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### The Political Economy of Embodied Technologies

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Selected Paper prepared for presentation at the 2016 Agricultural & Applied Economics Association Annual Meeting, Boston, Massachusetts, July 31-August 2

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# The Political Economy of Embodied Technologies

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MAY 2016

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<sup>&</sup>lt;sup>‡</sup> The authors thank Kathleen Segerson for the conversations they had and comments she made on earlier drafts, which significantly improved this work and its presentation. Any remaining errors are the authors' doing.

The presumption of this paper is that some governments value the environment, but others do not. Assuming political uncertainty and capital-intensive technologies, this circumstance yields a political economic process that emphasizes the effect of using current policy to influence future outcomes. The result of the analysis suggests that the optimal dynamic tax is larger than the Pigovian tax and that a standard results in more employment and output and yields higher adoption rates, thus achieving a predetermined pollution target with a lower political economic cost than a tax – with policy outcomes being more resilient to political change.

JEL code: L5, O2, O3, Q2, Q3

Keywords: Embodied technologies, political uncertainty, standard, tax, environmental policy

#### **I. Introduction**

2 A large body of political science literature on Western democracies 3 documents the great diversity in the list of values and actions supported by 4 various political parties and individual candidates (Gunther & Diamond, 2003) -5 diversity that results in policy changing over time. However, can choice of policy 6 instruments affect policy's long-term outcomes, resulting in incumbent 7 governments leaving a legacy attributed to the policy they ushered in? While 8 focusing on a subset of policies that address externalities and lead to the replacement of dirty technologies with clean ones, this work compares the 9 10 long-term impact of two policy instruments: a tax and a standard.

11 The literature on the selection of environmental policies is extensive, 12 starting with Pigou (1932) and later expanded by Baumol and Oates (1971), Weitzman (1974), Buchanan and Tullock (1975), and Laffont and Tirole (1996a, 13 14 1996b), among many others. This literature usually emphasizes efficiency while 15 taking a welfare economic perspective. To correct the pollution externality, the 16 literature usually recommends the use of a tax to provide incentives for polluters 17 to reduce pollution. The literature generally expresses a critical view of forcing 18 technological change and questions the effectiveness of command-and-control 19 policies (Jaffe et al., 2002; Bansal & Gangopadhyay, 2005), arguing that firms are 20 often unclear about the cost of compliance (Miller, 1995; Kemp, 1997; Gerard & 21 Lave, 2005) and that regulators' ability to enforce regulations is questionable 22 (Lutz et al., 2000; Bansal & Gangopadhyay, 2005; Gerard & Lave, 2005; Mohr, 23 2006; Puller, 2006; Mickwitz et al., 2008). Although the literature has criticized 24 the use of policies mandating that firms go beyond the existing technological 25 capabilities of their industry and force technological change, such policies have 26 numerous success stories, including the development of substitutes for chlorofluorocarbons (Ashford et al., 1985; McFarland, 1992), the development of
flue gas desulfurization systems for SO2 control in the power sector (Popp, 2003;
Taylor et al., 2005), and the control of automobile emissions (Jaegul et al., 2010),
among many other examples. While much of the literature takes a welfare
economic perspective, we approach the problem from a political economic
vantage point.

33 The political economic literature recognizes the limitations of 34 governments' survival over time and thus the challenge of enacting policies that 35 will not be overruled in the longer run. To address such issues, the study of public 36 policy problems that pertain to the long run has evolved into two strands of literature: (a) the inefficiencies associated with political opportunity costs that 37 38 draw on the differences between the objectives of politicians and those of voters 39 (e.g., Persson & Tabellini, 1999; Acemoglu & Robinson, 2001; Grossman & 40 Helpman, 2001; Bohn, 2007; Rausser et al., 2011) and (b) politicians' incentive to 41 manipulate current policy and influence both future elections and policy choices 42 of future governments (e.g., Persson & Svensson, 1989; Aghion & Bolton, 1990; 43 Tabellini & Alesina, 1990; Persson & Tabellini, 2000; Azzimonti, 2011; Millner 44 et al., 2013). While building on the second body of literature, this paper 45 investigates the impact of political uncertainty with respect to future governments 46 on the choice of environmental policy instruments, assuming some parties place 47 more weight on the environment than others.

This paper presents a political economic framework of a model for the selection of policy tools in the context of a choice of environmental policy. This framework will provide a new perspective to the literature that aims to explain the use of financial incentives versus direct control policies to address externality problems. The model developed is a two-period model that aims to capture a dynamic feature of political decision making, namely, the transition among ruling 54 parties. The model is based on the presumptions that (i) policy makers at the 55 present aim to design policy outcomes that will survive a political transition and 56 (ii) production units employ a capital-intensive technology that consists of several 57 activities, each with its own fixed proportion properties (i.e., a putty-clay 58 technology). The empirical literature has shown that these technologies – the 59 putty-clay technologies whereby the production coefficients are fixed in the short 50 run – fit well the energy sector (Dasgupta, 1970; Fuss, 1978).

We model a political economic process that emphasizes the effect of using current policy to influence future outcomes. We assumed an industry that relies on capital-intensive technologies and that regulation leads to modifications of the fixed capital-intensive assets and the adoption of clean technologies. However, because of political uncertainty, the party that places more weight on the environment sets an optimal dynamic tax that is larger than the Pigovian tax that simply maximizes the social welfare over time.

68 Using a static framework and assuming a predetermined aggregate 69 pollution level, Hochman and Zilberman (1978) demonstrated that a standard 70 results in more employment and higher output than a tax. Caparros et al. (2015) 71 used a dynamic putty-clay framework to compare the performance of pollution 72 taxes and upper bound regulation on pollution. The framework presented here 73 shows that an upper bound on pollution intensity results in a higher level of output 74 and, more importantly, greater adoption of clean technologies. Our numerical 75 analysis suggests that, given a predetermined aggregate pollution level, a standard 76 influences the rate of adoption much more than a tax but that a tax results in more 77 production units exiting and becoming idle. The standard leads to higher adoption 78 rates in the short run and thus makes reversing policy outcomes harder.

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However, although environmental policy influences future outcomes by

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altering the embodied technology employed in the industry, policy is time-inconsistent. The optimal policy calculated at time zero will not be optimal in future periods when the policy is reevaluated (Fischer, 1980). That is, a fixed policy is inferior to a dynamic policy that changes over time. The dynamic inconsistency occurs even though the government maximizes its political economic objectives at time zero. These results are similar to those presented in the macro literature (Kydland & Prescott, 1977; Calvo, 1978).

87 With embodied technologies, environmental policy yields changes that are 88 more stable over time. The Clean Air Act of 1970 allowed the U.S. 89 Environmental Protection Agency (EPA) to regulate motor vehicle pollution, with 90 the first-generation catalytic converters in 1975 significantly reducing 91 hydrocarbon and carbon monoxide emissions. A major milestone in vehicle 92 emission control technology came in 1980-1981 in response to standards 93 becoming tighter over time. This technological change included emission control 94 systems that optimized the efficiency of the catalytic converter. The 1990 Clean 95 Air Act included even tighter tailpipe standards and was followed with 96 manufacturers increasing the durability, control of evaporative emissions, and 97 computerized diagnostic systems that identify malfunctioning emission controls. 98 However, standards have also promoted more controversial technologies, and 99 some have expressed concern about the ramifications of choosing technologies as 100 opposed to letting the market find the solution. For example, Germany's 101 Renewable Energy Act led to prosperous wind, solar, and anaerobic digester 102 industries. Although several governments have come to power since enactment of 103 this act, and some have viewed the development of wind and solar industries as a 104 waste of taxpayers' money, these industries continue to thrive. Another example 105 of a controversial environmental policy mandating a change that resulted in an 106 economically viable industry in the U.S. is the Renewable Fuel Standards of 2005, which was revised in 2007, and led to the development of a flourishingcorn-ethanol industry.

