# THE FACTOR BIAS OF TECHNICAL CHANGE UNDER PRODUCTION UNCERTAINTY

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#### **ABSTRACT**

The impact of production uncertainty on factor-augmenting technical change is examined by using and Itô stochastic control model. The results indicate that thechnical progress will be biased toward risk-reducing inputs and against risk-increasing inputs. This bias may reinforce or counteract the bias caused by relative input price movements.

Hicks [8] proposed that changes in the relative prices of inputs would act as a spur to inventions that economized the use of the factor that had become relatively more expensive. This intuitive proposition is now called the induced innovation theory. The theory has been formulated in microeconomic terms and tested empirically [1; 4; 5; 6; 9; 11; 12; 15; 18; 19; 20; 21].

The purpose of this paper is to show how the incorporation of production uncertainty alters Hicks's proposition. The inputs that increase the variance of output are separated from those that reduce the variance of output. The results show that, under production uncertainty, firms will deviate from the technology development and adoption path implied by Hicks's proposition so as to economize the use of inputs that increase the variance of production. This divergence will depend on the level of production uncertainty and on the

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degree of risk aversion of the decision maker. The results are derived by Itô stochastic control. The results depend on the specified assumptions; consequently, the results may not have immediate use in policy analysis. Nevertheless, the results are meaningful in that they may help to explain some of the deviations from Hicks's proposition that have been found in the empirical studies listed previously. Should it be possible to validate the theoretical results that follow, the concepts would help researchers develop innovations with a greater likelihood of acceptance in a target audience.

In the next section, the stochastic control model is presented. This model contains endogenous factor-augmenting technical change and allows the derivation of some propositions regarding the impact of production uncertainty on factoraugmenting technical change. The results are presented in terms of the implied deviations from the induced innovation theory. The final section contains concluding remarks and implications for policy.

#### I. The Model

A production function that explicitly incorporates factoraugmenting technical change may be written as

$$Y(t) = F(A(t) X_1(t), B(t) X_2(t)),$$

where F(.) is assumed to be a differentiable, quasi-concave function,  $X_1(t)$  and  $X_2(t)$  are the levels of two inputs at time t. and A(t) and B(t) reperesent the levels of factor augmentation at time t(19). It may be useful to consider  $X_1$  to be an imput that increases production uncertainty (e.g., fertilizer or yield increasing varieties) and  $X_2$  to be an input that decreases production uncertainty (e.g., irrigation or disease-resistant varieties).

Refer to Arnold (2), Hertzler (7), and Malliaris and Brock (13, 80-118) for a detailed discussion of Itô differentiation and dynamic programming.

It is assumed that a representative firm can fully internalize the benefits from its research and that it finances all research internally. This assumption limits the analysis to firms that can obtain patents or to government-sponsored research programs that conduct research on problems specific to the domestic environment. A less formal interpretation would allow us to view the research that ensues as that which would be desired by firms that ultimately use the proceeds of the research. If these firms signal researchers via market forces or the political process, then research programs should reflect the desires of these end users in a manner consistent with the assumption.

The levels of research expenditures  $E_1$  and  $E_2$  are assumed to determine the rate of increase in A(t) and B(t), as shown in the technical progress functions:

(1) 
$$dA = Ab_1(E_1)dt, \text{ and}$$
$$dB = Bb_2(E_2)dt$$

[20, 60-69]. The function b(.) and b(.) are assumed to be continuous and twice differentiable, with positive first and negative second derivatives.

Assume that the firm maximizes the expected discounted utility of the ternimal state variables wealth W and the value of the stocks of accumulated knowledge A and B such that

(2) 
$$J(S_0) = \operatorname{Max} E\left(e^{-rt}V(S_T) \mid S_0 = \overline{S_0}\right)$$

where J is the indirect utility function resulting from the optimization; S is the vector of state variables W, A, and B; r is the firm's rate of time preference; T is the terminal time; and V(.) represents the utility function of terminal state variables.

