simulations. While the no adjustment cost model analyzed above is useful to
 450 build intuition, it has unrealistic implications for the responses of investment to shocks.
 451 In particular, it produces too large investment volatility for all plausible parametrizations.
 452 The model with adjustment costs, on the other hand, can be calibrated to match moments
 453 of the observed processes for profits and investment, so we can look at its quantitative
 454 implications.

455 First, our choice of parameters is presented and the equilibrium is characterized in
 456 terms of policy functions and impulse responses. We then run investment regressions on
 457 the simulated output and explore the model's ability to replicate empirical investment
 458 regressions.

459 In the baseline calibration, there are only productivity shocks. Shocks to the price of
 460 capital are added later.

461 4.1 Calibration

462 The time period in the model is one year. The baseline parameter values are summarized
 463 in Table 1. The first three parameters are pre-set, the remaining parameters are calibrated
 464 on Compustat data. We now describe their choice in detail.

465 The investors' discount factor $\hat{\beta}$ is chosen so that the implied interest rate is 8.7%. As

Table 1: Parameters

Preset	β	$\hat{\beta}$	θ	
	0.90	0.92	0.3	
Calibrated to cash flow moments	\bar{A}	ρ	σ_ε	σ_η
	0.246	0.743	0.0713	0.0375
Calibrated to investment and q moments	δ	ξ	γ	
	0.0250	1.75	0.095	

466 argued by Abel and Eberly (2011) the interest rate used in this type of exercises should
 467 correspond to a risk-adjusted expected return. The number chosen is in the range of rates
 468 of return used in the literature.¹⁰ The entrepreneurs' discount factor β has effects similar
 469 to the parameter γ which governs their exit rate. In particular, both affect the incentives
 470 of entrepreneurs to accumulate wealth and become financially unconstrained and both
 471 affect the forward looking component of q . Therefore, β is set at a level lower than $\hat{\beta}$ and
 472 γ is calibrated.¹¹ Regarding the fraction of non-divertible assets θ , there is only indirect
 473 empirical evidence, and existing simulations in the literature have used a wide range of
 474 values. Here $\theta = 0.3$ is chosen in line with evidence in Fazzari et al. (1988) and Nezafat
 475 and Slavic (2014). In particular, Fazzari et al. (1988) report that 30% of manufacturing
 476 investment is financed externally. Nezafat and Slavic (2014) use US Flow of Funds data
 477 for non-financial firms to estimate the ratio of funds raised in the market to finance fixed
 478 investment, and find a mean value of 0.284.

479 The parameters in the second line of Table 1 are calibrated to match moments of the
 480 firm-level cash flow time series in Compustat. Profits per unit of capital A_t are the sum
 481 of a persistent and a temporary component as in Example 2. Profits per unit of capital in
 482 the model, A_{it} , are identified with cash flow per unit of capital in the data, denoted by
 483 CFK_{it} .¹² The mean of A_t , denoted by \bar{A} , is set equal to average cash flow per unit of capital
 484 in the data. The values of ρ , σ_ε and σ_η are chosen to match the first and second order
 485 autocorrelation and the standard deviation of cash flow in the data, denoted, respectively,
 486 by $\rho_1(CFK)$, $\rho_2(CFK)$ and $\sigma(CFK)$. These moments are estimated using the approach of
 487 Arellano and Bond (1991) and Arellano and Bover (1995) and are reported in Table 1.¹³

¹⁰ Abel and Eberly (2011) and DeMarzo et al. (2012) choose numbers near 10%, while Moyen (2004) and Gomes (2001) use $r = 6.5\%$.

¹¹ Changing the chosen value of β in a reasonable range does not affect the results significantly.

¹² Cash flow is equal to net income before extraordinary items plus depreciation.

