

# ONLINE APPENDIX

## Relative Prices and Climate Policy: How the Scarcity of Non-Market Goods Drives Policy Evaluation

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**Abstract:** Climate change not only impacts production and market consumption, but also the relative scarcity of non-market goods, such as environmental amenities. We study fundamental drivers of the resulting relative price changes, their potential magnitude, and their implications for climate policy in Nordhaus' prominent DICE model, thereby addressing one of its key criticisms. We propose plausible ranges for these relative prices changes based on best available evidence. Our central calibration reveals that accounting for relative prices is equivalent to decreasing pure time preference by 0.6 percentage points and leads to a more than 50 percent higher social cost of carbon.

**JEL-Classifications:** Q01, Q54, H43, D61, D90

**Keywords:** Climate policy, discounting, non-market goods, social cost of carbon, substitutability

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## A.1 Derivation of the relative price effect

To derive the relative price effect of non-market goods,  $RPE_t = \frac{d}{dt} \left( \frac{U_{E_t}}{U_{c_t}} \right) \left( \frac{U_{E_t}}{U_{c_t}} \right)^{-1}$  (Equation 4), we first compute marginal utilities with respect to the two goods for utility function (2):

$$U_{E_t} = \alpha(E_t - \bar{E})^{\theta-1} [\alpha(E_t - \bar{E})^\theta + (1 - \alpha)c_t^\theta]^{\frac{1-\eta-\theta}{\theta}} \quad (\text{A.1})$$

$$U_{c_t} = (1 - \alpha)c_t^{\theta-1} [\alpha(E_t - \bar{E})^\theta + (1 - \alpha)c_t^\theta]^{\frac{1-\eta-\theta}{\theta}}. \quad (\text{A.2})$$

We thus have

$$\frac{U_{E_t}}{U_{c_t}} = \frac{\alpha}{(1 - \alpha)} \left( \frac{E_t - \bar{E}}{c_t} \right)^{\theta-1} \quad (\text{A.3})$$

The time derivative of this marginal rate of substitution is given by:

$$\frac{d}{dt} \left( \frac{U_{E_t}}{U_{c_t}} \right) = (\theta - 1) \frac{\alpha}{(1 - \alpha)} \left( \frac{E_t - \bar{E}}{c_t} \right)^{\theta-2} \left[ \frac{\dot{E}_t}{c_t} - \frac{(E_t - \bar{E})\dot{c}_t}{c_t^2} \right] \quad (\text{A.4})$$

With the growth rates  $g_i$  of the two goods  $i \in (E, c)$  defined as  $g_{i_t} = \frac{\dot{i}_t}{i_t}$ , we can rewrite this time derivative using  $\dot{i}_t = g_{i_t} i_t$  as:

$$\begin{aligned} \frac{d}{dt} \left( \frac{U_{E_t}}{U_{c_t}} \right) &= \frac{\alpha}{(1 - \alpha)} \left( \frac{E_t - \bar{E}}{c_t} \right)^{\theta-1} (\theta - 1) \left( \frac{c_t}{E_t - \bar{E}} \right) \left[ \frac{g_{E_t} E_t}{c_t} - \frac{(E_t - \bar{E})g_{c_t} c_t}{c_t^2} \right] \\ &= (1 - \theta) \frac{\alpha}{(1 - \alpha)} \left( \frac{E_t - \bar{E}}{c_t} \right)^{\theta-1} \left[ g_{c_t} - \frac{E_t}{E_t - \bar{E}} g_{E_t} \right]. \end{aligned} \quad (\text{A.5})$$

The relative price effect of non-market goods is therefore given by

$$RPE_t = \frac{\frac{d}{dt} \left( \frac{U_{E_t}}{U_{c_t}} \right)}{\left( \frac{U_{E_t}}{U_{c_t}} \right)} = (1 - \theta) \left[ g_{c_t} - \frac{E_t}{E_t - \bar{E}} g_{E_t} \right]. \quad (\text{A.6})$$

The relative price effect of non-market goods, i.e. the change in relative prices over time, is thus the same as the difference in the two good-specific discount rates (see Weikard and Zhu (2005) or Drupp (2018) for derivations in continuous time).

## A.2 Calibration of non-market damages

### A.2.1 Calibration for Section 3

In Section 3, we replicate the analysis of Sterner and Persson (2008) in DICE-2016R2. Thus, we do not consider a subsistence requirement in the consumption of non-market goods. The non-market good climate damage coefficient  $\psi$  is calibrated for a temperature increase of  $T = 3^\circ\text{C}$  as follows:

$$W_0(E_0, (1 - D_0^\phi)C_0, L_0) = W_0((1 - D_0^\psi)E_0, (1 - D_0^\kappa)C_0, L_0) \Leftrightarrow \quad (\text{A.7})$$

$$\alpha E_0^\theta + (1 - \alpha)((1 - D_0^\phi)C_0)^\theta = \alpha \left(\frac{E_0}{1 + \psi T^2}\right)^\theta + (1 - \alpha)((1 - D_0^\kappa)C_0)^\theta$$

We can solve this for the non-market climate damage parameter  $\psi$  as follows:

$$\psi = \left[ E_0 \left( E_0^\theta + \frac{1 - \alpha}{\alpha} \left( ((1 - D_0^\phi)C_0)^\theta - ((1 - D_0^\kappa)C_0)^\theta \right) \right)^{-\frac{1}{\theta}} - 1 \right] T^{-2}. \quad (\text{A.8})$$

Sterner and Persson (2008) assume that the initial amount of the non-market good is equal to the starting value for material consumption, i.e.  $C_0 = E_0$ . In this case equation (A.8) reduces to

$$\psi = \frac{1}{T^2} \left[ \left( \frac{1 - \alpha}{\alpha} (1 - D_0^\phi)^\theta + 1 - \frac{1 - \alpha}{\alpha} (1 - D_0^\kappa)^\theta \right)^{-\frac{1}{\theta}} - 1 \right]. \quad (\text{A.9})$$

### A.2.2 Calibration for Sections 4 and 5

In the presence of a subsistence requirement in the consumption of non-market goods the calibration is modified as follows:

$$W_0(E_0, (1 - D_0^\phi)C_0, L_0) = W_0((1 - D_0^\psi)E_0, (1 - D_0^\kappa)C_0, L_0) \Leftrightarrow \quad (\text{A.10})$$

$$\alpha (E_0 - \bar{E})^\theta + (1 - \alpha)((1 - D_0^\phi)C_0)^\theta = \alpha \left(\frac{E_0}{1 + \psi T^2} - \bar{E}\right)^\theta + (1 - \alpha)((1 - D_0^\kappa)C_0)^\theta$$

We can solve this for the non-market climate damage parameter  $\psi$  as follows:

$$\psi = \left[ E_0 \left( \bar{E} + \left[ (E_0 - \bar{E})^\theta + \frac{1 - \alpha}{\alpha} \left( ((1 - D_0^\phi)C_0)^\theta - ((1 - D_0^\kappa)C_0)^\theta \right) \right]^{\frac{1}{\theta}} \right)^{-1} - 1 \right] T^{-2}. \quad (\text{A.11})$$

### A.3 Relative prices and comparison of climate policy paths until 2300, with 100% additional non-market damages as in Sterner and Persson (2008)

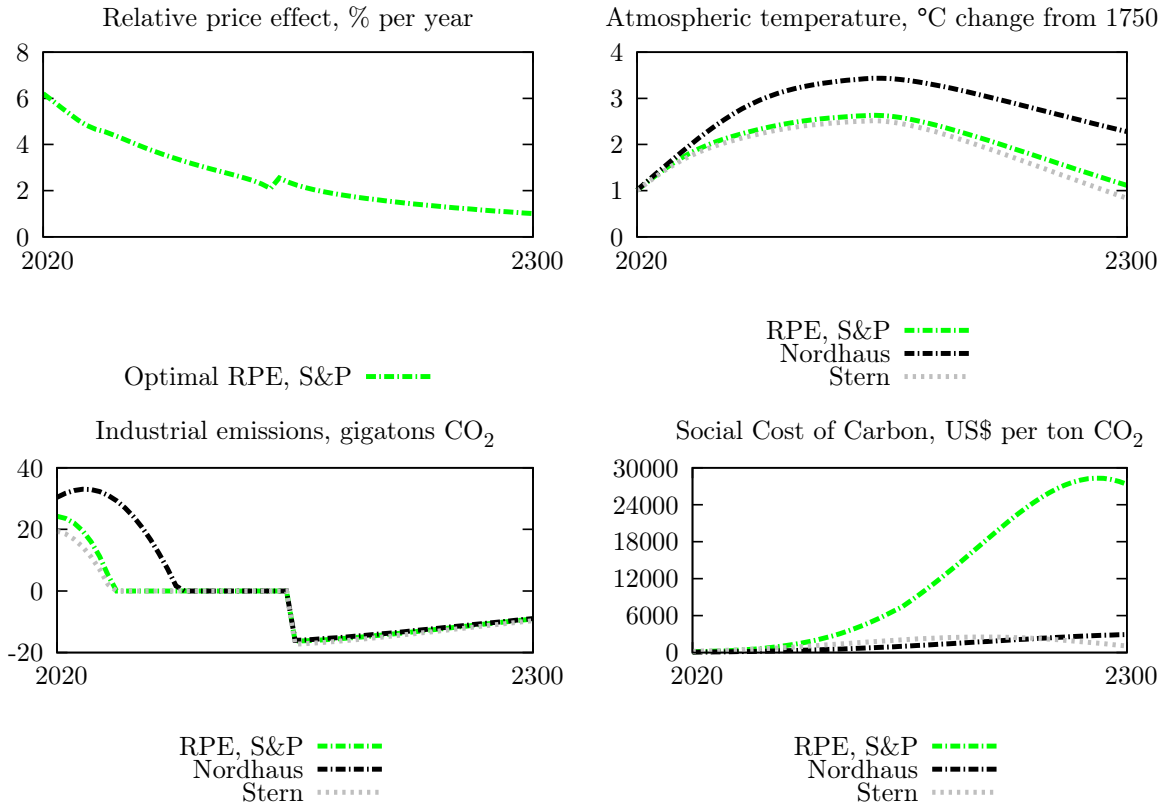


Figure A.1: Relative price effect (*RPE*) and comparison of climate policy paths for a time horizon up to 2300 and 100% additional non-market damages. Otherwise, see description of Figure 1.