Excluding the value of accumulated knowledge, the firm's wealth consists of an inventory of risk-free assets, L, valued at market price,  $P_L$  (Where negative L denotes liabilities), or  $W = P_L L$ . Using Itô's lemma, the change in wealth can be specified as

$$(3) dW = LdP_{L} + P_{L}dL + dP_{L}dL$$

(10;13). The term  $LdP_L$  represents capital gains on the current inventory of assets whereas the remaining two right-hand variables are the net value of additions to wealth from sources other than capital gains. In the absence of consumption, these last two terms represent the income or losses from production such that

$$(4) \qquad (P_{\perp} + dP_{\perp})dL = \pi dt + d\pi,$$

where  $\pi dt + d\pi$  is the income or losses from production and  $P_L + dP_L$  is the price of risk-free assets, dL assumed to be purchased with this production income (14).

If we specify the change in the value of risk-free assets as

$$\frac{dP_{L}}{P_{L}} = \delta_{L}dt,$$

where  $\delta_L$  is the known growth rate of  $P_L$ , by substituting this equation and (4), we can rewrite (3) as

(5) 
$$dW = (\delta_L W + \pi)dt + d\pi.$$

We can introduce distributions of production level and input prices as follows:

$$dY = Y\sigma_y dz_y$$
,  
 $dP_1 = P_1 \delta_1 dt$ , and  
 $dP_2 = P_2 \delta_2 dt$ ,

where  $\sigma_y$  is the standard deviation of the percentage change in production,  $\delta_i$  is the forecast of the growth rate for input price i, and  $z_y$  is a Wiener process in which  $E[dz_y(t)] = 0$  and  $E[dz_y(t)]^2 = dt$  [10]. If input i increases, the first-derivative of the standard deviation of output with respect to input  $X_i$ ,  $\partial \sigma_y / \partial X_i > 0$ , then,  $X_i$ , is risk increasing. The opposite is true for risk-

reducing technologies (7).2

Expected production income at the current time may be expressed as

$$\pi = PY - P_1X_1 - P_2X_2 - E_1 - E_2$$

where P is the known output price. By using Ito's lemma, we can stochastically differentiate for the change in production income to get

$$d\pi = PdY - X_1dP_1 - X_2dP_2$$
  
=  $PY\sigma_ydz_y - X_1P_1\delta_1dt - X_2P_2\delta_2dt$ .

By substituting these expressions into (5), we get the following dynamic budget constraint

(6) 
$$dW = (\delta_L W + PY - P_1 X_1 (1 + \delta_1) - P_2 X_2 (1 + \delta_2) - E_1 - E_2) dt + PY \sigma_V dz_V.$$

In summary, the objective of the firm is to maximize the expected discounted utility of terminal stock of state variables (2) subject to the dynamic budget constraint (6) and the technical progress constraints (1). The variables  $X_1$ ,  $X_2$ ,  $E_1$ , and  $E_2$  are the control and choice variables, whereas W, A, and B are the stock and state variables.

The Itô version of the Bellman equation associated with this model specification is

$$-J_{1} = \max\{J_{w}\{\delta_{L}W + PF - P_{1}X_{1}(1 + \delta_{1}) - P_{2}X_{2}(1 + \delta_{2}) - E_{1} - E_{2}\} + J_{A}\{Ab_{1}(E_{1})\} + J_{B}\{Bb_{2}(E_{2})\} + 1/2 J_{ww} \{P^{2}F^{2}\sigma_{y}^{2}\}\},$$

where  $J_w$  and  $J_{ww}$ , respectively, are the first- and second-order derivatives with respect to wealth, and  $J_A$  and  $J_B$  are the first-order derivatives with respect to A ant B, respectively.

<sup>&</sup>lt;sup>2</sup> The static uncertainty literature contains some other definitions of risk-reducing and risk-increasing inputs. See Pope and Kramer(16) for a detailed discussion.