¹³ We estimate the firm-specific variation in cash-flow by first taking out the aggregate mean for each year and then applying the function `xtabond2` in STATA. This implements the GMM approach of Arellano and Bover (1995). This approach avoids estimating individual fixed effects affecting both the dependent variable (cash flow) and one of the independent variables (lagged cash flow), by first-differencing the law of motion for cash flow, and then using both lagged differences and lagged levels as instruments. We use

Table 2: Target moments and model values

Moment	$\rho_1(CFK)$	$\rho_2(CFK)$	$\sigma(CFK)$	$\mu(IK)$	$\sigma(IK)$	$\mu(q^a)$
Target value	0.60	0.41	0.113	0.17	0.111	2.5
Model value	0.60	0.41	0.113	0.23	0.098	2.5

488 Notice that simply computing raw autocorrelations in the data—as sometimes done in
 489 the literature—would lead to biased estimates, given the short sample length.¹⁴ In terms
 490 of sample, we use the same sub-sample of Compustat used in Gilchrist and Himmelberg
 491 (1995) so that we can compare our simulated regressions to their results.¹⁵

492 The next three parameters in Table 1, δ , ξ , and γ , are chosen to match three moments
 493 from the Compustat sample: the mean and standard deviation of the investment rate,
 494 $\mu(IK)$ and $\sigma(IK)$, and the mean of average q , $\mu(q^a)$. The reason why δ and ξ help deter-
 495 mine the level and volatility of the investment rate is intuitive, as these two parameters
 496 determine the depreciation rate and the slope of the adjustment cost function. The param-
 497 eter γ controls the speed at which entrepreneurs exit, so it affects the discounted present
 498 value of the quasi-rents they expect to receive in the future and thus average q . However,
 499 the three parameters interact, so we choose them jointly—by a grid search—in order to
 500 minimize the average squared percentage deviation between the three model-generated
 501 moments and their targets. The target moments from the data and the model generated
 502 moments are reported in Table 2.¹⁶

503 Notice that there is a tension between hitting the targets for $\mu(IK)$ and $\sigma(IK)$. Increas-
 504 ing any of the parameters, δ, ξ, γ reduces $\mu(IK)$, bringing it closer to its target value, but
 505 also decreases $\sigma(IK)$, bringing it farther from its target. Notice also that it is important
 506 for our purposes that the model generates a realistic level of volatility in the investment
 507 rate, given that IK is the dependent variable in the regressions we will present in Section
 508 4.3 below.

509 Our calibration also determines the average size of the wedge between average and
 510 marginal q . In particular, $\mu(q^a) = 2.5$ is the mean value of average q while ξ and $\mu(IK)$
 511 determine the mean value of marginal q , which is $1 + \xi(\mu(IK) - \delta) = 1.25$. Therefore, the
 512 average wedge between average and marginal q is 1.25. Since the presence of the wedge

the first three available (non-autocorrelated) lags in differences as instruments, with lags chosen separately
 for the 1st and 2nd order autocorrelation estimation. One lagged level is also used as an instrument.

¹⁴This type of bias was first documented in Nickell (1981). The bias is non-negligible in our sample. For
 the first-order autocorrelation, the Arellano and Bond (1991) approach gives $\rho_1(CFK) = 0.60$, while the raw
 autocorrelation in the data is 0.42.

¹⁵In particular, we restrict attention to the sample period 1978-1989 and use the same 428 listed firms
 used in their paper.

¹⁶The target standard deviation $\sigma(IK)$ is a pooled estimate.

Table 3: Calibration of frictionless model

Parameter	δ	ζ	γ	Moment	$\mu(IK)$	$\mu(q^a)$	$\sigma(IK)$
	0.05	1.50	0.125	Target value	0.17	2.5	0.111
				Model value	0.18	1.2	0.116

513 is what breaks the sufficient statistic property of q it is useful that our calibration imposes
 514 some discipline on the wedge's size.

515 All the simulations assume that entrepreneurs enter the economy with a unit endow-
 516 ment of capital and zero financial wealth (i.e., zero current profits and zero debt). Since
 517 the entrepreneurs' problem is invariant to the capital stock and all our empirical targets
 518 are normalized by total assets, the choice of the initial capital endowment is just a normal-
 519 ization. We have experimented with different initial conditions for financial wealth, but
 520 they have small effects on our results given that—with our parameters—the state variable
 521 b converges quickly to its stationary distribution.