## A.4 Re-calibration of the model

### (1) Derivation of $\eta_C$

We make use of Equations 2 and Equation 8 with  $\bar{E} = 0$  to recalibrate the effective elasticity of marginal utility of market consumption,  $\eta_C$ , such that the model with relative prices yields the same paths of market consumption and investments as the standard DICE version. We have

$$U = \frac{1}{1-\eta} \left[ \alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta \right]^{\frac{1-\eta}{\theta}}, \quad (\text{A.12})$$

$$U_C = (1-\alpha) c_t^{\theta-1} \left[ \alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta \right]^{\frac{1-\eta-\theta}{\theta}}, \quad (\text{A.13})$$

$$U_{CC} = (1-\alpha) c_t^{\theta-1} \frac{1}{c} (\theta-1) \left[ \alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta \right]^{\frac{1-\eta-\theta}{\theta}} \quad (\text{A.14})$$

$$\begin{aligned} &+ (1-\alpha) c_t^{\theta-1} \left[ \alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta \right]^{\frac{1-\eta-\theta}{\theta}} \frac{(1-\alpha) c_t^{\theta-1}}{\alpha E^\theta + (1-\alpha) c^\theta} (1-\eta-\theta) \\ &= (1-\alpha) c_t^{\theta-1} c^{-1} \left[ \alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta \right]^{\frac{1-\eta-\theta}{\theta}} \times \\ &\quad \left[ (\theta-1) + (1-\eta-\theta) \frac{(1-\alpha) c^\theta}{\alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta} \right] \end{aligned} \quad (\text{A.15})$$

Combining these ingredients yields the effective elasticity of marginal utility of market consumption,  $\eta_C$ , as

$$\eta_C = -\frac{U_{CC} c}{U_C} = (1-\theta) - (1-\eta-\theta) \frac{(1-\alpha) c_t^\theta}{\alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta}. \quad (\text{A.16})$$

Defining the value share of the consumption good as  $\beta^* = \frac{(1-\alpha) c_t^\theta}{\alpha (E_t - \bar{E})^\theta + (1-\alpha) c_t^\theta}$ , (cf. Gerlagh and van der Zwaan (2002), Hoel and Sterner (2007), Traeger (2011)), this can be rewritten as

$$\eta_C = \beta^* \eta + (1-\beta^*) (1-\theta) \quad (\text{A.17})$$

That is, when the full value share accrues to market consumption goods, the effective elasticity of marginal utility of market consumption,  $\eta_C$ , equals the overall elasticity of marginal utility,  $\eta$ . Yet, as soon as non-market goods have a positive value share, the degree of substitutability matters for the effective elasticity of marginal utility of market consumption.

**(2) Time-path of  $\eta(t)$  used for re-calibration**

<u><math>\theta = -1</math> (left column)</u>	<u><math>\theta = 1</math> (right column)</u>
0 1.388889	0 1.611111
1 1.399128	1 1.584113
2 1.407008	2 1.563337
3 1.413283	3 1.546792
4 1.418342	4 1.533453
5 1.422468	5 1.522575
6 1.425869	6 1.513609
7 1.428699	7 1.506148
8 1.431075	8 1.499881
9 1.433088	9 1.494575
10 1.434805	10 1.490046
11 1.436281	11 1.486154
12 1.437558	12 1.482786
13 1.438670	13 1.479854
14 1.439644	14 1.477285
15 1.440501	15 1.475023
16 1.441261	16 1.473020
17 1.441937	17 1.471237
18 1.442543	18 1.469638
19 1.443090	19 1.468192
20 1.443587	20 1.466877
21 1.444036	21 1.465682
22 1.444443	22 1.464596
23 1.444813	23 1.463606
24 1.445148	24 1.462703
25 1.445452	25 1.461880
26 1.445729	26 1.461127
27 1.445982	27 1.460439
28 1.446211	28 1.459812
29 1.446423	29 1.459230
30 1.446618	30 1.458691
31 1.446798	31 1.458192
32 1.446965	32 1.457731
33 1.447119	33 1.457305

34	1.447261	34	1.456910
35	1.447393	35	1.456544
36	1.447515	36	1.456205
37	1.447629	37	1.455891
38	1.447735	38	1.455600
39	1.447833	39	1.455330
40	1.447925	40	1.455078
41	1.448011	41	1.454845
42	1.448091	42	1.454627
43	1.448166	43	1.454424
44	1.448237	44	1.454235
45	1.448303	45	1.454059
46	1.448365	46	1.453894
47	1.448423	47	1.453739
48	1.448478	48	1.453595
49	1.448530	49	1.453459
50	1.448579	50	1.453333
51	1.448625	51	1.453214
52	1.448668	52	1.453102
53	1.448710	53	1.452996
54	1.448749	54	1.452897
55	1.448786	55	1.452804
56	1.448821	56	1.452716
57	1.448855	57	1.452633
58	1.448887	58	1.452555
59	1.448917	59	1.452481
60	1.448946	60	1.452411
61	1.448973	61	1.452345
62	1.449000	62	1.452282
63	1.449025	63	1.452222
64	1.449048	64	1.452166
65	1.449071	65	1.452112
66	1.449093	66	1.452061
67	1.449114	67	1.452013
68	1.449134	68	1.451967
69	1.449153	69	1.451923

70	1.449171	70	1.451881
71	1.449188	71	1.451841
72	1.449205	72	1.451803
73	1.449221	73	1.451767
74	1.449236	74	1.451732
75	1.449251	75	1.451699
76	1.449265	76	1.451668
77	1.449278	77	1.451638
78	1.449291	78	1.451609
79	1.449303	79	1.451581
80	1.449315	80	1.451555
81	1.449327	81	1.451530
82	1.449337	82	1.451505
83	1.449348	83	1.451482
84	1.449358	84	1.451460
85	1.449367	85	1.451439
86	1.449377	86	1.451418
87	1.449386	87	1.451398
88	1.449394	88	1.451379
89	1.449402	89	1.451361
90	1.449410	90	1.451343
91	1.449418	91	1.451325
92	1.449426	92	1.451308
93	1.449435	93	1.451289
94	1.449443	94	1.451270
95	1.449453	95	1.451247
96	1.449466	96	1.451219
97	1.449483	97	1.451181
98	1.449508	98	1.451123
99	1.449552	99	1.451024
100	1.449493;	100	1.451158;