The paper proceeds as follows. The conceptual model is introduced in section 2. Section 3 describes the two-period game. The analysis begins in section 4, where the tax equilibrium is derived. We switch to a pollution-intensity upper bound in section 5 and compare the outcome of a tax to the outcome of an intensity upper bound in section 6, where we also calculate the effect of policy on adoption. General policy discussion and concluding remarks are presented in section 7.

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#### II. The Conceptual Model

117 We assume a fixed coefficients production function in the short run that 118 are the outcome of past decisions. In each moment, a production unit faces two 119 alternatives: (i) to keep the existing technology, or (ii) to adopt a new technology 120 with a different set of coefficients. This process yields technical coefficients that 121 are continuously modified over time by the choice of the technology. Assuming 122 many production-units face these alternatives, results in a different set of initial 123 coefficients at each point in time. This modeling is consistent with the empirical literature that estimates the adoption of discrete technologies and evaluates 124 125 technological change.

Both Fuss (1978) and Dasgupta (1970) suggested that the putty-clay approach fits the energy sector well, and several studies assessed the impact of energy regulation using the putty-clay specifications (see survey by Khanna and Rao, 2009). Moffitt et al. (1978) applied the putty-clay approach to analyze waste-management regulation, and Sunding and Zilberman (2002) used the putty-clay framework to assess the impact of water market reforms in California. 132 Furthermore, studies by Paris (1992) and Berck and Helfand (1990), among 133 others, showed that the fixed proportion Von-Liebig production function fits well 134 agricultural production systems, thus justifying, for example, the approach taken 135 by Babcock et al. (1997) and Wu et al. (2001) who used putty-clay specifications 136 to assess various payments for ecological service schemes. These and other empirical studies confirm the insight of Houthakker (1955) and Johansen (1972), 137 138 who showed that the putty-clay approach results in aggregate production functions that are well behaved and simple to construct and analyze. 139

140 Formally, we assume a production function of fixed proportions and 141 constant input prices. Also assume a one-variable input that can be measured in 142 monetary terms that capture the costs per output unit and normalize the price of 143 the input to 1. Production also generates pollutants in a fixed proportion. Formally, let  $x \in (0, \bar{x})$  denote the fixed input-output coefficient and  $\beta \in (0, \bar{\beta})$ 144 145 the fixed pollution-output coefficient. A lower x denotes a more cost-efficient 146 production unit and a higher  $\beta$  denotes the more pollution-intensive units. While 147 the model can be more complex to include changes in output and other inputs, 148 among other modifications, for brevity and simplicity, this basic model suffices.

149 Assume a production unit produces one unit or is idle. Define the 150 production unit current-period quasi-rents  $\pi^0$  as

$$\pi^0 \equiv p - x,\tag{1}$$

where *p* denotes output price. The production unit then chooses either to become idle and not produce (i.e.,  $\pi^0 < 0$ ) or remain active, earn non-negative quasi-rents, and produce at capacity (i.e.,  $\pi^0 \ge 0$ ).

154 The production unit may also make irreversible investments, thus

modifying its technology. Irreversible investment means that an investment cannot be fully recovered once installed and this irreversibility limits a production unit's ability to redeploy capital. Although it is sufficient to assume that production units cannot change technologies without costs, we chose the stronger assumption that simply disallowed the redeployment of capital because it made the presentation clearer and simpler.

161 Technically, assume that investment decisions are discrete choices and that a production unit may invest  $I_t^m > 0$  and modify its existing technology via 162 163 adopting the cleaner technology. Let superscript m denote the modified 164 technology, let subscript  $t = \{1,2\}$  denote period t, and assume the modified technology reduces emissions by  $\gamma \ge 0$ . The production unit may invest  $I_1^m > 0$ 165 in period 1 and modify its pollution-intensity coefficient, that is,  $\beta^m = (1 - \gamma)$ . 166  $\beta$ . However, the new technology may also affect the input-output coefficient. Let 167  $\rho \geq 0$  denote the effect of the adopted technology on production costs and 168 assume  $x^m = (1 + \rho) \cdot x$ . These assumptions capture the idea that while the 169 170 average annual fuel cost of generating one-megawatt hour using fossil steam is 171 \$3.73 U.S., it is \$6.71 U.S. using hydroelectricity – that is, it is cheaper to produce one-megawatt hour using the polluting technologies than the cleaner 172 173 ones.1

Improvement in productivity is obtained via practice, self-improvements, and small innovations (Arrow, 1962). Recent studies on adoption of renewable technologies, however, have argued that the negative correlation between cost and capacity is tenuous (Nordhaus, 2009). We do not aim to contribute to this debate and simply assume that the cost of adopting a technology declines with time. Specifically, we assume adoption of second-period technology requires less

<sup>&</sup>lt;sup>1</sup> Data are available at http://www.eia.gov/electricity/annual/html/epa\_08\_04.html (viewed February 18, 2014).

180 upfront capital and does not affect production costs. Formally, we assume  $I_2^m = (1 - \omega)I_1^m$  where  $\omega > 0$  and adopting clean technology in the second period 182 does not change the per-unit production costs, that is,  $\rho = 0$ . For simplicity, we 183 also assume that  $\gamma_2 = \gamma_1 (\equiv \gamma)$ .

184 To derive the aggregate supply of output, denoted by Y, we follow Hochman and Zilberman (1978) - who followed Johansen (1972) - and define an 185 186 output capacity distribution function  $g(\beta, x)$ . The output capacity of production 187 units located in the set  $(\beta, \beta + d\beta) \times (x, x + dx)$  for small  $d\beta$  and dx are 188 simply  $g(\beta, x)d\beta dx$ . We assume that  $g(\beta, x)$  is a smooth function with 189 compact support. This output-capacity distribution function is used to define the output produced by units located in region  $R \subseteq (0, \overline{\beta}) \times (0, \overline{x})$ , that is,  $Y^0 \equiv$ 190  $\int \int_{\mathbb{R}} g(\beta, x) d\beta dx.$ 191

192 We now transition from the individual production unit to the aggregate 193 industry level and define the survival region. Let  $R^0$  denote the survival region 194 in the  $\beta - x$  space, formally defined as follows:

195 
$$R^0 \equiv R(p) = \{(\beta, x) \mid 0 \le \beta \le \overline{\beta}, 0 \le x \le p\},\$$

196 assuming  $p < \bar{x}$ . Put differently, production units  $R - R^0 \equiv \{(\beta, x) \mid 0 \le \beta \le 197 \quad \bar{\beta}, p < x \le \bar{x}\}$  remain idle.

198 The industry generates aggregate output supplied, aggregate pollution 199 demanded, and aggregate input demanded, respectively, as follows:

$$Y^{0} \equiv \int \int_{R^{0}} g\left(\beta, x\right) d\beta dx, \qquad (2)$$

$$Z^{0} \equiv \int \int_{R^{0}} \beta \cdot g(\beta, x) d\beta dx, \text{ and}$$
$$X^{0} \equiv \int \int_{R^{0}} x \cdot g(\beta, x) d\beta dx.$$

Assume a downward-sloping demand function Q = D(p),  $\partial D(p)/$ 201  $\partial p < 0$ . Then, the equilibrium price is determined by  $Y^0 = Q$ . At this 202 equilibrium price  $(p^0)$ , the marginal firm earns zero quasi-rents (i.e., x =203  $p^0$ ). Then, assuming pollution does not affect consumers' benefit from the 204 good consumed, consumer surplus (CS) and producer surplus (PS) are, 205 respectively,

$$CS^{0} = \int_{p^{0}}^{\infty} D(p) dp$$

$$PS^{0} = p^{0} \cdot Y^{0} - X^{0}$$
(3)

The industry generates a flow of pollution, and this flow generates a stock of pollution. Let  $\mathbb{Z}_t = Z_{t-1}(1 - \Psi) + Z_t$  denote the pollution stock in period t, where the flow of pollution in period t is  $Z_t$  and where  $\Psi$  is the pollution-decay parameter. Note that if  $\Psi = 1$ , then only current-period pollution matters. For simplicity, we normalize the initial stock of pollution to 0; thus,  $\mathbb{Z}_1 = Z_1$ . In addition, we assume that policy makers know the period t social cost of pollution,  $C(\mathbb{Z}_t) = \xi \cdot \mathbb{Z}_t^2$  for  $\xi = 0$ .