The expression to be maximized is a dynamic certainty equivalent denominated in utils. For convenience, we call it a stochastic Hamiltonian [13]. The stochastic Hamiltonian and first-order conditions for  $X_1$ ,  $X_2$ ,  $E_1$  and  $E_2$  can be written as

$$H = J_{w} \left[ \delta_{L}W + PF(AX_{1}, BX_{2}) - P_{1}X_{1}(1 + \delta_{1}) - P_{2}X_{2} \right]$$

$$(1 + \delta_{2}) - E_{1} - E_{2} + J_{A}(Ab_{1}(E_{1})) + J_{B}(Bb_{2}(E_{2}))$$

$$+ 1/2J_{ww}P^{2}F^{2}\sigma_{y}^{2},$$

(7) 
$$\frac{\partial \boldsymbol{H}}{\partial \boldsymbol{X}_1} = \boldsymbol{J}_{w} [\boldsymbol{P} \boldsymbol{F}_1 \boldsymbol{A} - \boldsymbol{P}_1 (1 + \delta_1)] + \boldsymbol{J}_{ww} [\boldsymbol{P}^2 \sigma_y^2 \boldsymbol{F} \boldsymbol{F}_1 \boldsymbol{A} + \boldsymbol{P}^2 \boldsymbol{F}^2 \sigma_y \sigma_{yx1}] = 0,$$

(8) 
$$\frac{\partial \mathbf{H}}{\partial \mathbf{X}_2} = \mathbf{J}_{\mathbf{w}} [\mathbf{P} \mathbf{F}_2 \mathbf{B} - \mathbf{P}_2 (1 + \delta_1)] + \mathbf{J}_{\mathbf{w}\mathbf{w}} [\mathbf{P}^2 \sigma_y^2 \mathbf{F} \mathbf{F}_2 \mathbf{B} + \mathbf{P}^2 \mathbf{F}^2 \sigma_y \sigma_{yx2}] = \mathbf{0},$$

(9) 
$$\frac{\partial \mathbf{H}}{\partial \mathbf{E}_1} = -J_{\mathbf{w}} + J_{\mathbf{A}} A b_1' = 0$$
, and

(10) 
$$\frac{\partial \mathbf{H}}{\partial \mathbf{E}_2} = -\mathbf{J}_{\mathbf{w}} + \mathbf{J}_{\mathbf{B}}\mathbf{B}\mathbf{b}_2' = 0.$$

 $F_1$  and  $F_2$  are the first-order derivatives of the production function with respect to  $AX_1$  and  $BX_2$ , respectively,  $b_1'$  and  $b_2'$  are the first-order derivatives of the technical progress function with respect to  $E_1$  and  $E_2$ , respectively, and  $\sigma_{yx1}$  and  $\sigma_{yx2}$  are the first derivation of standard deviation with respect to inputs  $X_1$  and  $X_2$ , respectively.

The first-order conditions in (7) and (8), input decision rules under production uncertainty, can be depicted as

$$PF_1A = P_1(1 + \delta_1) - J_{ww}/J_w \left( P^2 \sigma_y^2 FF_1A + P^2 F^2 \sigma_y \sigma_{yx1} \right), \text{ and } PF_2B = P_1(1 + \delta_2) - J_{ww}/J_w \left( P^2 \sigma_y^2 FF_2B + P^2 F^2 \sigma_y \sigma_{yx2} \right).$$

These relationships have an intuitive interpretation. In the deterministic case with no uncertainty,  $\sigma_y = 0$ , the firm applies inputs to the point at which the marginal value product equals the expected input price. Because J denotes the indirect utility function relating from optimization,  $-J_{ww}/J_w$  will be the Arrow-Pratt measure of absolute risk aversion [3;17]. For a risk-neutral firm in which J is a linear function of wealth ( $J_{ww} = 0$ ), the result is the same as for the deterministic case. For a rist-

averse firm, for which  $J_{ww}/J_w$  is negative and marginal risk premium is positive, inputs will be utilized to the point at which their marginal value product equals input price plus a marginal risk premium. If an input is risk reducing, the marginal risk premium will decrease. On the other hand, if an input is risk increasing, the marginal risk premium will increase. Therefore, if  $X_i$  is risk reducing, a risk-averse firm will apply more  $X_i$ s than will a risk-neutral firm. On the contrary, if  $X_i$  is risk increasing, a risk-averse firm will apply less  $X_i$ s than will a risk-neutral firm.