(3) Figure 2 with re-calibrated  $\eta(t)$

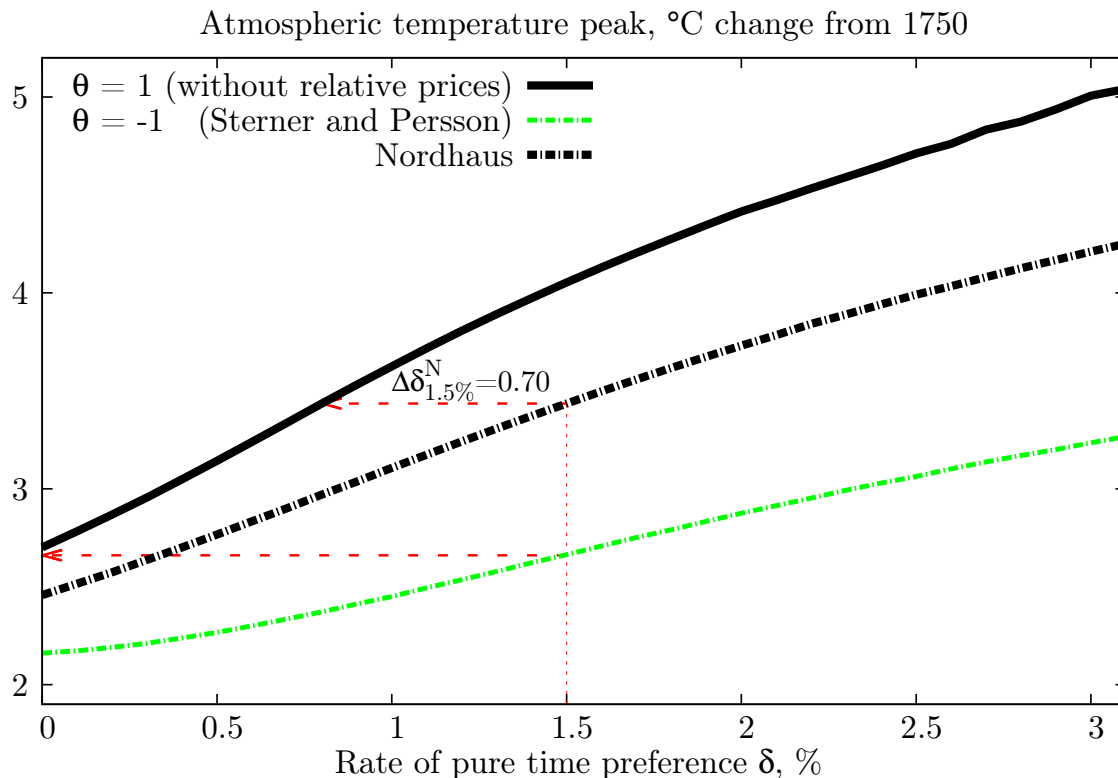


Figure A.2: The comparative influence of introducing relative prices on peak temperature with re-calibrated elasticity of marginal utility. The Figure depicts peak temperature as a function of the rate of pure time preference,  $\delta$ , for different degrees of substitutability,  $\theta$ . The solid black line shows the comparison case of perfect substitutability, i.e. without relative prices. The green line depicts the substitutability assumption of Sterner and Persson, with  $\theta = -1$ , and the dashed black line the ‘Nordhaus’ case. A model run with relative prices is compared to a run without but with a higher  $\delta$  such that peak temperature is the same across both runs.

## A.5 Relative prices and comparison of climate policy paths until 2100, with 25% additional non-market damages as in the standard DICE-2016R2

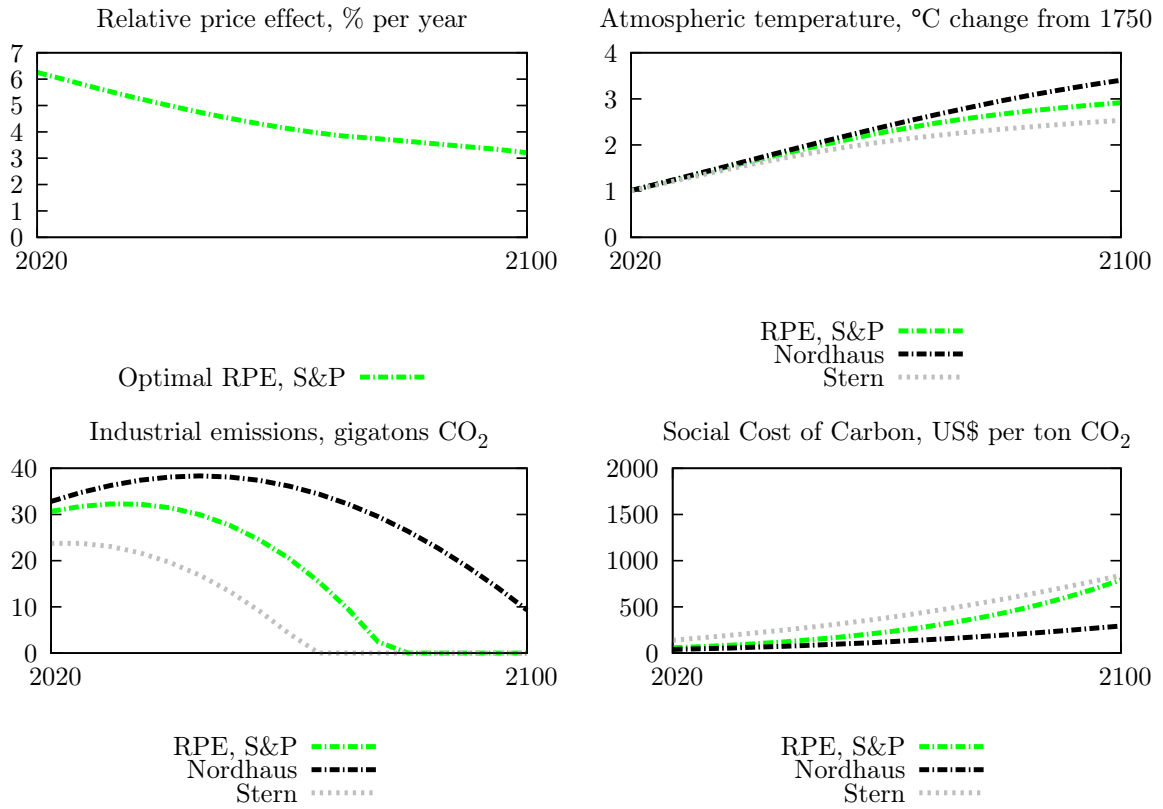


Figure A.3: Relative price effect ( $RPE$ ) and comparison of climate policy paths for a time horizon up to 2100 and 25% additional non-market damages.

## A.6 Relative prices, substitutability and the equivalent reduction in pure time preference

Table A.1: Degree of substitutability ( $\theta$ ) and the corresponding *RPE's* equivalent effect in terms of a reduction of pure time preference ( $\Delta\delta$ )\*

$\theta_i$	$\Delta\delta_{1.5\%}^{\theta_i}$	Reference or relation
1	0	Perfect substitutes
0.8	0.05	
0.6	0.12	Drupp (2018)
0.4	0.18	
0.2	0.4	
0	0.6	Cobb-Douglas, Gollier (2010)
-0.2	0.85	
-0.333	1.01	Kopp et al. (2012)
-0.4	1.09	
-0.66	1.4	Equivalent to Stern's (2007) lower $\delta$
-0.78	1.5	
-1	not defined	Sterner and Persson (2008)
-2.3	not defined	Lowest indirect empirical estimates

\* This analysis is conducted without our replication of Sterner and Persson (2008).

## A.7 Additional drivers of the relative price effect: Initial value of non-market goods and share of non-market goods in utility

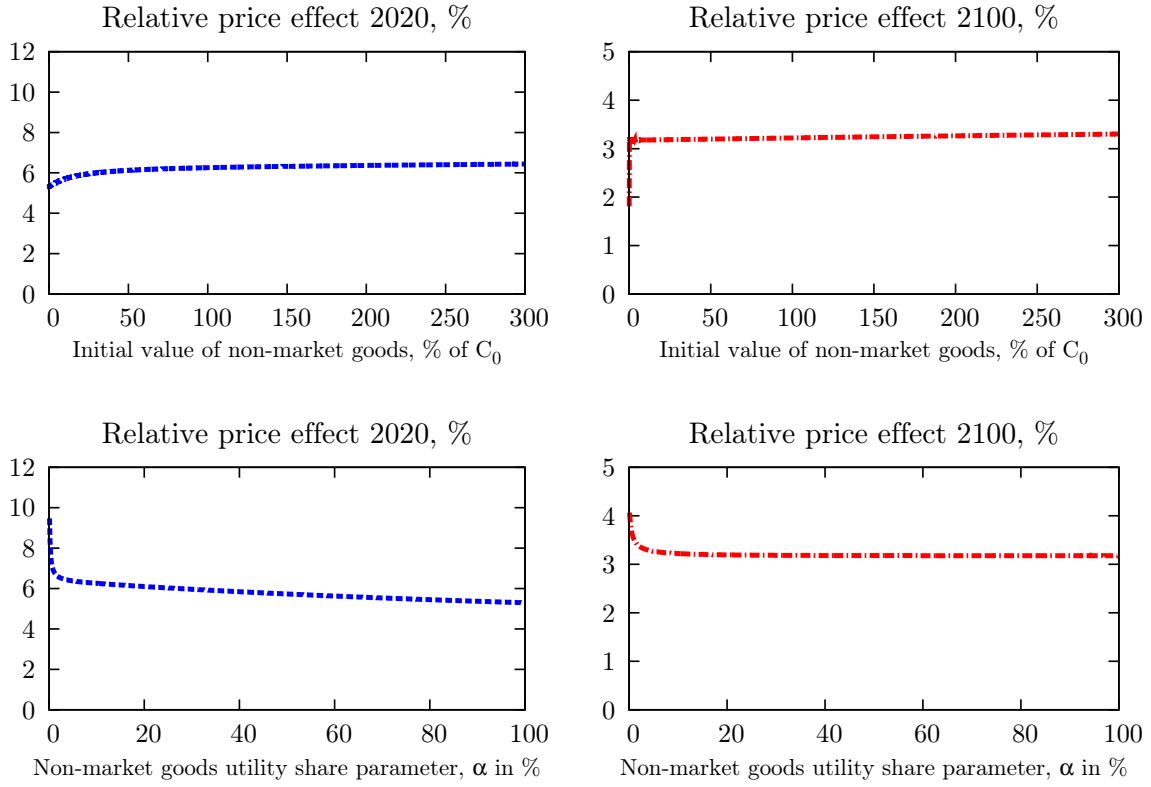


Figure A.4: Additional drivers of the relative price effect (I). Top to bottom: The impact of the initial value of non-market goods and the share of non-market goods in utility on the *RPE* in 2020 (left) and in 2100 (right).