Finally, we define social welfare assuming separability between economic activities and environmental amenities, and assume that period t social welfare is

$$W_t = CS_t + PS_t - C(\mathbb{Z}_t). \tag{4}$$

#### III. The Two-Party Two-Period Game

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217 We wanted to assess the implications of the presupposition of uncertainty 218 with respect to future governments and to better understand the incumbent 219 governments' strategic incentives to manipulate policy and tie the hands of future 220 governments. Thus, and to maintain clarity, we assume a two-party system (i.e., 221 Party A and Party B) and a two-period game. Furthermore, although in the first 222 period Party A is in power, we assume a random draw determines which party 223 will be in power in period 2; that is, with probability  $\alpha \in (0,1)$ , Party A is in 224 power in the second period, but with probability  $(1 - \alpha)$ , Party B is in power in 225 the second period. We also assume that Party A strives to maximize social 226 welfare, but Party B places no weight on the environment.

Formally, let  $V_{A,t}$  denote Party A's objective function and  $V_{B,t}$  denote Party B's objective function, where subscript  $t \in \{1,2\}$  denotes period t. Also, let  $\delta$  denote the discount rate and  $W_{j,t}$  the social welfare of period t when Party j is in power.

The first period starts with Party A in power. Party A chooses policy to maximize  $V_{A,1}$ :

$$V_{A,1} \equiv W_{A,1} + \delta \big( \alpha W_{A,2} + (1-\alpha) W_{B,2} \big).$$
(5)

Let  $\varrho_{A,1}$  denote the policy that maximizes Eq. (5). Given policy  $\varrho_{A,1}$ , production units decide whether to remain active and whether to adopt the clean technology. These decisions define the first-period survival region. Then, given the policy-modified survival region, current-period profits and welfare materialize. At the beginning of the second period, a random draw transitions the political system to the second period. If this results in Party A staying in power (i.e., Party A is reelected), then the second-period policy maximizes  $V_{A,2}$ :

$$V_{A,2} \equiv W_{A,2} = CS_{A,2} + PS_{A,2} - C(\mathbb{Z}_{A,2}).$$
(6)

Let  $\varrho_{A,2}$  denote the policy that maximizes (6). However, if the random draw results in Party B being in power, then policy is set to maximize the sum of consumer and producer surplus,  $V_{B,2} = CS_{B,2} + PS_{B,2}$ . Once the ruling government sets policy, previously active units decide whether to remain active and active units that have not yet adopted the clean technology decide whether to adopt it. Then, second-period profits and welfare materialize.

This structure mimics, for instance, the response of U.S. power plants to the EPA's proposal to limit emissions of coal-fired plants to 1,100 pounds of CO<sub>2</sub> per megawatt hour. If such proposed regulation is enacted, then coal-fired power plants would either invest in cleaner technologies (e.g., natural gas-fired boilers, carbon storage and capture technologies, co-generation with biomass) or retire the plant. On the other hand, natural gas-fired plants continue to produce electricity using their existing technologies.

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#### IV. The Optimal Dynamic Tax

Even though we assume competitive markets, the optimal tax may be larger than the dynamic Pigovian tax. To illustrate this, we modify Eq. (1) to include a tax  $\tau$ , and let  $\pi_t^{\tau}$  denote the period 1 expected quasi-rents, where units modified the technology in period *t*,  $t \in \{1,2\}$ , or never adopted the technology (i.e.,  $\pi_0^{\tau}$ ):

$$\pi_1^{\tau} = P - (1+\delta) \cdot (1+\rho) \cdot x - l_1^m - (1-\gamma) \cdot T \cdot \beta$$
(7)

$$\pi_2^{\tau} = P - (1 + \delta) \cdot x - \delta \cdot \alpha \cdot I_2^m - T \cdot \beta + \delta \cdot \alpha \cdot \tau_{A,2} \cdot \gamma \cdot \beta \tag{8}$$

$$\pi_0^{\tau} = P - (1 + \delta) \cdot x - T \cdot \beta \tag{9}$$

259 We use Eqs. (7), (8), and (9) to define the first-period survival region (i.e.,  $R^{A,1}$ ),

260 where  $P \equiv p_{A,1} + \delta \{ \alpha p_{A,2} + (1 - \alpha) \cdot p_{B,2} \}$  and  $T \equiv \tau_{A,1} + \delta \cdot \alpha \cdot \tau_{A,2}$ ; that is,

261 
$$R^{A,1} \equiv \{(\beta, x) \mid \{0 \le \pi_1^\tau\} \cup \{0 \le \pi_2^\tau\} \cup \{0 \le \pi_0^\tau\}, 0 \le \beta \le \bar{\beta}, 0 \le x \le p\}.$$

In addition, let  $R^{m,1}$  denote the first-period modification region whereby 262 units located in this region are active units that modified their technology in the 263 first period. Production units located in region  $R^{m,1}$  earn the highest quasi-rents 264 265 when adopting the pollution abatement technology in the first period. These units 266 are the dirtier yet efficient production units (see section 6.1). This suggests that if 267 the EPA's proposal to regulate coal-fired power plants is enacted and results in a 268 pollution tax on these plants, it will lead to inefficient coal plants shutting down, 269 but the relatively efficient plants will shift to more environmentally benign 270 technologies.

How does policy affect the survival and modification regions and thus aggregate pollution? Recall that the pollution-production coefficient of units that adopted the cleaner technology is  $\beta^m = (1 - \gamma)\beta$ . Then, let  $\Delta Z_1 = Z_1|_{\tau_1=0} Z_1|_{\tau_1=\tau_{A,1}}$  denote the reduction of flow of pollution in period 1 (where  $Z_1|_{\tau_1=0}$ denotes pollution assuming no regulation, and  $Z_1|_{\tau_1=\tau_{A,1}}$  denotes pollution assuming first-period tax of  $\tau_{A,1}$ ); that is,

277 
$$\Delta Z_1 = \gamma \cdot \int \int_{R^{m,1}} \beta \cdot g(\beta, x) d\beta dx + \int \int_{R^0 - R^{A,1}} \beta \cdot g(\beta, x) d\beta dx$$

Party A's policy lowered pollution in the first period (i.e.,  $\Delta Z_1 > 0$ ); it 278 279 reduced the number of active units and induced some active units to modify their 280 technology and adopt cleaner technologies. However, policy enacted in the first period also affected the second period's pollution stock;  $\tau_{A,1}$  not only reduced 281 the first-period stock of pollution yielding a decline of  $(1 - \Psi)\Delta Z_1$  in the 282 second-period pollution stock, but first-period policy also permanently changed 283 284 the technology employed by the industry. This suggests that environmental policy 285 that yields changes to existing coal-powered plants will result in a permanent 286 change to the pollution generated by the power sector, a change that will not be 287 reversed if policy is revoked in the future (recall that investment is irreversible). 288 Thus, we propose the following:

*Proposition 1.* Given the aforementioned assumptions, the equilibrium policy
choices made by governments are as follows:

- 291  $\tau_{A,1} = \left(2\xi \mathbb{Z}_{A,1} \qquad \delta(1 \quad \Psi) \left(\alpha 2\xi \mathbb{Z}_{A,2} \qquad (1 \quad \alpha) 2\xi \mathbb{Z}_{B,2}\right) + \delta(1 \quad \alpha) 2\xi \mathbb{Z}_{B,2}\right)$
- 292  $\tau_{A,2} = 2\xi \mathbb{Z}_{A,2}$

$$\tau_{B,2} = 0$$

#### **Proof:** The proof is in Appendix A.

The optimal dynamic tax  $\tau_{A,1}$  is the outcome of current-period pollution (*the static Pigovian tax effect*:  $2\xi\mathbb{Z}_{A,1}$ ), pollution being a stock and thus affecting the next period (*the pollution stock effect*:  $\delta(1 \ \Psi) 2\xi(\alpha\mathbb{Z}_{A,2}$  $(1 \ \alpha)\mathbb{Z}_{B,2})$ ) and uncertainty regarding the future (*the political uncertainty effect*:  $\delta(1 \ \alpha) 2\xi\mathbb{Z}_{B,2})$ . On the other hand, the Pigovian tax denoted  $\tau_{A,1}^{Pigou}$ equals the difference between the marginal private cost and the social cost 301 calculated at the optimal solution; that is,  $\tau_{A,1}^{Pigou} 2\xi \mathbb{Z}_{A,1}$ 302  $\delta(1 \Psi)(\alpha 2\xi \mathbb{Z}_{A,2} (1 \alpha) 2\xi \mathbb{Z}_{B,2})$ . Proposition 1, then, leads to the 303 following:

304 (a) Proposition 1 suggests that the dynamic Pigovian tax may not be the 305 optimal tax from a political economic perspective, even when the number 306 of producers and consumers is large.<sup>2</sup> Given political uncertainty (i.e., 307  $\alpha < 1$ ) and  $\delta > 0$ , in equilibrium, Party A sets a higher pollution tax than 308 the dynamic Pigovian tax; that is,  $\tau_{A,1} > \tau_{A,1}^{Pigou}$ .

309 (b) However, assuming no political uncertainty (i.e.,  $\alpha = 1$ ),

- 310 i. The first-period tax equals the dynamic Pigovian tax; that is, 311  $\tau_{A,1} = \tau_{A,1}^{Pigou}$ .
- 312 ii. In addition, if  $\delta = 0$  and/or  $\psi = 1$ , then the first-period optimal 313 tax is the static Pigovian tax; that is,  $\tau_{A,1} = 2\xi \mathbb{Z}_{A,1}$ .

314 When there is no uncertainty regarding future governments (i.e.,  $\alpha = 1$ ) and pollution is a flow,  $\Psi = 1$ , Party A has no incentive to diverge from the 315 static Pigovian tax (bullet point (b.ii)). However, when  $\Psi < 1$  and  $\alpha = 1$ , the 316 optimal policy that maximizes social welfare is the dynamic Pigovian tax (bullet 317 point (b.i)); that is,  $\tau_{A,1} = 2\xi \mathbb{Z}_{A,1} + \delta 2\xi (1 - \Psi) \mathbb{Z}_{A,2}$ . However, if political 318 319 uncertainty exists (i.e.,  $\alpha < 1$ ), then Party A's optimal policy diverges from the 320 dynamic Pigovian tax and, because investment is irreversible, Party A uses 321 current policy to tie the hands of future governments and force larger changes in 322 current period than suggested by the Pigovian tax (bullet point (a)).

 $<sup>^2</sup>$  Baumol and Oates (1988) showed that when the number of polluting units is small (i.e., one firm pollutes) or the damage is affecting a small number of consumers (i.e., the pollution is negatively affecting one firm), the Pigovian tax would not result in the optimal solution.

Because the game ends after the second period, period 2 tax is simply the Pigovian tax (i.e., $\tau_{A,2} = 2\xi \mathbb{Z}_{A,2}$ ). However, because Party B does not care about the environment,  $\tau_{B,2} = 0$ .

Politicians may elect not to choose a tax instrument because they care about the distributional implications of the regulatory system deployed. We explore the distributional implications of a standard and contrast them with a tax in section 6, but we first introduce the standard in section 5 and characterize its equilibrium outcome.

#### 331 V. An Upper Bound on Pollution

Assume that a tax is not politically feasible, but that the government mandates an intensity upper bound (e.g., upper bound on chemical concentration of polluting elements in waste, air, and water, speed limits, and the California low carbon fuel standard). In the following paragraphs, we assess the efficacy of such a pollution restriction.

337 Technically, let  $\theta_{j,i}$  denote the intensity upper bound set by Party j in 338 period *i* for  $j \in \{A, B\}$  and  $i \in \{1, 2\}$ . We also modify Eq. (1) to include an upper 339 bound on pollution per unit of output and denote production units' quasi-rents as 340  $\pi^{\theta}$ . Using the definition of production units' quasi-rents, we transition from the micro to the macro level. Technically, a unit remains active if  $\beta \in [0, \theta_{A,1}]$  and 341  $\pi^{\theta} \geq 0$ , and a unit adopts clean technology in the first period if  $\beta \in$ 342  $(\theta_{A,1}, \theta_{A,1}/(1-\gamma)]$  and expected quasi-rents are not negative. Otherwise, a unit 343 becomes idle. Similarly, we can describe second-period survival and modification 344 345 regions.

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Monetary charges that equal the marginal social damage result in Pareto

347 optimality. However, estimating the social damage function is challenging 348 (Baumol and Oates, 1971). An alternative cost-efficient approach introduced in 349 the literature, assumes policy is set to achieve a predetermined aggregate pollution 350 level. Building on this alternative approach, we assumed that at the beginning of 351 the game Party A chooses the per-period expected aggregate level of pollution  $\overline{\mathbb{Z}}$ 352 (e.g., stock of pollution should not result in global average temperatures increasing by more than 2°C of their pre-industrial level), and then period 1 353 354 begins and Party A sets policy. Furthermore, assume the intensity upper bound is set at the beginning of each period:  $\theta_{A,1}$  for the first period and  $\theta_{A,2}$  for the 355 second. Note that although pollution is constrained by a predetermined aggregate 356 level in period 1 (i.e.,  $\mathbb{Z}_{A,1} \leq \overline{\mathbb{Z}}$ ), political uncertainty results in policy containing 357 only the expected value of the pollution stock in period 2 (i.e.,  $\alpha \mathbb{Z}_{A,2}$  + 358  $(1-\alpha)\mathbb{Z}_{B,2} \leq \overline{\mathbb{Z}}$ ). Formally, the Lagrangian of Party A's period 1 359 constraint-maximization problem is 360

361 
$$\mathcal{L}_1 = W_{A,1} + \delta \left( \alpha W_{A,2} + (1-\alpha) W_{B,2} \right) + \lambda_1 \left( \mathbb{Z}_{A,1} - \overline{\mathbb{Z}} \right)$$

$$+ \lambda_2 \left( \alpha \mathbb{Z}_{A,2} + (1-\alpha) \mathbb{Z}_{B,2} - \overline{\mathbb{Z}} \right)$$

363 where  $0 < \lambda_1, \lambda_2$  are the Kuhn-Tucker multipliers. Party A sets  $\theta_{A,1}$  and, if 364 elected in period 2, sets  $\theta_{A,2}$ . In period 2, it chooses an intensity upper bound 365  $\theta_{A,2}$  that maximizes Eq. (6) subject to  $\mathbb{Z}_{A,2} \leq \overline{\mathbb{Z}}$ . The Lagrangian of Party A's 366 period 2 constraint-maximization problem is

367 
$$\mathcal{L}_2 = W_{A,2} + \mu (\mathbb{Z}_{A,2} - \overline{\mathbb{Z}})$$

368 where  $0 < \mu$  is the Kuhn-Tucker multiplier. However, if Party B is elected in the 369 second period, then  $\theta_{B,2} = \overline{\beta}$  (recall that Party B does not care about the 370 environment). 371 Assumption 1: Output capacity distribution function  $g(\beta, x)$  is a single-peak

372 *distribution function with the peak up and to the right of the equilibrium outcome.* 

373 Assumption 2: 
$$\frac{\partial^2}{\partial Q^2}(CS + PS) \le 0 \le \frac{\partial}{\partial Q}(CS + PS)$$

While Assumption 1 suggests that most production units are polluting units, Assumption 2 states that in the neighborhood of the equilibrium solution the economic surplus (i.e., producer plus consumer surpluses) increases with quantity at a decreasing rate. Proposition 2 is as follows:

**Proposition 2.** Assume expected aggregate pollution is set at a predetermined per-period aggregate level  $\overline{\mathbb{Z}}$ . Then, the equilibrium intensity upper bound equates the marginal economic cost of regulation (i.e., the effect of the standard on consumer and producer surpluses) to the marginal pollution damage, and this equilibrium is unique.