522 It is useful to compare our results to those of a benchmark model with no financial
 523 frictions. To make the parametrization of the two models comparable, the parameters δ, ζ
 524 and γ are re-calibrated for the frictionless case. The moments and associated parameters
 525 are reported in Table 3. Notice that the frictionless model generates a low value of $\mu(q^a)$.
 526 For given IK , increasing ζ would increase marginal and average q (which are the same in
 527 the frictionless case), but it would reduce the volatility of investment.

528 4.2 Model dynamics

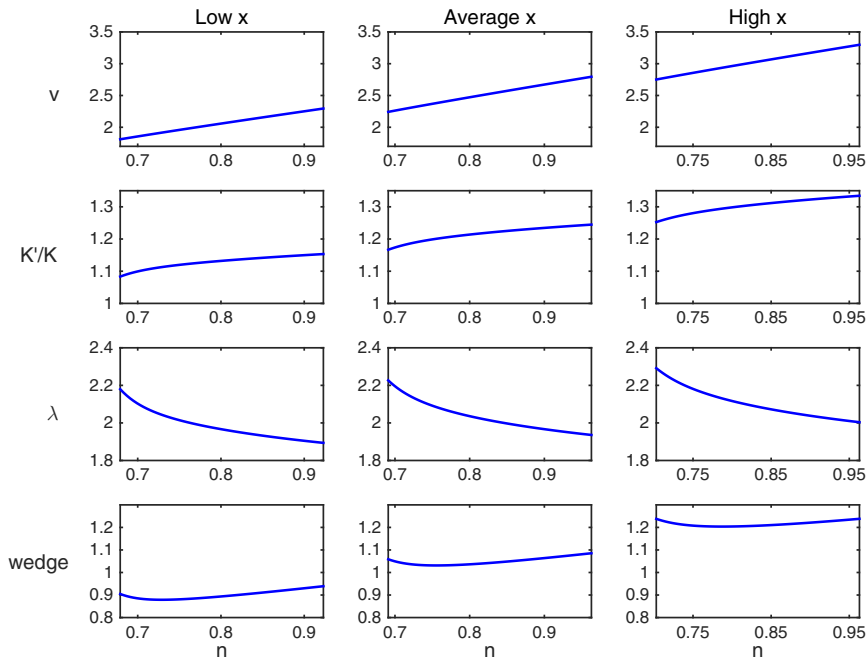
529 To describe the model behavior, it helps intuition to use as state variables A_t and n_t rather
 530 than A_t and b_t , where $n_t = A_t + \phi_t(1 - \delta_t) - b_t$, is the ratio of net worth (excluding
 531 adjustment costs) to assets K_t .

532 Each row of Figure 3 plots the value function (per unit of capital) v , the optimal in-
 533 vestment ratio K'/K , the Lagrange multiplier λ on the entrepreneur's budget constraint,
 534 and the wedge between average q and marginal q . Each column corresponds to different
 535 values of x_t . In particular, the values reported correspond to the the 20th, 50th and 80th
 536 percentile of the unconditional distribution of x . On the horizontal axis there is n , but the
 537 domain differs between columns as we plot values between the 10th to 90th percentile of
 538 the conditional distribution of n , conditional on the reported value of x .¹⁷

539 A higher level of n leads to a higher value v and a higher level of investment K'/K .
 540 Moreover, the value function is concave in n . The Lagrange multiplier λ is equal to the

¹⁷The joint distribution of (n, x) is computed numerically as the invariant joint distribution generated by the optimal policies.

Figure 3: Characterization of equilibrium



Note: The three columns correspond to the 20th, 50th, and 80th percentile of the persistent component of productivity x . The range for the net worth variable n is between the 10th and 90th percentiles of the distribution of n conditional on x .