## A.8 Empirical estimates of the substitutability parameter

Table A.2: Indirect empirical estimates of the substitutability parameter\*

Study	Mean $\theta$	Type of good
Alberini et al. (2018)	0.46	Environmental goods: Climate change mitigation
Alberini et al. (2006)	0.53	Reduction in mortality risk
Aldy et al. (1999)	0.58	Environmental goods: Nature conservation
Ara and Tekesin (2016)	0.43	Health improvements
Ara and Tekesin (2017)	0.10	Health improvements
Barbier et al. (2017)	0.74	Environmental goods: Water or air quality
Basili et al. (2006)	0.58	Environmental goods: Water or air quality
Broberg (2010)	0.75	Environmental goods: Nature conservation
Clark and Kahn (1988)	0.10	Cultural goods
Cuevas-Alvarado et al. (2016)	0.73	Environmental goods: All others
Czajkowski et al. (2017)	0.72	Environmental goods: Water or air quality
Czajkowski and Scasny (2010)	0.36	Environmental goods: Water or air quality
Dickie (2005)	0.86	Health improvements
Doucouliafos et al. (2014)	0.48	Reduction in mortality risk
Francisco (2015)	0.52	Environmental goods: Water or air quality
Hammit et al. (2019)	-2.00	Reduction in mortality risk
Hammit and Herrera-Araujo (2018)	0.11	Reduction in mortality risk
Hökby and Söderqvist (2003)	0.32	Environmental goods: All others
Huhtala and Pouta (2008)	0.90	Environmental goods: All others
Jacobsen and Hanley (2009)	0.62	Environmental goods: Nature conservation
Johannesson et al. (1993)	0.82	Health improvements
Li et al. (2016)	0.90	Environmental goods: Climate change mitigation
Lindhjem et al. (2011)	0.20	Reduction in mortality risk
Lindhjem and Tuan (2012)	0.27	Environmental goods: Nature conservation
Martini and Tiezzi (2014)	-0.16	Environmental goods: Water or air quality
Masiye and Rehnberg (2005)	0.71	Health improvements
Masterman and Viscusi (2018)	0.18	Reduction in mortality risk
McLeod (1984)	0.34	Cultural goods
Milligan et al. (2014)	0.45	Reduction in mortality risk
Nastis and Mattas (2018)	0.04	Environmental goods: Climate change mitigation
Navrud and Strand (2018)	-0.02	Environmental goods: Nature conservation
Ready et al. (2002)	0.41	Environmental goods: Water or air quality
Riera et al. (2006)	0.01	Reduction in mortality risk
Robinson et al. (2019)	-0.50	Reduction in mortality risk
Scandizzo and Ventura (2010)	0.55	Environmental goods: Nature conservation
Tuan et al. (2009)	-0.98	Cultural goods
Tyllianakis and Skuras (2016)	0.26	Environmental goods: Water or air quality
Viscusi and Aldy (2003)	0.50	Reduction in mortality risk
Wang and Mullahy (2006)	-0.42	Environmental goods: Water or air quality
Zan and Scharff (2017)	-2.30	Reduction in mortality risk

\* A separate data file contains more details on these 40 as well as the 41 non-relevant studies.

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## A.9 Optimal versus business as usual growth rates of the market and non-market good

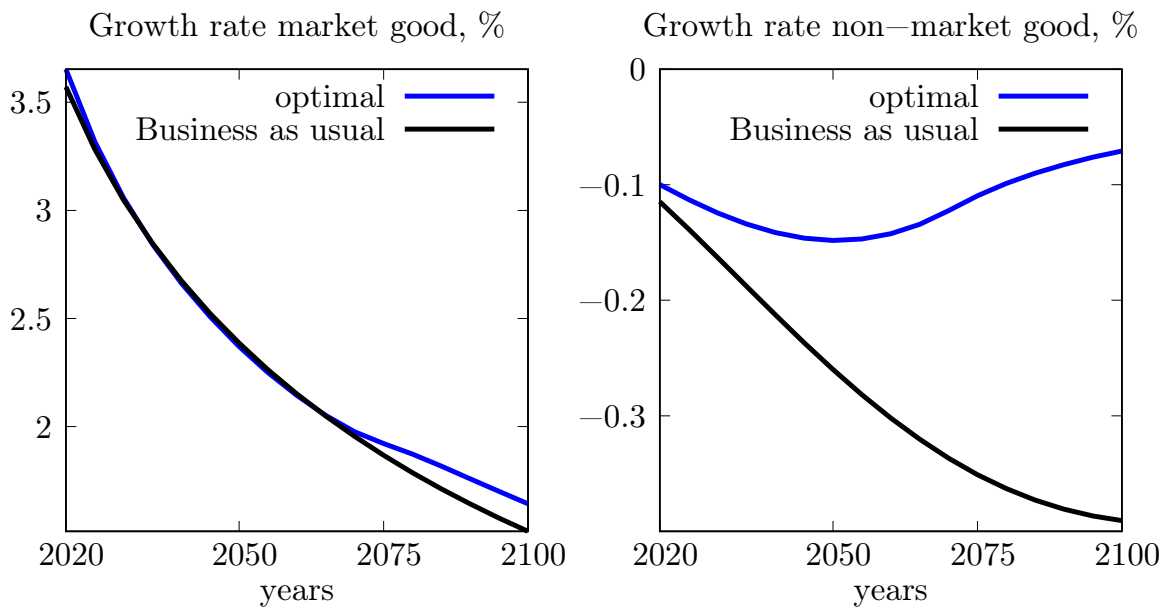


Figure A.5: Optimal versus business as usual growth rates of the market good (left) and non-market good (right) using the central calibration of our plausible ranges model.

## A.10 Code

### A.10.1 AMPL code to produce Figure 1

#### (1) AMPL mod-file Nordhaus

```
# Nordhaus, DICE 2016R2
# To work with the run-file, this mod-file should be named: DICE_Nordhaus_2016.mod

# PARAMETERS
#Time horizon
param T:=100;
# Preferences
param eta default 1.45; # I-EMUC
param rho default 0.015; # time preference rate
# for Stern use param rho default 0.001;
# discount factor
param R {t in 0..T}>=0;
let R[0]:=1;
let {t in 1..T} R[t] := R[t-1]/((1+rho)^5);

# for sensitivity analysis and figure 2 use: param R {t in 0..T}:= 1*exp(-rho*5*t);

# Population and its dynamics
param L0:=7403; #initial world population 2015 (millions)
param gL0:=0.134; #growth rate to calibrate to 2050 pop projection
param L {t in 0..T}>=0;
let L[0]:=L0;
let {t in 1..T} L[t]:=L[t-1]*((11500/L[t-1])^gL0);

# Technology and its dynamics
param gamma:=0.3; #capital elasticity in production function
param deltaK:=0.100; #depreciation rate on capital (per year)
param Qgross0:=105.5; #Initial world gross output 2015 (trill 2010 USD)
param K0:=223; #initial capital value 2015 (trillions 2010 USD)
param A0:=5.115; #initial level of total factor productivity
param gA0:=0.076; #initial growth rate for TFP per 5 years
param deltaA:=0.005; #decline rate of TFP per 5 years
```

```

param gA {t in 0..T}>=0;
let {t in 0..T} gA[t]:=gA0*exp(-deltaA*5*(t));

param A {t in 0..T}>=0;
let A[0]:=A0;
let {t in 1..T} A[t]:=A[t-1]/(1-gA[t-1]);

# Emission parameters, where sigma is the carbon intensity or CO2-output ratio
param gsigma0:=-0.0152; #initial growth of sigma (coninuous per year ),
param deltasigma:=-0.001; #decline rate of decarbonization per period
param ELand0:=2.6; #initial Carbon emissions from land 2015 (GtCO2 per period)
param deltaLand:=0.115; #decline rate of land emissions (per period)
param EInd0:=35.85; #Industrial emissions 2015 (GtCO2 per year)
param Ecum0:=400; #Initial cumulative emissions (GtCO2)
param mu0:=.03; #Initial emissions control rate; under BAU: 0.00
param Lambda0:=0; #Initial abatement costs
param sigma0:=EInd0/(Qgross0*(1-mu0));
#initial sigma (kgCO2 per output 2005 USD in 2010)

param gsigma {t in 0..T};
let gsigma[0]:=gsigma0;
let {t in 1..T} gsigma[t]:=gsigma[t-1]*((1+deltasigma)^5);

param sigma {t in 0..T}>=0;
let sigma[0]:=sigma0;
let {t in 1..T} sigma[t]:=sigma[t-1]*exp(gsigma[t-1]*5);

param ELand {t in 0..T}>=0;
let ELand[0]:=ELand0;
let {t in 1..T} ELand[t]:=ELand[t-1]*(1-deltaLand);

# Carbon cycle
param MAT0=851; # Initial Concentration in atmosphere 2015 (GtC)
param MUP0=460; # Initial Concentration in upper strata 2015 (GtC)
param MLO0=1740; # Initial Concentration in lower strata 2015 (GtC)
param MATEQ=588; # Equilibrium concentration in atmosphere (GtC)
param MUPEQ=360; # Equilibrium concentration in upper strata (GtC)
param MLOEQ=1720; # Equilibrium concentration in lower strata (GtC)