Combining first-order conditions (9) and (10) yields

(11) 
$$J_{w} = J_{A}Ab'_{1}(E_{1}) = J_{B}Bb'_{2}(E_{2}).$$

This equation implies that  $E_1$  and  $E_2$  will be determined at the point the marginal utility of wealth equals the marginal utility of A or B times the stock level of factor augmentation and marginal productivity of the technical progress function. To retrieve the relationship between production uncertainty and the bias of technical change, we differentiate (11) partially with respect to W, A and B, and also differentiate (11) by using Itô's lemma:

(12) 
$$J_{ww} = J_{Aw}Ab'_1 = J_{Bw}Bb'_2$$
, and  
(13)  $Ab'_1 dJ + J_Ab'_1 dA + b'_1 dJ_AdA = Bb'_2 dJ_B + J_Bb'_2 dB + b'_2 dJ_BdB$ .

The evolution of the marginal indirect utility of state, that is another group of first-order conditions [13, 111-112], can be expressed as

(14) 
$$dJ_{w} = -J_{w}\delta_{L}dt + J_{ww}PF^{*}\sigma_{y}dz_{y}$$
(15) 
$$dJ_{A} = -[J_{w}PF_{1}^{*}X_{1}^{*} + J_{A}b_{1}^{*} + J_{ww}P^{2}F^{*}F_{1}^{*}X_{1}^{*}\sigma_{y}^{2}]dt + J_{Aw}PF\sigma_{y}dz_{y}, \text{ and}$$
(16) 
$$dJ_{B} = -[J_{w}PF_{2}^{*}X_{2}^{*} + J_{B}b_{2}^{*} + J_{ww}P^{2}F^{*}F_{2}^{*}X_{2}^{*}\sigma_{y}^{2}]dt + J_{Bw}PF\sigma_{y}dz_{y}.$$

where \* represents the optimal value. Substituting (11), (12), (15), and (16) into (13) and simplifying yields.

(17) 
$$\frac{b_1'(E_1^*)}{b_2'(E_2^*)} = \frac{PF_2^* X_2^* B + J_{ww}/J_w P^2 F^* F_2^* X_2^* \sigma_y^2 B}{PF_1^* X_1^* A + J_{ww}/J_w P^2 F^* F_1^* X_1^* \sigma_y^2 A}$$

This set of equations constitutes the model.

### II. The Results

The framework described in the preceding in based on the existing literature. The derivation of the propositions that are the principal contribution of this paper is now straightforward. Substitute (7) and (8) into (17) to obtain

(18) 
$$\frac{b_1'(E_1^*)}{b_2'(E_2^*)} = \frac{X_2'[P_2(1+\delta_2)-J_{ww}/J_w P^2 P^{2*}\sigma_y \sigma_{yx2}]}{X_1'[P_1(1+\delta_1)-J_{ww}/J_w P^2 P^{2*}\sigma_y \sigma_{yx1}]}.$$

Equation (18) summarizes the important results of this paper. Assume, without loss of generality, that  $b_1(E_1)$  and  $b_2(E_2)$  have the same functional form. Then, if  $b_1(E_1)/b_2(E_2)$  is less (equal or greater) than 1, the rate of factor augmentation for  $X_1$  is higher (equal or lower) than that for  $X_2$ , i.e.,  $\dot{A}/A > (= \text{or } <) \dot{B}/B$ , where  $\dot{A}$  and  $\dot{B}$  are the time rates of change. The term  $J_{ww}/J_w$  will be negative for the individual's concave utility functions. The term  $P_1(1 + \delta_1)$  represents the expected costs of input i caused by an expected increase in its price. The term  $J_{ww}/J_w$  ( $P^2F^{2*}\sigma_y\sigma_{yx_1}$ ) will be negative for as long as the i-th input increases the variability of output.