541 derivative of the value function and therefore is decreasing in n . The fact that λ is de-
 542 creasing in n reflects the fact that a higher ratio of net worth to capital allows firms to
 543 invest more, leading to a higher shadow cost of capital G_1 and thus to a lower expected
 544 returns on investment. Eventually, for very high values of n we reach $\lambda = 1$. However, as
 545 the figures show this does not happen for the range of n values more frequently visited
 546 in equilibrium.

547 The bottom row documents how the wedge varies with the level of net worth n and
 548 with the persistent component of productivity x . Let us first look at the effect of n . Even
 549 though λ is decreasing in n , the wedge, $q^a - q^m$, does not vary much with n for a given
 550 value of x . Our analytical derivations in Section 2 help explain this outcome. Recall from
 551 equation (11) that the wedge is equal to

$$\frac{\lambda - 1}{\lambda} \beta \mathbb{E} [v' | s].$$

552 When we reach the unconstrained solution and $\lambda = 1$ the wedge disappears. However,
 553 for lower levels of n , for which the constraint is binding, the relation is in general non-

554 monotone. An increase in n reduces the marginal gain from an extra unit of net worth.
 555 However, at the same time it increases the future growth rate of firm's capital stock and
 556 so it increases the base to which this marginal quasi-rent is applied. This second effect
 557 is captured by the expression $\mathbb{E}[v'|s]$, because the value per unit of capital v' embeds the
 558 future growth of the firm and is increasing in n . The plots in the bottom row of Figure 3
 559 show that in the relevant range of n these two effects roughly cancel.

560 On the other hand, comparing the values of the wedge across columns, shows that
 561 persistent component of productivity x has large effects on the wedge and that the wedge
 562 is increasing in x . The reason is that higher values of x lead both to higher values of λ ,
 563 as the marginal benefits of extra internal funds increase with productivity, and to higher
 564 values of K'/K and v , because higher productivity allows the firm to raise more external
 565 funds and grow faster. Therefore both elements of the wedge increase with higher values
 566 of x .

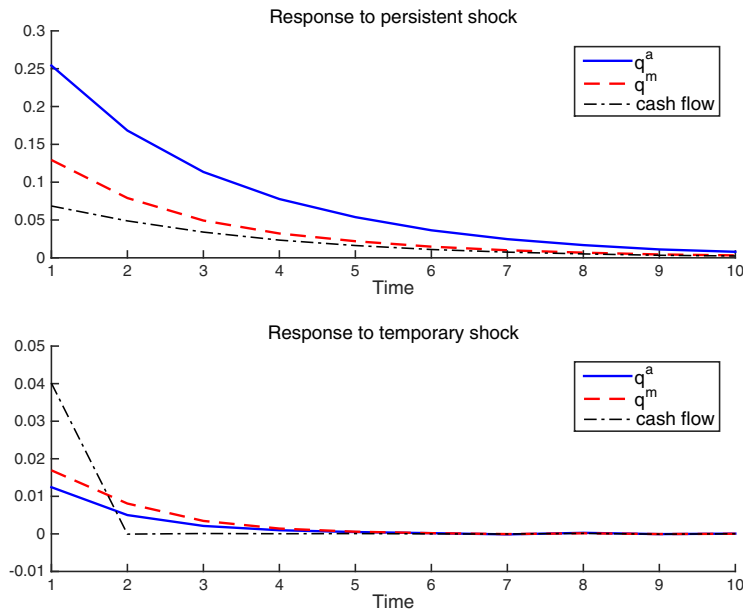
567 We now present impulse response functions following the two shocks. To construct
 568 these impulse response functions, we start at the median values of the state variables n
 569 and x . We then introduce a shock, simulate 10^6 paths following the shock, and report
 570 the difference between the average simulated paths, with and without the initial shock.
 571 Given the non-linearity of the model, the initial conditions for n and x in general affect
 572 the responses. However, in our simulations these non-linear effects are relatively small,
 573 so the plots below are representative.