```

```

# Flow parameters (carbon cycle transition matrix)
# correspond to the bXX parameters in Nordhaus)
param phi12:=0.12;
param phi23:=0.007;
param phi11=1-phi12;
param phi21=phi12*MATEQ/MUPEQ;
param phi22=1-phi21-phi23;
param phi32=phi23*MUPEQ/MLOEQ;
param phi33=1-phi32;

# Climate model parameters
param nu:=3.1; # Equilibrium temperature impact (°C per doubling CO2)
param Fex0:=0.5; # 2015 forcings of non-CO2 GHG (Wm-2)
param Fex1:=1.0; # 2100 forcings of non-CO2 GHG (Wm-2)
param TL00:=0.0068; # Initial lower stratum temperature change (°C from 1900)
param TAT0:=0.85; # Initial atmospheric temp change (°C from 1900)
param xi1:=0.1005; # Speed of adjustment parameter for atmospheric temperature
param xi3:=0.088; # Coefficient of heat loss from atmosphere to oceans
param xi4:=0.025; # Coefficient of heat gain by deep oceans
param kappa:=3.6813; # Forcings of equilibrium CO2 doubling (Wm-2)
param xi2=kappa/nu; # climate model parameter
param Fex {t in 0..T}>=0;
let {t in 0..17} Fex[t]:=Fex0+1/17*(Fex1-Fex0)*(t); # external forcing (Wm-2)
let {t in 18..T} Fex[t]:=Fex1;
# external forcing (Wm-2)
# is assumed to be constant and equal to Fex1 from 2100 onward
# see e.g. Traeger (2014, Fig.1)

# climate damage parameters
param Psi:=0.003622;
# 0.00236 damage quadratic term for 25% add on NMD; for 100% add on: 0.003622
param TATlim default 12; # upper bound on atm. temperature change

# abatement cost
param Theta:=2.6; # Exponent of control cost function
param pback0:=550; # Cost of backstop 2010 $ per tCO2 2015
param gback:=0.025; # Initial cost decline backstop cost per period

```

```

param cprice0:=2; # Initial base carbon price (2010$ per tCO2)

param pback {t in 0..T}>=0;
let pback[0]:=pback0;
let {t in 1..T} pback[t]:=pback[t-1]*(1-gback);

param phead {t in 0..T}=pback[t]*sigma[t]/Theta/1000;

# VARIABLES

# capital (trillions 2010 USD)
var K {t in 0..T}>=1;

# Gross output (trillions 2010 USD)
var Qgross {t in 0..T}=A[t]*((L[t]/1000)^(1-gamma))*(K[t]^gamma);

# carbon reservoir atmosphere (GtC)
var MAT {t in 0..T}>=10;

# carbon reservoir upper ocean (GtC)
var MUP {t in 0..T}>=100;

# carbon reservoir lower ocean (GtC)
var MLO {t in 0..T}>=1000;

# total radiative forcing (Wm-2)
var F {t in 0..T}=kappa*((log(MAT[t]/MATEQ))/log(2))+Fex[t];

# atmospheric temperature change (°C from 1750)
var TAT {t in 0..T}>=0, <=TATlim;

# ocean temperature (°C from 1750)
var TLO {t in 0..T}>=-1, <=20;

# damage fraction
var Omega {t in 0..T}=Psi*(TAT[t])^2;

# damages (trillions 2010 USD)

```

```

var damage {t in 0..T}=Omega[t]*Qgross[t];

# emission control
var mu {t in 0..T}>=0;

# abatement costs (fraction of output)
var Lambda {t in 0..T}=Qgross[t]*phead[t]*(mu[t]^Theta);

# industrial emissions
var EInd {t in 0..T}=sigma[t]*Qgross[t]*(1-mu[t]);

# total emissions
var E {t in 0..T};

# maximum cumulative extraction fossil fuels (GtC)
var Ecum {t in 0..T}<=6000;

# Marginal cost of abatement (carbon price)
var cprice {t in 0..T}=pback[t]*mu[t]^(Theta-1);

# output net of damages and abatement (trillions 2010 USD)
var Q {t in 0..T}=(Qgross[t]*(1-Omega[t]))-Lambda[t];

# per capita consumption (1000s 2010 USD)
var c {t in 0..T} >= .1;

# aggregate consumption
var C {t in 0..T} = L[t]*c[t]/1000;

# Investment (trillions 2005 USD)
var I {t in 0..T}>=0;

# utility
var U {t in 0..T} =c[t]^(1-eta)/(1-eta);

# total period utility
var U_period {t in 0..T}=U[t]*R[t];

```



```

# welfare/objective function
var W=sum{t in 0..T} L[t]*U[t]*R[t];

# welfare optimization
maximize objective_function: W;
subject to constr_accounting {t in 0..T}:
C[t]=Q[t]-I[t];
subject to constr_emissions {t in 0..T}:
E[t]=EInd[t]+ELand[t];
subject to constr_capital_dynamics {t in 1..T}:
K[t]=(1-deltaK)^5*K[t-1]+5*I[t-1];
subject to constr_cumulativeemissions {t in 1..T}:
Ecum[t]=Ecum[t-1]+(EInd[t-1]*5/3.666);
subject to constr_atmosphere {t in 1..T}:
MAT[t]=E[t]*(5/3.666)+phi11*MAT[t-1]+phi21*MUP[t-1];
subject to constr_upper_ocean {t in 1..T}:
MUP[t]=phi12*MAT[t-1]+phi22*MUP[t-1]+phi32*MLO[t-1];
subject to constr_lower_ocean {t in 1..T}:
MLO[t]=phi23*MUP[t-1]+phi33*MLO[t-1];
subject to constr_atmospheric_temp {t in 1..T}:
TAT[t]=TAT[t-1]+xi1*((F[t]-xi2*TAT[t-1])-(xi3*(TAT[t-1]-TLO[t-1])));
subject to constr_ocean_temp {t in 1..T}:
TLO[t]=TLO[t-1]+xi4*(TAT[t-1]-TLO[t-1]);

# Initial conditions
subject to initial_capital: K[0] = K0;
subject to initial_Ecum: Ecum[0]=Ecum0;
subject to initial_MAT: MAT[0]=MAT0;
subject to initial_MUP: MUP[0]=MUPO;
subject to initial_MLO: MLO[0]=MLO0;
subject to initial_TLO: TLO[0]=TLO0;
subject to initial_TAT: TAT[0]=TAT0;
subject to initial_control: mu[0]=mu0;
subject to control1 {t in 1..28}: mu[t]<=1;
subject to control2 {t in 29..T}: mu[t]<=1.2; # from 2150
#subject to control_BAU {t in 1..T}: mu[t]=0;

```

## (2) AMPL mod-file, RPE, S&P

```
# DICE 2016R2 with Relative Prices
# To work with the run-file, this mod-file should be named: DICE_2016_RPE.mod
# for central calibration for Section 5 change parameters according to table 3
# PARAMETERS
#Time horizon
param T default 100;

# Preferences
param eta default 1.45; #I-EMUC central calibration: 1.35
param rho default 0.015; #time preference rate; central calibration: 0.011

# relative prices additions
param zeta default -1; #substitution parameter S\&P 2008

#central calibration: -0.11
#for a model run without relative prices:1

param beta default 0.1; #share of non-market good in utility function
param EQbar default 0; #subsistence level of non-market good
#central calibration: 7.77

param cbar default 0; #subsistence level of consumption per capita

# Discount factor
param R {t in 0..T}>=0;
let R[0]:=1;
let {t in 1..T} R[t] := R[t-1]/((1+rho)^5);

#for sensitivity analysis and figure 2/5: param R {t in 0..T}:= 1*exp(-rho*5*t);

# Population and its dynamics
param L0:=7403; #initial world population 2015 (millions)
param gL0:=0.134; #growthrate to calibrate to 2050 pop projection
param L {t in 0..T}>=0;
let L[0]:=L0;
let {t in 1..T} L[t]:=L[t-1]*((11500/L[t-1])^gL0);
```

```

# Technology and its dynamics
param gamma:=0.3; #capital elasticity in production function
param deltaK:=0.1; #depreciation rate on capital (per year)
param Qgross0:=105.5; #Initial world gross output 2015 (trill 2010 USD)
param K0:=223; #initial capital value 2015 (trillions 2010 USD)
param A0:=5.115; #initial level of total factor productivity
param gA0 :=0.076; #initial growth rate for TFP per 5 years
param deltaA default 0.005; #decline rate of TFP per 5 years
#central calibration: 0.005

param gA {t in 0..T} := gA0*exp(-deltaA*5*t); # growth rate for TFP per period

param A {t in 0..T}>=0;
let A[0]:=A0;
let {t in 1..T} A[t]:=A[t-1]/(1-gA[t-1]);

# Emission parameters
param gsigma0:=-0.0152; #initial growth of sigma (coninuous per year )
param deltasigma:=-0.001; #decline rate of decarbonization per period
param ELand0:=2.6; #initial Carbon emissions from land 2015 (GtCO2 per period)
param deltaLand:=0.115; #decline rate of land emissions (per period)
param EInd0:=35.85; #Industrial emissions 2015 (GtCO2 per year)
param Ecum0:=400; #Initial cumulative emissions (GtCO2)
param mu0:=.03; #Initial emissions control rate for base year 2010
param Lambda0:=0; #Initial abatement costs
param sigma0:=EInd0/(Qgross0*(1-mu0));#initial sigma
#(kgCO2 per output 2005 USD in 2010)

param gsigma {t in 0..T};
let gsigma[0]:=gsigma0;
let {t in 1..T} gsigma[t]:=gsigma[t-1]*((1+deltasigma)^5);

param sigma {t in 0..T}>=0;
let sigma[0]:=sigma0;
let {t in 1..T} sigma[t]:=sigma[t-1]*exp(gsigma[t-1]*5);

param ELand {t in 0..T}>=0;
let ELand[0]:=ELand0;