If there is no uncertainty,  $\sigma_y$  will equal zero and  $\delta_1$  and  $\delta_2$  will be the known input prices. To further simplify, assume that both inputs have similar costs,  $P_1X_1 = P_2X_2^* = P_1X_1^*$ . This simplifies (18) to

(19) 
$$\frac{b_1'(E_1^*)}{b_2'(E_2^*)} = \frac{P_i X_i^*(1+\delta_2)}{P_i X_i^*(1+\delta_1)}.$$

Equation (18) and (19) leads directly to the following propositions.

**Proposition 1 (Hicksian).** When there is no uncertainty or when firms are risk neutral, relative input prices will determine the rate of bias of endogenous technical progress.<sup>3</sup> This Hicksian proposition is, however, a special case. In situations in which firms are risk averse and the production process is uncertain, the bias of technical change will be different from that predicted by Hicks. The magnitude of this difference will depend on the difference between  $\sigma_{yxl}$  and  $\sigma_{yx2}$  and on  $\sigma_y$  and  $J_{ww}/J_{wv}$ 

If  $X_1$  is risk increasing  $(\sigma_{yx1} > 0)$  and  $X_2$  is risk reducing  $(\sigma_{yx2} < 0)$ , then technical change will be biased toward  $X_2$  in a manner that is independent of relative input prices. This result can best be seen by assuming that  $\delta_1 = \delta_2 = \delta_i$ ; i.e., the expected growth rates of input prices are similar, and if this is the case, (18) can be written as

(20) 
$$\frac{b_1'(E_1^*)}{b_2'(E_2^*)} = \frac{P_i X_i^*(1+\delta_i) - J_{ww}/J_w X_2^* P^2 F^{2*} \sigma_y \sigma_{yx_2}}{P_i X_i^*(1+\delta_i) - J_{ww}/J_w X_1^* P^2 F^{2*} \sigma_y \sigma_{yx_1}}.$$

This equation leads directly to proposition 2.

Proposition 2. For a risk-averse firm facing production uncertainty, the rate of endogenous technical progress will be biased even if there is no difference in the growth rates of input prices. Given risk aversion, the rate of factor-augmenting technical change for those inputs whose use increases production risk is higher than the rate of factor-augmenting technical change for those inputs whose use decreases production risk.

The increased production risk is an implicit cost associated with the use of the risk increasing input, while the decreased production risk is an implicit benefit associated with the use of the risk reducing input. The firm seeks to economize, through higher factor-augmenting technical change, on the use

<sup>&</sup>lt;sup>3</sup> Sato and Ramachandran (19) define the rate of bias of endogenous technical progress as the difference in the rates of change of factor augmentation and show that in a steady state this will equal the difference in the rates of growth of input prices. Using their definition, we call this result the Hicksian proposition.

of inputs with higher explicit and implicit relative costs.

The risk-induced bias may reinforce or counteract the bias caused by relative input price movements. This result implies that A/A can be equal to B/B, even if the firm faces different expected growth rates in input prices. This implication leads to corollary 1.

Corollary 1. For a risk-averse firm facing different expected growth rates in input prices and production uncertainty, the Hicks-neutral technical change can not be excluded because of the existence of production uncertainty and risk.

## **III. Concluding Remarks**

The theory of endogenous technical progress assumes that changes in the relative prices of inputs will lead to the invention of technologies. Such inventions may be undertaken by a public institution or by individual firms. In this paper, a dynamic. microeconomic, factor-augmenting technical change model is developed to investigate how production uncertainty affects the bias of technical change. Uncertainty has been modeled as a set of Wiener processes in whih the level of variance of these processes implies the level of uncertainty. We show that uncertainty in prices or production may affect the rate of bias of technical progress.