574 The top panel of Figure 4 plots responses to a 1-standard-deviation persistent shock
 575 ε .¹⁸ Following a persistent shock all variables increase and return gradually to trend. The
 576 response of average q is larger than that of marginal q , thus producing an increase in the
 577 wedge. The bottom panel plots responses to the temporary shock η . Also in this case all
 578 three variables respond positively, but the response is more short-lived. Moreover, now
 579 the response of average q is slightly smaller than the response of marginal q , so the wedge
 580 shows a small decrease after the shock.

581 Average q is a forward-looking variable that incorporates the quasi-rents that the en-
 582 trepreneur is expected to receive in the future. These quasi-rents are only marginally
 583 affected by a temporary shock. In the model with no adjustment costs, the effect is zero.
 584 Here, because of adjustment costs, there is a positive effect, due to the fact that the invest-
 585 ment response displays a small but positive degree of persistence and high investment
 586 in the future increases the future value of installed capital. But the effect is small. In the
 587 case of a persistent shock, instead, future quasi-rents are directly affected by higher future

¹⁸The response of investment K'/K is always proportional to the response of marginal q and is thus omitted.

Figure 4: Impulse response functions



Note: Average paths following a shock at time 1, in (level) deviations from average paths following no shock. Cash flow is cash flow per unit of capital.

588 productivity, which will lead to faster growth (as shown in Figure 3), thus explaining the
 589 large increase in q^a in the top panel of Figure 4.

590 The discussion following Figure 3, helps to explain the response of the wedge $q^a - q^m$.
 591 A temporary shock leads to a pure increase in net worth per unit of capital. The effect of
 592 such an increase on the wedge is in general ambiguous and, with our parameter choices,
 593 close to zero. In the case of a persistent shock, instead, the effect is unambiguously to
 594 increase the wedge.

595 4.3 Investment regressions

596 We now turn to investment regressions, and ask whether the model can replicate the
 597 coefficients on q and cash flow observed in the data. In particular, we ask to what extent
 598 does the presence of a financial friction help to obtain a smaller coefficient on q and a
 599 positive and large coefficient on cash flow. To answer this question, simulated data are
 600 generated from our model and they are used to run the investment regression (28). In line
 601 with the empirical literature, we generate a balanced panel of 500 firms for 20 periods,

Table 4: Investment regressions

	a_1	a_2	R^2	Univariate q^a coefficient	R^2	Univariate CFK coefficient	R^2
Baseline model	0.22	0.14	0.98	0.26	0.98	0.81	0.89
Frictionless model	0.67	0.00	1.00	0.67	1.00	0.95	0.86
GH (1995)	0.033	0.24		0.05			

602 and allow for firm-level fixed effects in the regression.¹⁹ All reported results are the mean
603 values for 50 simulated panels.

604 The regression coefficients for the baseline model are presented in the first row of Ta-
605 ble 4. As reference points, the second row reports the coefficients that arise in the model
606 without financial frictions and in the last row the empirical estimates in Gilchrist and
607 Himmelberg (1995), which are representative of the orders of magnitude obtained in em-
608 pirical studies.²⁰ The table also reports coefficients of univariate regressions of investment
609 on average q and cash-flow separately.

610 The results for the frictionless benchmark are reported in the second line of Table 4.
611 In this case, average q is a sufficient statistic for investment, the coefficient on cash flow
612 is zero and the coefficient on q is equal to the inverse of the adjustment cost coefficient ζ ,
613 which is calibrated to 1.5. This line shows the standard empirical failure of the benchmark
614 adjustment cost model.

615 Adding financial frictions helps to get a smaller coefficient on q and a positive coeffi-
616 cient on cash flow. The effect is sizable, although the coefficient on q is still large compared
617 to the very small numbers found in empirical regressions. Notice also that the R^2 of the
618 regression is very close to 1. This is not surprising given the simple two-shock structure
619 and the presence of two explanatory variables.²¹ Given that the model is non-linear, the
620 R^2 can in general be smaller than 1. However, by experimenting with impulse responses
621 for different initial values of the state variables we have confirmed that, given our param-
622 eter values, the model is close to linear in its responses to the two shocks, which helps to
623 explain the high R^2 in Table 4.