```

```

let {t in 1..T} ELand[t]:=ELand [t-1]*(1-deltaLand);

# Carbon cycle
param MAT0=851; # Initial Concentration in atmosphere 2015 (GtC)
param MUP0:=460; # Initial Concentration in upper strata 2015 (GtC)
param MLO0:=1740; # Initial Concentration in lower strata 2015 (GtC)
param MATEQ:=588; # Equilibrium concentration in atmosphere
#(pre-industrial atmos. carbon) (GtC)
param MUPEQ:=360; # Equilibrium concentration in upper strata (GtC)
param MLOEQ:=1720; # Equilibrium concentration in lower strata (GtC)

# Flow parameters
param phi12:=0.12;
param phi23:=0.007;
param phi11=1-phi12;
param phi21=phi12*MATEQ/MUPEQ;
param phi22=1-phi21-phi23;
param phi32=phi23*MUPEQ/MLOEQ;
param phi33=1-phi32;

# Climate model parameters
param nu:=3.1; # Equilibrium temperature impact (°C per doubling CO2)
param Fex0:=0.5; # 2015 forcings of non-CO2 GHG (Wm-2)
param Fex1:=1.0; # 2100 forcings of non-CO2 GHG (Wm-2)
param TLO0:=0.0068; # Initial lower stratum temperature change (°C from 1900)
param TAT0:=0.85; # Initial atmospheric temp change (°C from 1900)
param xi1:=0.1005; # Speed of adjustment parameter for atmospheric temperature
param xi3:=0.088; # Coefficient of heat loss from atmosphere to oceans
param xi4:=0.025; # Coefficient of heat gain by deep oceans
param kappa:=3.6813; # Forcings of equilibrium CO2 doubling (Wm-2)
param xi2=kappa/nu; # climate model parameter

# external forcing (Wm-2)
#assumed to be constant and equal to Fex1 from 2100 onward,
#see e.g. Traeger (2014, Fig.1)
param Fex {t in 0..T}>=0;
let {t in 0..18} Fex[t]:=Fex0+1/18*(Fex1-Fex0)*(t);
let {t in 19..T} Fex[t]:=Fex1;

```

```

# Climate damage parameters
param Psi default 0.00181; # market damage term without 25% adjustment
# damage quadratic term with 25% adjustment is 0.00236
param MD default 0.0163;
# market damages for 3°C warming above preindustrial according to Nordhaus (2017)
param NMD default 0.0163;
# corresponds to 100% NMD add on; with 25% add on 0.00494
# central calibration: 0.0165
param TD=MD+NMD; # total climate damages

param TATlim default 12; # upper bound on atm. temperature change

# Abatement cost
param Theta:=2.6; # Exponent of control cost function
param pback0:=550; # Cost of backstop 2010 $ per tCO2 2015
param gback:=0.025; # Initial cost decline backstop cost per period

param pback {t in 0..T}>=0;
let pback[0]:=pback0;
let {t in 1..T} pback[t]:=pback[t-1]*(1-gback);

param phead {t in 0..T}=pback[t]*sigma[t]/Theta/1000;

# VARIABLES
# capital (trillions 2010 USD)
var K {t in 0..T}>=1;

# Gross output (trillions 2010 USD)
var Qgross {t in 0..T}=A[t]*((L[t]/1000)^(1-gamma))*(K[t]^gamma);

# carbon reservoir atmosphere (GtC)
var MAT {t in 0..T}>=10;

# carbon reservoir upper ocean (GtC)
var MUP {t in 0..T}>=100;

# carbon reservoir lower ocean (GtC)

```

```

var MLO {t in 0..T}>=1000;

# total radiative forcing (Wm-2)
var F {t in 0..T}=kappa*((log(MAT[t]/MATEQ))/log(2))+Fex[t];

# atmospheric temperature change (°C from 1750)
var TAT {t in 0..T}>=0, <=TATlim;

# ocean temperature (°C from 1750)
var TLO {t in 0..T}>=-1, <=20;

# damage fraction
var Omega {t in 0..T}=Psi*(TAT[t])^2;

# damages (trillions 2010 USD)
var damage {t in 0..T}=Omega[t]*Qgross[t];

# emission control
var mu {t in 0..T}>=0;

# abatement costs (fraction of output)
var Lambda {t in 0..T}=Qgross[t]*phead[t]*(mu[t]^Theta);

# industrial emissions
var EInd {t in 0..T}=sigma[t]*Qgross[t]*(1-mu[t]);

# total emissions
var E {t in 0..T};

# maximum cumulative extraction fossil fuels (GtC)
var Ecum {t in 0..T}<=6000;

# Marginal cost of abatement (carbon price)
var cprice {t in 0..T}=pback[t]*mu[t]^(Theta-1);

# output net of damages and abatement(trillions 2010 USD)
var Q {t in 0..T}=(Qgross[t]*(1-Omega[t]))-Lambda[t];

```

```

# per capita consumption (1000s 2010 USD)
var c {t in 0..T} >= .1;

# aggregate consumption
var C {t in 0..T} = L[t]*c[t]/1000;

# Investment(trillions 2005 USD)
var I {t in 0..T}>=0;

# non-market good
var EQ {t in 0..T}>=0.0000001 <=1000;

# Non-market damages scaling parameter including subsistence requirement

# including sub
var a {t in 0..T} =(1/(nu^2))*(EQ[0]*(EQbar+((EQ[0]-EQbar)^(zeta)
+((1-beta)/beta)*(((1-TD)*C[0])^(zeta)-((1-MD)*C[0])^(zeta)))^(1/zeta))^(1-zeta));

# growth rate of market good

var g_C {t in 0..T-1} = (C[t+1]-C[t])/C[t];

# growth rate of non market good

var g_EQ {t in 0..T-1} = ((EQ[t+1]-EQ[t])/EQ[t]);

# relative price effect

var RPE {t in 0..T-1} =(1-zeta)*(g_C[t]-((EQ[t]/(EQ[t]-EQbar))*g_EQ[t]));

# utility

var U {t in 0..T}= (((1-beta)*(c[t])^(zeta)+
beta*((EQ[t]-EQbar)*1000/L[t])^(zeta))^(1-zeta))/(1-zeta);

# welfare/objective function
var W=sum{t in 0..T} L[t]*U[t]*R[t];

```

```

maximize objective_function: W;
subject to initial_consumption: c[0]=10.4893;
subject to constr_accounting {t in 0..T}:
C[t]=Q[t]-I[t];
subject to constr_emissions {t in 0..T}:
E[t]=EInd[t]+ELand[t];
subject to constr_capital_dynamics {t in 1..T}:
K[t]=(1-deltaK)^5*K[t-1]+5*I[t-1];
subject to constr_cumulativeemissions {t in 1..T}:
Ecum[t]=Ecum[t-1]+(EInd[t-1]*5/3.666);
subject to constr_atmosphere {t in 1..T}:
MAT[t]=E[t]*(5/3.666)+phi11*MAT[t-1]+phi21*MUP[t-1];
subject to constr_upper_ocean {t in 1..T}:
MUP[t]=phi12*MAT[t-1]+phi22*MUP[t-1]+phi32*MLO[t-1];
subject to constr_lower_ocean {t in 1..T}:
MLO[t]=phi23*MUP[t-1]+phi33*MLO[t-1];
subject to constr_atmospheric_temp {t in 1..T}:
TAT[t]=TAT[t-1]+xi1*((F[t]-xi2*TAT[t-1])-(xi3*(TAT[t-1]-TLO[t-1])));
subject to constr_ocean_temp {t in 1..T}:
TLO[t]=TLO[t-1]+xi4*(TAT[t-1]-TLO[t-1]);

# Initial conditions
subject to initial_capital: K[0] = K0;
subject to initial_Ecum: Ecum[0]=Ecum0;
subject to initial_MAT: MAT[0]=MAT0;
subject to initial_MUP: MUP[0]=MUP0;
subject to initial_MLO: MLO[0]=MLO0;
subject to initial_TLO: TLO[0]=TLO0;
subject to initial_TAT: TAT[0]=TAT0;
subject to initial_control: mu[0]=mu0;
subject to control1 {t in 1..28}: mu[t]<=1;
subject to control2 {t in 29..T}: mu[t]<=1.2; # from 2150
subject to initial_EQ: EQ[0]=C[0];
subject to constr_EQ {t in 1..T}: EQ[t]=(EQ[0]/(1+a[t]*(TAT[t]^2)));