383 Given Assumptions 1 and 2, the F.O.C. of the constraint-maximization problems (i.e.,  $\frac{\partial \mathcal{L}_1}{\partial \theta_{A,1}} = 0$  and  $\frac{\partial \mathcal{L}_2}{\partial \theta_{A,2}} = 0$ ) suggest that marginal 384 economic cost of regulation *decreases* in  $\theta$  while the marginal pollution damage 385 386 *increases* – a less stringent standard, and therefore a larger  $\theta$ , results in a smaller impact on the consumer and producer surpluses yet leads to more pollution and 387 388 thus larger marginal pollution damage. The F.O.C. of the first period constraint-maximization problem (i.e.,  $\frac{\partial \mathcal{L}_1}{\partial \theta_{A,1}} = 0$ ) also suggests that the 389 390 optimal intensity upper bound increases with the decay parameter and that the 391 intensity upper bound is largest if pollution is a flow.

392 Next, we compare the regulatory outcome of a tax to that of an intensity393 upper bound and discuss employment, output, and adoption.

#### 394 VI. Comparing a Tax to a Standard: The Pareto Distribution

395 This section compares a tax to an intensity upper bound, assuming a 396 generalized Pareto output-capacity distribution function. This distribution 397 function has been used extensively in the trade literature (Helpman et al., 2007; 398 Melitz & Ottaviano, 2008; Chaney, 2008), and its aggregation across firms yields 399 a Cobb-Douglas production function (Houthakker, 1955-1956). Employing a 400 specific distribution function allows us to numerically, as well as conceptually, 401 quantify and evaluate the differences between a tax and an intensity upper bound 402 and assess their impact on the power utility sector in the U.S.

403 Technically, assume production units distributed according to the 404 following generalized Pareto distribution function:

405 
$$g(\beta, x) = A\beta^{\varphi_1 - 1} x^{\varphi_2 - 1} \text{ for } A, \varphi_1, \varphi_2 > 0.$$

This functional form suggests that when  $\varphi_1 = 1$  and  $\varphi_2 = 1$ , the density function is a uniform distribution function where  $g(\beta, x) = A$  for  $\{\{\beta, x\} \mid 0 \le \beta \le \beta, 0 \le x \le \overline{x}\}$  and 0 otherwise. However, when  $\varphi_1 < 1$  ( $\varphi_2 < 1$ ), the density function places more weight on the low-polluting (efficient) production units. However, if  $\varphi_1 > 1$  ( $\varphi_2 > 1$ ), then the density function places more weight on the high-polluting (inefficient) units.

We use the generalized Pareto distribution function to derive the survival regions and calculate output, employment, and adoption. We calculate the value of these variables for both the tax and the standard regimes and use these calculations to compare the two aforementioned policy instruments.

416 The parameters used for the calculations are depicted in Table 1, where

417 total employment in the power utility sector is used to calibrate the productivity parameter (i.e., A) using Eq. (2) and the definition of  $X^0$ . The Energy 418 419 Information Administration (EIA) included data on 6,668 plants that generated a total of about 4 billion megawatts.<sup>3</sup> We also use information on the power plant 420 421 industry in the U.S. to derive estimates of price and investment. We assume a 422 competitive industry that does not affect the equilibrium price. We also assume 423 more density given to high-polluting and inefficient units, a 50% decline in 424 upfront costs in the second period, and that the probability that Party A is 425 reelected is 55%.

Parameter	Value	Parameter	Value	Parameter	Value
Α	2115.3	φ1	2.0	φ <sub>2</sub>	2.0
<b>p</b> <sub>A,1</sub>	10.34	<b>p</b> <sub>A,2</sub>	10.34	р <sub>В,2</sub>	10.34
I <sub>1</sub> <sup>m</sup>	1.5982	I <sup>m</sup> <sub>2</sub>	0.79909	ω	1/2
δ	0.95	α	0.55	γ	0.5
ρ	0.05	ξ	$9.8 * 10^{-5}$	β	1
Ψ	0.75				

426 Table 1. *The baseline parameters*.

427 Our assumptions suggest that in the unregulated environment, the 428 economic conditions yield 56,540 active production units (using Eq. (2) and the 429 definition of  $Y^0$ ), each generating 70,750 megawatts (recall that although 430 production technology varies across the different units, each production unit is

<sup>&</sup>lt;sup>3</sup> Data are available at http://www.eia.gov/electricity/data/browser/

431 assumed to generate one unit of output, which in the calibration is equivalent to 432 70,750 megawatts), with pollution flow of 37,390 units (using Eq. (2) and the 433 definition of  $Z^0$ ). Assuming the power utility sector is producing 30% of annual 434 greenhouse gases produced in the U.S., and given that the U.S. generated about 435 6,500 million tons of CO<sub>2</sub> in 2012,<sup>4</sup> the result is 0.05 million tons of CO<sub>2</sub> per 436 pollution unit (= 6500 · 0.3/37390).

In what follows, assume predetermined pollution stock in period 1 of 20,158 units (a reduction of 46.5% in the pollution level). This level of pollution is compatible with an intensity standard of  $\theta_{A,1} = 0.43759$  and  $\theta_{A,2} =$ 0.39591 and with an optimal (dynamic) tax of  $\tau_{A,1} = 6.58$  and  $\tau_{A,2} = 3.95$ per pollution unit. Because of political uncertainty and pollution not being a flow, the dynamic Pigovian tax of period 1 is smaller than the optimal dynamic tax; that is, the dynamic Pigovian tax is  $4.89 < 6.58 = \tau_{A,1}$ .

However, the question of how adoption rates vary across regimes remains.
We address this question by deriving the output, employment, and adoption rates
under the two alternative regimes.

447 *A. The Tax Regime* 

Building on the aforementioned assumptions, we describe the survival region  $\mathbb{R}^{A,1}$  in the  $\beta$ -x plane assuming a tax regime. Formally, let first-period expected quasi-rents  $\tilde{\pi}_i^{\tau}$  for  $i \in \{0,1,2\}$  denote production units that made a technology choice (0 denotes units that do not adopt, whereas i=1 or i=2 denotes units that adopt technology in period *i*) and assume units are indifferent about remaining active; that is,  $\tilde{\pi}_i^{\tau} = 0$ . The slopes of these lines are

<sup>&</sup>lt;sup>4</sup> Data are available at

http://www.epa.gov/climatechange/Downloads/ghgemissions/US-GHG-Inventory-2014-Chapter-2-Trends.pdf