624 The presence of the wedge breaks the one-to-one relation between q and investment
625 and allows for cash flow to have explanatory power in the the investment regression. In
626 particular, as can be seen in Figure 4 the wedge responds in opposite directions to the two
627 shocks, while q^m respond positively to both. So the wedge plays a role somewhat similar

¹⁹The model features random exit, so to generate a balanced panel we only keep firms for which exit does not occur for 20 periods.

²⁰We do not report standard errors, but they are small (less than 0.04) for both coefficients in our simulated data. They are also small in the empirical estimates of Gilchrist and Himmelberg (1995).

²¹For the same reason, in the linear model of Example 2, Section 3, the R^2 is 1.

Table 5: Investment regressions: changing shock variances

σ_ε	σ_η	σ_η^2/σ_A^2	a_1	a_2	R^2	Univariate q^a coefficient	R^2	Univariate CFK coefficient	R^2
0.113	0.000	0.00	0.38	-0.48	0.99	0.25	0.99	0.90	0.98
0.071	0.037	0.11	0.22	0.15	0.98	0.27	0.98	0.81	0.89
0.033	0.080	0.50	0.28	0.11	0.96	0.32	0.95	0.48	0.56
0.006	0.107	0.90	0.34	0.10	0.84	0.38	0.75	0.18	0.32
0.000	0.113	1.00	2.47	0.01	0.92	2.53	0.92	0.11	0.37

628 to measurement error in dampening the regression coefficient. Notice however that the
629 model still features a strong positive relation between q^a and investment, as documented
630 by the fifth and sixth columns of Table 4, which show that a univariate regression between
631 investment and average q produces a large coefficient and a large R^2 in simulated data
632 (unlike in actual data). In the rest of the paper we investigate shock structures that can
633 potentially weaken this relation.

634 It is useful to look at how the shock structure affects investment regressions. Table
635 5 reports regression coefficients and R^2 for different combinations of σ_ε and σ_η , keeping
636 constant the total volatility of A_t . The second row corresponds to the baseline case of
637 Table 4. The third column reports the fraction of variance due to the temporary shock.
638 Here all remaining parameters are kept at their baseline level, in order to focus on how
639 variance parameters affect the result.

640 The first row of Table 5 shows an extreme case with no temporary shocks. In this case,
641 the coefficient on q is larger than in our baseline and the coefficient on cash flow is actu-
642 ally negative. The last row of the table shows the opposite extreme, with only temporary
643 shocks. Interestingly, also this row displays a larger coefficient on q . The coefficient on
644 cash flow in this case is close to zero. So going to a one-shock model, worsens the model
645 performance in terms of replicating investment regressions. In this case q and investment
646 tend to comove simply because they are driven by the same shock. In these cases, we get
647 close to the sufficient statistic result obtained in the one-shock linear model of Example 1.
648 Example 1 has indeterminate implications for the coefficients, due to the perfect collinear-
649 ity of q and cash flow. Here, the perfect collinearity result does not hold for two reasons:
650 first, the model displays inertia so past values of x_t determine investment and q , which
651 complicates the correlation structure of investment, q and cash flow; second, the model is
652 non-linear. For these reasons, the bivariate coefficients are determinate even with a single
653 shock, and, in particular, the model prefers to assign a large coefficient on q .²²

²²The results in this table may help reconcile our results with the results of Gomes (2001). In particular, although Gomes (2001) uses a different model of financial frictions, it is possible that his result—that q is