```



### (3) AMPL run-file

```
reset;
model DICE_2016_RPE.mod; # add a "#" in a S\&P-RPE run
#model DICE_Nordhaus_2016.mod; # delete "#" in a Nordhaus run
option solver knitroampl;
solve;
# Produce overview of results in a csv format
# change file name to "Results_Figure1_Nordhaus.csv" during the Nordhaus run
# change file name to "Results_Figure1_RPE-SP.csv" during the RPE-SP run

for {i in 0..T-1}
{printf "%f\t", i>Results_Figure1_RPE-SP.csv;
printf "%f\t", (((RPE[i]+1)^(1/5))-1)*100>Results_Figure1_RPE-SP.csv;
# delete this RPE line during Nordhaus run
printf "%f\t", EInd[i]> Results_Figure1_RPE-SP.csv;
printf "%f\t", TAT[i]>Results_Figure1_RPE-SP.csv;
printf "%f\n", -1000*constr_emissions[i]/constr_accounting[i]>
Results_Figure1_RPE-SP.csv;}
```

### A.10.2 AMPL code to produce Figure 2

```
# use the same mod files as for figure 1 with the following changes
# change equation for time preference rate to
# param R {t in 0..T}:= 1*exp(-rho*5*t);
# set the substitution parameter zeta equal to 1 for the run without RPE
# run the following AMPL run file
# change the file name of the csv-file for each run as preferred, e.g.:
# Nordhaus, Sterner and Persson (SP), without relative prices
reset;
model DICE_2016_RPE.mod; # add a "#" in a Nordhaus run
#model DICE_Nordhaus_2016.mod; # delete "#" in a Nordhaus run
option solver knitroampl;
solve;
# Produce sensitivity analysis in csv format
for {i in 0.000001 .. 0.032 by 0.001}
{ let rho:=i;
solve;
printf "%.5f\t", rho>Results_Figure2_SP.csv;
printf "%.5f\n", max {t in 0..T} TAT[t]>Results_Figure2_SP.csv;
}
```

### A.10.3 AMPL code for Figure 3 and Figure 4

```
# use the AMPL mod file RPE, S\&P (as for figure 1 and 2)
# change equation for time preference rate to param R {t in 0..T}:= 1*exp(-rho*5*t);
# run the following AMPL run-file
# note that for the decline rate of TFP deltaA the sensitivity analysis
# needs to be done manually, i.e.
# change deltaA in the mod-file from 0 to 0.05 in some steps
# solve the model each time
# print the RPE in 2020 and 2100 for every deltaA similar to other variables

reset;
model DICE_2016_RPE.mod;
option solver knitroampl;
solve;

# Produce sensitivity analysis in csv format
for {i in -4 .. 1.1 by 0.03}
{let zeta:=i;
solve;
printf "%.5f\t", i>Results_Figure3_Theta.csv;
printf "%.5f\t", (((RPE[1]+1)^(1/5))-1)*100>Results_Figure3_Theta.csv;
printf "%.5f\n", (((RPE[17]+1)^(1/5))-1)*100>Results_Figure3_Theta.csv;
}

reset;
model DICE_2016_RPE.mod;
option solver knitroampl;
solve;

for {i in 0.000 .. 0.11 by 0.001}
{let NMD:=i;
solve;
printf "%.5f\t", i>Results_Figure3_NMD.csv;
printf "%.5f\t", (((RPE[1]+1)^(1/5))-1)*100>Results_Figure3_NMD.csv;
printf "%.5f\n", (((RPE[17]+1)^(1/5))-1)*100>Results_Figure3_NMD.csv;
}
```

```

reset;
model DICE_2016_RPE.mod;
option solver knitroampl;
solve;

for {i in 0 .. 40 by 0.5}
{ let EQbar:=i;
solve;
printf "%.5f\t", i>Results_Figure3_Sub.csv;
printf "%.5f\t", (((RPE[1]+1)^(1/5))-1)*100>Results_Figure3_Sub.csv;
printf "%.5f\n", (((RPE[17]+1)^(1/5))-1)*100>Results_Figure3_Sub.csv;
}

reset;
model DICE_2016_RPE.mod;
option solver knitroampl;
solve;

for {i in 0.000001 .. 0.085 by 0.001}
{let rho:=i;
solve;
printf "%.5f\t", rho>Results_Figure4_delta.csv;
printf "%.5f\t", (((RPE[1]+1)^(1/5))-1)*100>Results_Figure4_delta.csv;
printf "%.5f\n", (((RPE[17]+1)^(1/5))-1)*100>Results_Figure4_delta.csv;
}

reset;
model DICE_2016_RPE.mod;
option solver knitroampl;
solve;

for {i in 0.0001 .. 5.2 by 0.02}
{let eta:=i;
solve;
printf "%.5f\t", eta>Results_Figure4_eta.csv;
printf "%.5f\t", (((RPE[1]+1)^(1/5))-1)*100>Results_Figure4_eta.csv;
printf "%.5f\n", (((RPE[17]+1)^(1/5))-1)*100>Results_Figure4_eta.csv;}

```

#### A.10.4 AMPL code for figure 5

##### (1) AMPL dat-file, Plausible Ranges

```
# this is the random data generated from raw data by Drupp et al. (2018)
# save this file as random_delta_eta.dat to be compatible with the run-file
```

```
param nruns:=1000;
```

```
param rhos:=
```

```
1 0.015
```

```
2 0.00000001
```

```
3 0.02
```

```
4 0.01
```

```
5 0.03
```

```
6 0.03
```

```
7 0.001
```

```
8 0.005
```

```
9 0.00000001
```

```
10 0.00000001
```

```
11 0.01
```

```
12 0.00000001
```

```
13 0.001
```

```
14 0.001
```

```
15 0.005
```

```
16 0.01
```

```
17 0.00000001
```

```
18 0.02
```

```
19 0.015
```

```
20 0.00000001
```

```
21 0.04
```

```
22 0.00000001
```

```
23 0.00000001
```

```
24 0.00000001
```

```
25 0.03
```

```
26 0.00000001
```

```
27 0.00000001
```

```
28 0.001
```

```
29 0.02
```

30 0.04  
31 0.02  
32 0.00000001  
33 0.01  
34 0.00000001  
35 0.01  
36 0.03  
37 0.003  
38 0.03  
39 0.02  
40 0.00000001  
41 0.00000001  
42 0.02  
43 0.001  
44 0.02  
45 0.025  
46 0.005  
47 0.005  
48 0.01  
49 0.001  
50 0.01  
51 0.03  
52 0.02  
53 0.015  
54 0.01  
55 0.01  
56 0.01  
57 0.002  
58 0.01  
59 0.00000001  
60 0.00000001  
61 0.001  
62 0.025  
63 0.00000001  
64 0.00001  
65 0.02  
66 0.005  
67 0.04

68 0.01  
69 0.01  
70 0.01  
71 0.001  
72 0.07  
73 0.005  
74 0.04  
75 0.02  
76 0.04  
77 0.005  
78 0.00000001  
79 0.01  
80 0.01  
81 0.01  
82 0.02  
83 0.02  
84 0.00000001  
85 0.01  
86 0.00000001  
87 0.02  
88 0.001  
89 0.02  
90 0.03  
91 0.06  
92 0.00000001  
93 0.02  
94 0.01  
95 0.005  
96 0.001  
97 0.001  
98 0.03  
99 0.06  
100 0.06  
101 0.01  
102 0.00000001  
103 0.00000001  
104 0.03  
105 0.00001

106 0.00000001  
107 0.02  
108 0.005  
109 0.02  
110 0.00000001  
111 0.00000001  
112 0.00000001  
113 0.00000001  
114 0.008  
115 0.00001  
116 0.06  
117 0.00000001  
118 0.02  
119 0.015  
120 0.02  
121 0.005  
122 0.005  
123 0.005  
124 0.00000001  
125 0.01  
126 0.001  
127 0.00001  
128 0.01  
129 0.01  
130 0.005  
131 0.00000001  
132 0.01  
133 0.005  
134 0.08  
135 0.001  
136 0.00000001  
137 0.00000001  
138 0.00000001  
139 0.00000001  
140 0.00000001  
141 0.001  
142 0.025  
143 0.00000001



144 0.01  
145 0.01  
146 0.003  
147 0.00000001  
148 0.02  
149 0.015  
150 0.015  
151 0.00001  
152 0.04  
153 0.01  
154 0.00000001  
155 0.00000001  
156 0.01  
157 0.00000001  
158 0.00000001  
159 0.03  
160 0.00005  
161 0.00000001  
162 0.001  
163 0.01  
164 0.01  
165 0.00000001  
166 0.001  
167 0.03  
168 0.02  
169 0.00001  
170 0.003  
171 0.01  
172 0.00000001  
173 0.005  
174 0.00000001  
175 0.005  
176 0.0025  
177 0.07  
178 0.01  
179 0.01  
180 0.02  
181 0.02