454 
$$-\frac{\partial \widetilde{\pi}_{1}^{\tau}/\partial \beta}{\partial \widetilde{\pi}_{1}^{\tau}/\partial x} = -2.1106 > -\frac{\partial \widetilde{\pi}_{2}^{\tau}/\partial \beta}{\partial \widetilde{\pi}_{2}^{\tau}/\partial x} = -3.9029 > -\frac{\partial \widetilde{\pi}_{0}^{\tau}/\partial \beta}{\partial \widetilde{\pi}_{0}^{\tau}/\partial x} = -4.4322$$

and their intercepts are  $x(\beta = 0, \tilde{\pi}_0^{\tau} = 0) = 10.3400 > x(\beta = 0, \tilde{\pi}_2^{\tau} = 0) =$ 455  $10.1259 > x(\beta = 0, \widetilde{\pi}_1^{\tau} = 0) = 9.0671$ . We depict  $\widetilde{\pi}_i^{\tau}$  for  $i \in \{0, 1, 2\}$  in 456 Figure 1, and note that DCBEO is the survival region R<sup>A,1</sup>. The tax policy results 457 in inefficient and dirty units exiting the industry (i.e., units located in region  $R^0$  – 458 R<sup>A,1</sup>). The effect of the tax regime on output, employment, and adoption is shown 459 in Table 2. Given that a production unit generates 70,750 megawatts, a tax regime 460 461 yielded a reduction of 1.72 billion megawatts in electricity generated (=  $\frac{(56,540-32,241)\cdot 70,750}{10^9}).$ 462

	No Regulation	Tax Regime	Standard Regime	% Increase Relative to the Tax
Output	56,540	32,241	35,801	10.46%
Employment	389,750	171,480	225,590	31.55%
Pollution	37,693	20,158	20,158	0%
Adoption (output)		21,290	24,974	17.33%

463 Table 2. *The effect of the period 1 policy instruments.* 

464 Some of the active production units adopt the clean technology in the first 465 period, but others do not. We characterize these early adopters and separate them 466 from the late adopters and identify units that do not adopt the technology in either467 period 1 or period 2 (see Appendix B):

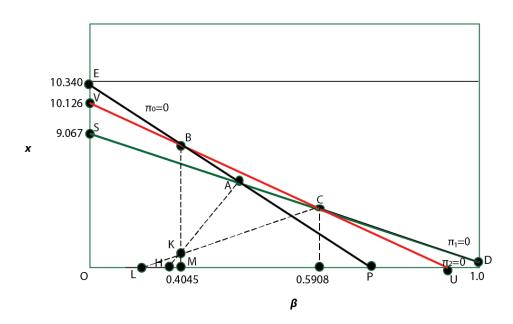
468 a. The line 
$$\pi_1^{\tau} = \pi_0^{\tau}$$
 (i.e., line HA in Figure 1):

$$x_{0.1} = 44.3222 \cdot \beta - 16.3915 \tag{10}$$

469 b. The line  $\pi_1^{\tau} = \pi_2^{\tau}$  (i.e., line LC in Figure 1):

$$x_{1,2} = 33.7355 \cdot \beta - 12.1092 \tag{11}$$

470 c. The line 
$$\pi_2^{\tau} = \pi_0^{\tau}$$
 (i.e., line MB in Figure 1):



$$\beta = 0.4045 \tag{12}$$

471

472 *Figure 1*. First-period survival region under a tax.

473 We use the above-defined lines, together with the boundary of the survival 474 region, to identify units that adopt the new technology in period 1. Units located 475 in region HKCD, namely,  $R_1^m$ , adopt the clean technology in the first period. 476 These are the efficient yet polluting units.

477 However, if Party A is reelected, then some of the units that did not adopt the clean technology in the first period may adopt it in the second period. The late 478 479 adopters are less efficient but less polluting units than the early adopters. The late 480 adopters purchase the clean technology only if Party A is reelected, and they 481 make the purchase at a lower upfront cost than the early adopters (recall that  $\rho =$ 0 in period 2, but that  $\rho > 0$  in period 1 and that  $I_2^m = (1 - \omega)I_1^m$  for  $\omega =$ 482 0.5 > 0). In the numerical model, 21,290 production units (i.e., power plants) 483 484 adopt the clean technology in period 1, while 171,480 production units are active. 485 In the second period, because investment is irreversible, the number of active 486 units does not change. However, the number of active units that adopt the cleaner 487 technology increases by 10,580 units. Note that although more output is produced 488 via a tax in period 2 (see Table 3), more people are employed under a standard 489 (7.7% more people are employed under a standard).

	Tax Regime	Standard Regime	% Change
Output	32,410	29,759	-8.8%
Employment	171,480	184,660	7.7%
Adoption (output)	10,580	20,897	97.5%

490 Table 3. *The effect of the policy instrument in period 2.* 

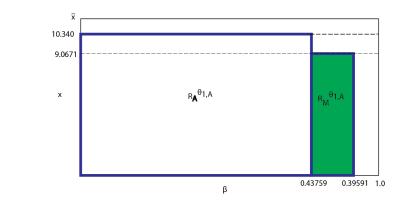
491 *B. Adoption and the emission upper bound* 

492 In section 5, we modify Eq. (1) to include an upper bound on pollution per

unit of output and identify early adopters (i.e., characterize production units that
adopted the clean technology in period 1). We then transition from the micro to
the macro level. We depict this first-period survival region in Figure 2.

Our baseline model suggests that under an intensity upper bound 24,970 production units adopt the clean technology in the first period, whereas the total number of active units in period 1 is 35,801 (Table 2). Let  $R_M^{\theta_{A,1}}$  denote the first-period adoption region (i.e., the green rectangle in Figure 2), and let  $R_A^{\theta_{A,1}}$ denote the region where active units do not adopt the clean technology in period 1. The union of these two regions is the survival region, namely,  $R^{\theta_{A,1}} = R_M^{\theta_{A,1}} \cup$  $R_A^{\theta_{A,1}}$  (i.e., the survival region 35,801 = 10,831 + 24,970).

503 While an upper bound does not affect operation costs, it does affect 504 upfront costs, leading units with large pollution output coefficients to modify their technology (i.e.,  $\{\beta, x\} \in R_M^{\theta_{A,1}}$ ) or exit the industry and become idle (i.e., 505  $\{\beta, x\} \in \mathbb{R}^0 - \mathbb{R}^{\theta_{A,1}}$ ). Our analysis suggests that a standard leads to about 10% 506 more production units remaining active, thus resulting in a significantly lower 507 508 impact on the amount of megawatt generated that becomes idle because of 509 regulation (see Figure 2). It also results in almost 36% more employment in the 510 sector and about 17% more units adopting the clean technology than under a tax.



511

512 *Figure 2*. First-period survival region under a standard.

The Clean Air Act of 1970 gave the EPA authority to regulate motor vehicle pollution and introduce emission control policies. The advent of first-generation catalytic converters in 1975 significantly reduced hydrocarbon and carbon monoxide emissions (see EPA website – http://www.epa.gov); it pushed dirty technologies out of the market and resulted in fewer emissions from motor vehicles.

519 Similar to the tax scenario, in this scenario, the second-period outcome is 520 also conditional on the party in power. The second-period survival region, assuming Party B is in power, equals that of the first period ( $R^{\theta_{B,2}} = R^{\theta_{A,1}}$  and a 521 total of 35,801 active units). Active units in period 1 (i.e., units in region  $R^{\theta_{A,1}}$ ) 522 remain active in period 2. Furthermore, units that modified their technology in the 523 first period, namely,  $R_M^{\theta_{A,1}}$  (i.e., 24,974 active units that modified their 524 525 technology), continue using the cleaner technology in the second period. 526 However, if Party A remains in power, then the survival region may shrink further 527 and we may observe late adopters. Then, the possible outcomes are as follows:

- 528 a. The first-period upper bound is stricter than that of the second period; that 529 is,  $\theta_{A,1} \le \theta_{A,2}$ . This outcome yields a second-period survival region 530 that equals that of the first period, that is,  $R^{\theta_{A,2}} = R^{\theta_{A,1}}$ .
- b. The second-period upper bound is stricter than that of the first period (which is the outcome of the numerical example); that is,  $\theta_{A,1} > \theta_{A,2}$ . This outcome results in late adopters denoted  $R_M^{\theta_{A,2}}$ . In this scenario, early adopters may serve as a bridge to a less polluting industry, where the transition to a cleaner production structure is gradual.