Under production uncertainty, technical progress for the risk-averse firm will be affected by the characteristics of inputs (i.e., risk increasing or risk reducing). Technical change will be biased toward risk-reducing inputs and against risk-increasing inputs. It is also shown that the Hicksian proposition is a special case in which there is no uncertainty or in which firms are risk neutral.

From a policy perspective, the principal conclusion from this analysis is that the degree of output uncertainty will influence the types of technology that are developed and adopted. For example, new hybrid varieties of corn might be more favorably received in parts of the United States where weather conditions are more stable, whereas risk-reducing innovations would be more useful in areas where weather patterns are more volatile. Internationally, one would expect the risk-increasing technologies of the Green Revolution to be successfully adopted in countries or regions with stable output patterns. Also, the results suggest that research should concentrate on technologies that reduce risk in regions where production uncertainty is high.

#### REFERENCES

- [1] Antle, J. M. "The Structure of U.S. Agricultural Technology, 1910-78." American Journal of Agricultural Economics, November 1984, 414-21.
- [2] Arnold, L. Stochastic Differential Equations: Theory and Applications. Nwe York: John Wiley & Sons, 1974.
- [3] Arrow, K. J. Essays in the Theory of Risk Bearing. Chicago: Markham, 1971.
- [4] Binswanger, H. P., "The Measurement of Technical Change Biases with Many Factors of Production." *American Economic Review*, December 1974, 964-76.
- [5] Binswanger, H. P., "A Microeconomic Approach to Induced Innovation." *Economic Journal*, December 1974, 940-58.
- [6] Fellner, W., "Two Propositions in the Theory of Induced Innovations." *Economic Journal*, June 1961, 305-8.
- [7] Hertzler, G., "Dynamic Production, Marketing, and Investment Decision under Uncertainty: A User's Guide to Itô Stochastic Control for Agriculture." unpublished manuscript, Iowa State University, Ames, Iowa, 1987.
- [8] Hicks, J. R. The Theory of Wages. London: Macmillan, 1963.
- [9] Kamien, M. I. and N. L. Schwartz, "Induced Factor Augmenting Technical Progress from a Microeconomic Viewpoint." *Econometrica*, October 1969, 668-84.
- [10] Kamien, M. I. and N. L. Schwartz. *Dynamic Optimization*. Amsterdam: North-Holland, 1981.
- [11] Kamien, M. I. and N. L. Schwartz. Market Structure and Innovation. Cambridge: Cambridge University Press, 1982.
- [12] Kennedy, C., "Induced Bias in Innovation and the Theory of Distribution." *Economic Journal*, September 1964, 541-47
- [13] Malliaris, A. G. and W. A. Brock. Stochastic Methods in Economics and Finance. New York: North-Holland, 1982.

- [14] Merton, R. C., "Optimum Consumption and Portfolio Rules in a Continuous-Time Model." *Journal of Economic Theory*, June 1971, 373—413.
- [15] Nordhaus, W., "Some Sceptical Thoughts on the Theory of Induced Innovations." Quarterly Journal of Economics, May 1973, 208-19.
- [16] Pope, R. D. and R. A. Kramer, "Production Uncertainty and Factor Demands for the Competitive Firm." Southern Economic Journal, October 1979, 489-501
- [17] Pratt, J., "Risk Aversion in the Small and in the Large." *Econometrica*, January 1964, 122-36.
- [18] Sato, R., "The Estimation of Biased Technical Change." *International Economic Review*, June 1970, 1979—207.
- [19] Sato, R., and R. Ramachandran, "Factor Price Variation and the Hicksian Hypothesis; A Microeconomic Model." Oxford Economic Papers, June 1987, 343-56.
- [20] Sato, R. and G. S. Suzawa. Research and Productivity. Boston: Auburn House, 1983.
- [21] Stevenson, R., "Measuring Technical Bias," American Economic Review, March 1980, 162-73.