Table 6: Investment regressions: shocks to the price of capital

σ_v	a_1	a_2	R^2	Univariate q^a		Univariate CFK	
				coefficient	R^2	coefficient	R^2
0.00	0.2212	0.1433	0.9837	0.2622	0.9799	0.8098	0.8883
0.01	0.2238	0.1351	0.9826	0.2622	0.9793	0.8132	0.8846
0.02	0.2309	0.1109	0.9798	0.2624	0.9775	0.8146	0.8740
0.03	0.2421	0.0741	0.9753	0.2626	0.9743	0.8166	0.8562
0.04	0.2561	0.0241	0.9695	0.2627	0.9694	0.8192	0.8314
0.05	0.2715	-0.0337	0.9632	0.2625	0.9630	0.8219	0.8002

654 The remaining rows of Table 5 illustrate intermediate cases in which both shocks are
655 present. As argued above, both shocks increase investment but they have opposite effects
656 on the wedge and that is what reduces the predictive power of q . So there is some inter-
657 mediate mix of shocks that adds maximum noise to the information contained in average
658 q and reduces the overall explanatory power of the investment regression. In the table
659 this is visible in the non-monotone relation between the ratio σ_η^2/σ_A^2 and the R^2 of the
660 regression.

661 While it is intuitive that mixing the two shocks reduces the total explanatory power
662 of investment regressions and reduces R^2 , the quantitative effects on the two coefficients
663 a_1 and a_2 are more complex to interpret, as they also depend on the magnitudes of the re-
664 sponses of investment, cash flow, and q to the underlying shocks. In particular, persistent
665 shocks tend to affect more, in relative terms, q than investment, due to the forward look-
666 ing nature of q and the presence of the financial constraint which dampens the response
667 of investment (see Figure 4).²³ Persistent shocks lead to a smaller response of investment
668 for a given response of q , when compared to temporary shocks. This is immediately vis-
669 ible in the monotone increase in the univariate coefficient with σ_η^2/σ_A^2 . The effect on the
670 bivariate coefficient a_1 is more complex as, at the same time, the presence of temporary
671 shocks increases the coefficient on cash flow. Therefore, the relation between each of the
672 coefficients a_1 and a_2 and the variance ratio σ_η^2/σ_A^2 is non-monotone.

673 The overall take out from Table 5 is that, given all other model parameters, the relative
674 variance of temporary and persistent shocks matter for both the explanatory power and
675 for the individual coefficients in investment regressions.

676 We can now add to the model additional shocks, as done in the case of no adjustment
677 costs in Section 3. In particular, we add the same AR1 process for the price of capital, with

almost a sufficient statistic for investment—could be driven by his one-shock structure.

²³The same two reasons identified above (inertia and non-linearity) for one-shock models, explain why in the two-shock model the relative size of the two variances matter for the regression coefficients, unlike in the simple linearized model with no adjustment costs of Section 3, Example 2.

Table 7: News shocks: calibration

	Parameters			Moments			
	δ	ξ	γ	$\mu(IK)$	$\frac{\sigma(IK)}{\sigma(CFK)}$	$\mu(q^a)$	$\sigma(q^a)$
Targets				0.17	0.98	2.5	0.97
No news (7 states)	0.0250	2.00	0.09	0.22	0.79	2.49	0.27
No news (2 states)	0.0200	2.00	0.10	0.22	0.94	2.24	0.33
$J = 1$	0.0275	3.00	0.08	0.21	0.86	2.39	0.42
$J = 2$	0.0225	3.50	0.08	0.20	0.85	2.24	0.45
$J = 3$	0.0225	3.50	0.08	0.19	0.91	2.67	0.59
$J = 4$	0.0275	3.50	0.08	0.19	0.95	2.48	0.59
$J = 5$	0.0300	3.50	0.08	0.19	0.97	2.50	0.63

678 parameters from Justiniano et al. (2010) as in Section 3.2. Table 6 reports the regression
679 results for different values of the variance of the price of capital shocks. As in the case
680 of no adjustment costs the coefficient on q increases and the coefficient on cash flow de-
681 creases. Quantitatively, the slope of the relation between a_1 and a_2 is of a similar order of
682 magnitude, but the relation is a bit flatter (i.e., the negative effect on the cash flow coef-
683 ficient is relatively smaller than the positive effect on the q coefficient) in the calibration
684 considered here. The underlying intuition is the same. Shocks to the cost of capital affect
685 investment and q but do not affect cash flow, so they weaken the relation between cash
686 flow and investment.