536 b.i. Policy affected the extensive margins and led units to exit the industry 537 in period 2. The units that exited the industry in period 2 belong to the early adopters group, namely,  $R_M^{\theta_{A,1}}$ . The numerical simulation 538 539 suggests that when the predetermined pollution results in an intensity upper bound  $\theta_{A,1} = 0.4376$  and  $\theta_{A,2} = 0.3959$ , 6,042 units that 540 541 adopted the clean technology in period 1 exit the industry in period 2. 542 Many view natural gas-fired power plants as a short-term substitute to 543 aging coal-fired power plants that will be phased out in the long run 544 when technologies with significantly lower carbon footprints become 545 economically viable.

b.ii. Policy also influenced the intensive margins, resulting in active units
adopting the alternative technology in period 2. In the numerical
example, an addition of 1,965 production units adopted cleaner
technology only in the second period.

Returning to our real-world example, EPA emission control policies became progressively more stringent after their introduction in the early 1970s. From 1975 to 2014, light vehicles' average CO<sub>2</sub> grams per mile declined by about 50% while miles per gallon increased by more than 100%.5

554 Our dynamic framework suggests that conditions exist where a standard 555 yields more adoption than a tax, as well as more employment and lower prices. 556 While the economically efficient instrument (i.e., the tax) results in a large impact 557 on the extensive margins leading many units to exit and become idle, the standard 558 seems like the politically efficient instrument of choice because it emphasizes the 559 effect on the intensive margins much more. The standard results in significantly

<sup>&</sup>lt;sup>5</sup> See http://www.epa.gov/otaq/fetrends.htm

fewer units exiting and more employment, but much more adoption in the short term. The increase in adoption yields policy outcomes that are more resilient to political change.

A broad set of parameters results in a standard yielding more output, employment, and adoption in the short run than a tax. For instance, we obtain similar outcomes while revising our baseline parameters (i.e., assuming  $\varphi_1 =$  $3.5 = \varphi_2$  and/or varying the value of  $\gamma$  between 0.15 and 0.85, as well as calibrating the model to a different set of decay parameters).

#### 568 VII. Discussion and Concluding Remarks

569 This paper reevaluates the proposition that market-based instruments 570 should be used to address long-term environmental problems. The paper shows 571 that in a world with uncertainty regarding future governments, establishing a 572 pollution tax in industries relying on capital-intensive technologies may result in 573 the optimal dynamic tax being larger than the Pigovian tax. The paper also shows 574 that, given predetermined aggregate pollution, a standard may result in higher 575 adoption rates than a tax, as well as more employment, higher output, and lower 576 prices.

577 The foundational work of Weitzman (1974) introduced demand and 578 supply uncertainties and concluded that under certain conditions quantity 579 instruments are preferred over price instruments as a mode of regulation. This 580 work expands that line of thinking and shows that a standard may be the preferred 581 mode of regulation because of political uncertainty. Political uncertainty 582 regarding future elections may induce governments to employ a standard to 583 regulate the environment. The standard is less costly politically (i.e., more 584 employment), and it achieves a pre-determined level of aggregate pollution with

585 more adoption and thus solidifies the transition toward clean technologies more 586 than a tax. It is interesting that although economists argue for the use of price 587 instruments, politicians are much keener to employ an intensity upper bound, as 588 the examples in the introduction suggest.

589 Although the model analyzed above suffices to shed new light on the 590 political economy of environmental policy while highlighting the benefits of 591 using an intensity upper bound, this work can be extended in various ways. For 592 instance, we can assume non-random elections. This is motivated by the real 593 world where elections are not random but the outcome of actions taken by the 594 incumbent government. However, how does the analysis change when elections 595 are influenced by existing policy? Assume, for simplicity, that the consumer does 596 not factor into her/his calculation the benefits of policy to the environment (e.g., 597 the horizon is too long and/or the benefits to a consumer are too small to notice). 598 Then, environmental policy is costly to the incumbent government and will 599 dampen the stringency of the policy chosen in the first period but result in a more 600 stringent policy in the second period if Party A is reelected.

601 A second extension introduces commitment. However, our analysis 602 suggests that commitment is not credible. When Party A establishes a binding 603 policy in period 1, the policy choice will be the average across the two periods. 604 Assuming the discretionary tax is higher in the first period suggests that a binding 605 policy will dampen the first-period tax but increase the second-period tax. 606 However, such commitment is not optimal: Party A's choice is less preferred than a policy choice without commitment, and Party B clearly does not benefit from a 607 608 policy that is grandfathered to it.

609 Capital-intensive industries (e.g., power plants) are the main source of 610 anthropogenic emissions. These industries make large capital investments and are 611 not quick to make changes. This study suggests that environmentally conscious 612 governments can exploit this production structure to tie the hands of future 613 governments and yield a permanent environmental change. It also provides 614 political-economic justification for forcing technological change. In future work, 615 we plan to explore empirically the dynamics of capital-intensive industry (i.e., the 616 power sector) and how it responds to regulation. This work will shed new light, 617 for example, on the implications of the Obama administration regulations 618 regarding carbon pollution of existing power plants under Section 111(d) of the 619 Federal Clean Air Act.

620 More generally, while building on the presumption that political survival 621 of parties and individuals is uncertain, this work contributes to the strand of 622 literature that aims to understand politicians' incentives to manipulate current 623 policy and influence both future elections and policy choices of future 624 governments. An example of current policy aiming to influence choices of future 625 governments is the Vienna nuclear deal signed by Iran and the P5+1 (the 626 permanent members of the UN Security Council and Germany) that aspires to 627 prevent the manufacturing of nuclear technologies by Iran. Our study shows that 628 choice of policy instrument may result in more adoption, thus making the reversal 629 of policy outcomes in future periods less attractive.

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# **Appendices:**

# **Appendix A: Proof of Proposition 1**

Before depicting the proof of Proposition 1, we introduce the following notation. Let  $\pi_j^{\tau,t}$  denote production unit quasi-rents in period t, and let subscript *j* denote early adopters (i.e., *j*=1), late adopters (i.e., *j*=2), or units that do not adopt (i.e., *j*=0).

We use this notation to define the first- and second-period survival regions:

I) The first-period survival region is

$$R_{A,1} = \left\{ \{\beta, x\} \middle| 0 \le \beta \le \overline{\beta}, 0 \le x \le \overline{\overline{X}}(\beta) \right\}$$

where  $\overline{\overline{X}}(\beta) \equiv max\{\overline{\overline{x_0}}(\beta), \overline{\overline{x_1}}(\beta), \overline{\overline{x_2}}(\beta)\}\)$ , and the following curves define  $\overline{\overline{\beta}}$ ,  $\overline{\overline{x_0}}(\beta), \overline{\overline{x_1}}(\beta)$ , and  $\overline{\overline{x_2}}(\beta)$ :

1) 
$$\pi_1^{\tau,1}(\beta, x = 0) = 0 \Rightarrow \overline{\beta}$$
 (i.e., point D in Figure 1):  $\overline{\beta} = \frac{P - l_1^m}{T \cdot (1 - \gamma)}$   
2)  $\pi_0^{\tau,1}(\beta, x) = 0 \Rightarrow \overline{x_0}(\beta)$  (i.e., line EP in Figure 1):  $\overline{x_0}(\beta) = \frac{P - T \cdot \beta}{(1 + \delta)}$   
3)  $\pi_1^{\tau,1}(\beta, x) = 0 \Rightarrow \overline{x_1}(\beta)$  (i.e., line SD in Figure 1):  $\overline{x_1}(\beta) = \frac{P - T \cdot (1 - \gamma) \cdot \beta - l_1^m}{(1 + \rho)(1 + \delta)}$   
4)  $\pi_2^{\tau,1}(\beta, x) = 0 \Rightarrow \overline{x_2}(\beta)$  (i.e., line VU in Figure 1):  $\overline{x_2}(\beta) = \frac{P - T \cdot \beta - \delta \cdot \alpha \cdot (l_2^m - \tau_{A,2} \cdot \gamma \cdot \beta)}{(1 + \delta)}$ 

Note that the set  $\{\{\beta, x\} | \beta \in [0, \overline{\beta}], x = max\{\overline{x_0}(\beta), \overline{x_1}(\beta), \overline{x_2}(\beta)\}\}$  defines the line EBCD in Figure 1.

II) The second-period survival region is

$$R_{A,2} = \left\{ \{\beta, x\} \middle| 0 \le \beta \le \min\{\bar{\beta}, \bar{\beta}\}, 0 \le x \le \min\{\bar{X}, \bar{X}\} \right\}$$

where  $\breve{X} \equiv max\{\breve{x}_0(\beta), \breve{x}_1(\beta), \breve{x}_2(\beta)\}\)$ , and the values  $\breve{\beta}, \breve{x}_0(\beta), \breve{x}_1(\beta), \breve{x}_2(\beta)\)$  are defined

as follows:

1) 
$$\pi_1^{\tau,2}(\beta, x = 0) = 0 \Rightarrow \breve{\beta} = \frac{p_{A,2}}{\tau_{A,2} \cdot (1-\gamma)}$$
  
2)  $\pi_0^{\tau,2}(\beta, x) = 0 \Rightarrow \breve{x_0} = p_{A,2} - \tau_{A,2} \cdot \beta$   
3)  $\pi_1^{\tau,2}(\beta, x) = 0 \Rightarrow \breve{x_1} = \frac{p_{A,2} - \tau_{A,2} \cdot (1-\gamma) \cdot \beta}{(1+\rho)}$   
4)  $\pi_2^{\tau,2}(\beta, x) = 0 \Rightarrow \breve{x_2} = p_{A,2} - \tau_{A,2} \cdot (1-\gamma) \cdot \beta - I_2^m$ 

These definitions suggest that  $R_{A,1}(\tau_{A,1}, \tau_{A,2})$  and  $R_{A,2}(\tau_{A,1}, \tau_{A,2})$ . Because policy  $\tau_{A,1}$  affects both first-period and second-period survival regions, it affects not only  $CS_{A,1}$   $PS_{A,1}$ , and  $\mathbb{Z}_{A,1}$ , but also  $CS_{j,2}$   $PS_{j,2}$ , and  $\mathbb{Z}_{j,2}$  for j=A,B.

We use the aforementioned survival regions to characterize the optimal solution to Party A, while beginning with period 2.

Party A chooses  $\tau_{A,2}$  to maximize Eq. (6) and sets  $\tau_{A,2} = 2\xi \mathbb{Z}_{A,2}$  (i.e., the static second-period Pigovian tax).

Party A chooses  $\tau_{A,1}$  to maximize Eq. (5) and the First Order Coditions (F.O.C.) of this maximization problem is

$$0 = \frac{\partial W_{A,1}}{\partial \tau_{A,1}} + \delta \left[ \alpha \frac{\partial W_{A,2}}{\partial \tau_{A,1}} + (1-\alpha) \frac{\partial W_{B,2}}{\partial \tau_{A,1}} \right]$$
$$= \left( \frac{\partial CS_{A,1}}{\partial \tau_{A,1}} + \frac{\partial PS_{A,1}}{\partial \tau_{A,1}} \right) + \delta (1-\alpha) \left( \frac{\partial CS_{B,2}}{\partial \tau_{A,1}} + \frac{\partial PS_{B,2}}{\partial \tau_{A,1}} \right)$$
$$- \left[ \left( 2\xi \mathbb{Z}_{A,1} \quad \delta (1 \quad \Psi) (\alpha 2\xi \mathbb{Z}_{A,2} \quad (1-\alpha) 2\xi \mathbb{Z}_{B,2}) \right) \frac{\partial Z_{A,1}}{\partial \tau_{A,1}} \right]$$
$$+ \delta (1-\alpha) 2\xi \mathbb{Z}_{B,2} \frac{\partial Z_{B,2}}{\partial \tau_{A,1}} \right]$$

In deriving the F.O.C., we use the chain rule and employ the envelope theorem and thus

$$\frac{\partial W_{A,2}}{\partial \tau_{A,1}} = \underbrace{\left(\frac{\partial CS_{A,2}}{\partial R_{A,2}} + \frac{\partial PS_{A,2}}{\partial R_{A,2}} - \frac{\partial C_{A,2}}{\partial R_{A,2}}\right)}_{=0} \frac{\partial R_{A,2}}{\partial \tau_{A,1}} - \frac{\partial C_{A,2}}{\partial R_{A,1}} \frac{\partial R_{A,1}}{\partial \tau_{A,1}} = -2\xi \mathbb{Z}_{A,2} (1 \quad \Psi) \frac{\partial Z_{A,1}}{\partial \tau_{A,1}}$$

(follows from the F.O.C. of the second period assuming Party A remains in power). Furthermore, because investment is irreversible and  $\tau_{B,2} = 0$ , the optimal first-period tax per pollution unit is

$$\tau_{A,1} = 2\xi \mathbb{Z}_{A,1} \qquad \underbrace{\underbrace{\delta(1 \quad \Psi) 2\xi \left(\alpha \mathbb{Z}_{A,2} \quad (1 \quad \alpha) \mathbb{Z}_{B,2}\right)}_{The \ stock \ effect} + \underbrace{\delta(1 \quad \alpha) 2\xi \mathbb{Z}_{B,2}}_{The \ political \ uncertainty \ effect}$$

The final step of the proof is to show that the optimal tariff scheme derived above yields, in equilibrium, the solution that maximizes Party A's objective function. To derive this conclusion, recall that a production unit is active and produces at capacity if its profit is non-negative but becomes idle otherwise, and that  $g(\beta, x)$  is a smooth function with compact support. Furthermore, assuming that policy is binding suggests that the marginal unit earns zero profits; that is, the marginal unit equates its benefit from producing one unit with the cost of the pollution it creates. Let  $\tilde{\tau}_{A,t}$  for  $t \in \{1,2\}$  denote the policy in equilibrium in period *t* and assume  $\tilde{\tau}_{A,1} = \tau_{A,1}$  and  $\tilde{\tau}_{A,2} = \tau_{A,2}$ . By construction, marginal production units under the optimal tariff scheme are the marginal units in the equilibrium and vis versa. Because production units' expected profits decline with  $\beta$  and x, the tax rates,  $\tilde{\tau}_{A,1}$  and  $\tilde{\tau}_{A,2}$ , maximize  $V_{A,1}$ .

The proposition follows.

# **Appendix B:**

Using Eqs. (7) and (9), we derived the linear relationship between input-output and the pollution-output coefficients of production units that are indifferent between adopting the modification in period 1 or not adopting it at all (i.e., the line at which  $\pi_1^{\tau} = \pi_0^{\tau}$ ):

$$x_1(\beta) = \frac{\gamma \cdot \tau_{A,1} + \delta \cdot \alpha \cdot \gamma \cdot \tau_{A,2}}{\rho \cdot (1+\delta)} \cdot \beta - \frac{l_1^m}{\rho \cdot (1+\delta)}.$$
 (1d)

Using Eqs. (7) and (8), we derived the line at which  $\pi_1^{\tau} = \pi_2^{\tau}$ :

$$x_2(\beta) = \frac{\gamma \cdot \tau_{A,1}}{\rho \cdot (1+\delta)} \cdot \beta - \frac{l_1^m - \delta \cdot \alpha l_2^m}{\rho \cdot (1+\delta)}.$$
 (2d)

Finally, using Eqs. (8) and (9), we derived the line at which  $\pi_2^{\tau} = \pi_0^{\tau}$ :

$$\beta = \frac{I_2^m}{\tau_{A,2} \cdot \gamma} \tag{3d}$$

Q.E.D.