687 We now turn to news shocks. Example 4 in Section 3 shows that in the case of no
688 adjustment costs news shocks introduce additional noise in average q , thus reducing its
689 predictive power. Here we want to investigate whether the same forces are at work in our
690 full model with adjustment costs and see their quantitative implications.

691 Introducing news shocks increases the number of state variables, since we need to
692 keep track of anticipated values of x_t . Therefore, to simplify computations, we employ
693 a coarser description of the permanent component of the productivity process, using a
694 two-state Markov process for x_t . The stochastic process for A_t is specified and calibrated
695 as in our baseline but we assume agents observe x_t J periods in advance as in Example
696 4 in Section 3. We experiment with $J = 1, 2, \dots, 5$, re-calibrating the parameters δ , ξ and γ
697 for each value of J . Table 7 reports the calibrated parameters for each value of J . The table
698 also reports our baseline calibration (no news, 7 states) and a calibration with no news
699 and a 2 states Markov chain, which help to evaluate the effect of news on our results.

700 Table 7 shows that introducing news shocks improves the model's ability to match
701 the empirical level of the investment rate, reducing the value of $\mu(IK)$, while producing
702 similar values for $\sigma(IK)$ and $\mu(q^a)$. The table also reports the volatility of q^a (which is not

Table 8: Investment regressions: news shocks

	a_1	a_2	R^2
No news (7 states)	0.2047	0.1530	0.984
No news (2 states)	0.2434	0.0829	0.985
$J = 1$	0.1920	-0.0121	0.982
$J = 2$	0.1774	0.0161	0.974
$J = 3$	0.1417	0.0502	0.978
$J = 4$	0.1467	0.0628	0.976
$J = 5$	0.1394	0.0824	0.971

used as a target for our calibration), and the table shows that introducing news improves the model's realism in this dimension. The analytical derivations in Section 3 (Example 4) suggest a reason for this: anticipated shocks seem to introduce an additional source of volatility in q^a .

Turning to investment regressions, Table 8 shows regression coefficients and R^2 for different values of J . The coefficient on q^a and the R^2 behave in a similar way as suggested by Example 4: increasing the horizon adds noise in q^a thus reducing the coefficient and the overall R^2 . The cash flow coefficient goes down when going from no news to 1 period anticipation, and then increases monotonically in J .

Comparing the cases of no news and the case $J = 5$, the overall take away from Tables 7 and 8 is that news shocks improve the model's ability to match the observed behavior of investment, q and cash flow, both in terms of levels and volatility and in terms of the cross-correlations captured by investment regressions. The central intuition is that news shocks introduce a source of variation in q due to anticipated future shocks, which have little bearing on the contemporaneous movements in investment.

Due to the use of a 2 state Markov chain, the model with news does worse than the baseline in terms of the cash flow coefficient, so it is an open question for future work whether increasing the state space and possibly using alternative models of anticipated news that economize on state variables can further improve the model's empirical performance.²⁴

5 Conclusions

The paper shows that financial frictions can help dynamic investment models move closer to the correlations observed in the data. The model in this paper is stylized, but the main conclusions on the role of different shocks are likely to extend to more complex

²⁴See for example the information structure in Blanchard et al. (2013).

727 models. In particular, a promising avenue seems to be to build models where a substantial
728 fraction of the volatility in q is associated to news about profitability relatively far in the
729 future and where these news have relatively small effects on current investment decisions.
730 By assuming risk neutrality, we have omitted an important source of volatility in asset
731 prices, namely volatility in discount factors and risk premia. It is an important open
732 question how these additional sources of volatility affect the correlations investigated
733 here, especially because these factors are likely to correlate with the stringency of financial
734 constraints for individual firms.

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