182 0.03  
183 0.001  
184 0.00000001  
185 0.02  
186 0.0001  
187 0.00000001  
188 0.005  
189 0.00000001  
190 0.00000001  
191 0.00000001  
192 0.01  
193 0.00001  
194 0.005  
195 0.01  
196 0.00000001  
197 0.00000001  
198 0.04  
199 0.01  
200 0.015  
201 0.01  
202 0.005  
203 0.002  
204 0.00000001  
205 0.001  
206 0.02  
207 0.00000001  
208 0.005  
209 0.02  
210 0.005  
211 0.001  
212 0.02  
213 0.00000001  
214 0.02  
215 0.01  
216 0.001  
217 0.00000001  
218 0.08  
219 0.00000001

220 0.02  
221 0.00000001  
222 0.04  
223 0.02  
224 0.00000001  
225 0.025  
226 0.02  
227 0.005  
228 0.00000001  
229 0.00000001  
230 0.08  
231 0.005  
232 0.03  
233 0.00000001  
234 0.001  
235 0.025  
236 0.00000001  
237 0.04  
238 0.005  
239 0.03  
240 0.01  
241 0.015  
242 0.03  
243 0.02  
244 0.01  
245 0.02  
246 0.02  
247 0.001  
248 0.005  
249 0.02  
250 0.01  
251 0.01  
252 0.01  
253 0.02  
254 0.01  
255 0.01  
256 0.01  
257 0.02

258 0.01  
259 0.00000001  
260 0.001  
261 0.008  
262 0.001  
263 0.01  
264 0.00000001  
265 0.00000001  
266 0.02  
267 0.00000001  
268 0.005  
269 0.001  
270 0.00000001  
271 0.03  
272 0.00000001  
273 0.005  
274 0.01  
275 0.003  
276 0.025  
277 0.01  
278 0.06  
279 0.01  
280 0.00000001  
281 0.0025  
282 0.00000001  
283 0.00000001  
284 0.02  
285 0.00000001  
286 0.00000001  
287 0.00000001  
288 0.01  
289 0.025  
290 0.001  
291 0.001  
292 0.02  
293 0.005  
294 0.00000001  
295 0.001

296 0.00000001  
297 0.001  
298 0.012  
299 0.01  
300 0.00000001  
301 0.00000001  
302 0.00000001  
303 0.005  
304 0.00000001  
305 0.02  
306 0.01  
307 0.00000001  
308 0.005  
309 0.00000001  
310 0.00000001  
311 0.00000001  
312 0.001  
313 0.005  
314 0.02  
315 0.03  
316 0.02  
317 0.00000001  
318 0.03  
319 0.01  
320 0.03  
321 0.025  
322 0.02  
323 0.00000001  
324 0.005  
325 0.01  
326 0.005  
327 0.005  
328 0.01  
329 0.00000001  
330 0.00000001  
331 0.00000001  
332 0.02  
333 0.01

334 0.005  
335 0.03  
336 0.00000001  
337 0.00000001  
338 0.005  
339 0.00000001  
340 0.01  
341 0.001  
342 0.005  
343 0.00000001  
344 0.02  
345 0.03  
346 0.00000001  
347 0.005  
348 0.01  
349 0.001  
350 0.02  
351 0.00000001  
352 0.03  
353 0.01  
354 0.01  
355 0.00000001  
356 0.01  
357 0.05  
358 0.02  
359 0.008  
360 0.008  
361 0.005  
362 0.02  
363 0.03  
364 0.03  
365 0.001  
366 0.02  
367 0.02  
368 0.00000001  
369 0.02  
370 0.00000001  
371 0.02

372 0.003  
373 0.001  
374 0.00000001  
375 0.00001  
376 0.02  
377 0.0001  
378 0.00000001  
379 0.06  
380 0.01  
381 0.00000001  
382 0.005  
383 0.00000001  
384 0.002  
385 0.01  
386 0.03  
387 0.02  
388 0.01  
389 0.01  
390 0.08  
391 0.00000001  
392 0.001  
393 0.00000001  
394 0.00000001  
395 0.01  
396 0.005  
397 0.025  
398 0.00000001  
399 0.00000001  
400 0.001  
401 0.00000001  
402 0.00000001  
403 0.00000001  
404 0.06  
405 0.00000001  
406 0.003  
407 0.00000001  
408 0.005  
409 0.02

410 0.00000001  
411 0.005  
412 0.02  
413 0.00001  
414 0.00000001  
415 0.001  
416 0.00000001  
417 0.001  
418 0.01  
419 0.00000001  
420 0.01  
421 0.005  
422 0.00000001  
423 0.015  
424 0.01  
425 0.01  
426 0.02  
427 0.00000001  
428 0.001  
429 0.005  
430 0.02  
431 0.01  
432 0.003  
433 0.002  
434 0.01  
435 0.00000001  
436 0.01  
437 0.01  
438 0.01  
439 0.06  
440 0.01  
441 0.001  
442 0.01  
443 0.012  
444 0.002  
445 0.03  
446 0.00000001  
447 0.00000001



448 0.00001  
449 0.001  
450 0.02  
451 0.03  
452 0.02  
453 0.008  
454 0.04  
455 0.00000001  
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457 0.01  
458 0.01  
459 0.01  
460 0.001  
461 0.001  
462 0.00000001  
463 0.01  
464 0.01  
465 0.02  
466 0.00000001  
467 0.00000001  
468 0.01  
469 0.001  
470 0.015  
471 0.00000001  
472 0.005  
473 0.001  
474 0.00000001  
475 0.001  
476 0.00000001  
477 0.001  
478 0.00000001  
479 0.002  
480 0.0001  
481 0.005  
482 0.04  
483 0.001  
484 0.003  
485 0.002

486 0.001  
487 0.01  
488 0.005  
489 0.01  
490 0.00000001  
491 0.03  
492 0.02  
493 0.025  
494 0.005  
495 0.01  
496 0.005  
497 0.02  
498 0.0001  
499 0.00000001  
500 0.025  
501 0.00000001  
502 0.02  
503 0.005  
504 0.005  
505 0.025  
506 0.001  
507 0.02  
508 0.02  
509 0.005  
510 0.001  
511 0.02  
512 0.02  
513 0.0005  
514 0.00000001  
515 0.005  
516 0.015  
517 0.002  
518 0.02  
519 0.01  
520 0.00000001  
521 0.00000001  
522 0.01  
523 0.03

524 0.00000001  
525 0.01  
526 0.001  
527 0.02  
528 0.01  
529 0.07  
530 0.01  
531 0.00000001  
532 0.00000001  
533 0.00000001  
534 0.07  
535 0.05  
536 0.025  
537 0.02  
538 0.01  
539 0.00000001  
540 0.02  
541 0.005  
542 0.005  
543 0.03  
544 0.00001  
545 0.00000001  
546 0.00000001  
547 0.00000001  
548 0.02  
549 0.00000001  
550 0.025  
551 0.001  
552 0.00000001  
553 0.001  
554 0.00000001  
555 0.00000001  
556 0.001  
557 0.01  
558 0.01  
559 0.001  
560 0.00000001  
561 0.005

562 0.01  
563 0.02  
564 0.03  
565 0.00000001  
566 0.00000001  
567 0.03  
568 0.00000001  
569 0.001  
570 0.0001  
571 0.00000001  
572 0.001  
573 0.0005  
574 0.00000001  
575 0.00000001  
576 0.005  
577 0.008  
578 0.001  
579 0.01  
580 0.001  
581 0.02  
582 0.02  
583 0.00005  
584 0.00000001  
585 0.001  
586 0.005  
587 0.001  
588 0.01  
589 0.001  
590 0.00000001  
591 0.02  
592 0.005  
593 0.02  
594 0.005  
595 0.00000001  
596 0.01  
597 0.005  
598 0.01  
599 0.0005

600 0.015  
601 0.025  
602 0.00000001  
603 0.02  
604 0.01  
605 0.025  
606 0.00000001  
607 0.02  
608 0.02  
609 0.02  
610 0.00000001  
611 0.01  
612 0.025  
613 0.001  
614 0.00001  
615 0.01  
616 0.02  
617 0.02  
618 0.00000001  
619 0.02  
620 0.01  
621 0.00000001  
622 0.03  
623 0.01  
624 0.008  
625 0.001  
626 0.015  
627 0.001  
628 0.00000001  
629 0.00000001  
630 0.00000001  
631 0.00000001  
632 0.02  
633 0.001  
634 0.02  
635 0.03  
636 0.005  
637 0.005

638 0.00000001  
639 0.001  
640 0.08  
641 0.0005  
642 0.00000001  
643 0.005  
644 0.02  
645 0.001  
646 0.00000001  
647 0.02  
648 0.01  
649 0.00000001  
650 0.01  
651 0.00000001  
652 0.03  
653 0.06  
654 0.01  
655 0.00005  
656 0.005  
657 0.00000001  
658 0.002  
659 0.01  
660 0.025  
661 0.00005  
662 0.015  
663 0.00000001  
664 0.02  
665 0.015  
666 0.0025  
667 0.01  
668 0.002  
669 0.001  
670 0.015  
671 0.02  
672 0.01  
673 0.01  
674 0.01  
675 0.001

676 0.00000001  
677 0.01  
678 0.02  
679 0.00000001  
680 0.02  
681 0.005  
682 0.07  
683 0.002  
684 0.07  
685 0.005  
686 0.015  
687 0.00001  
688 0.05  
689 0.03  
690 0.00000001  
691 0.00000001  
692 0.00000001  
693 0.01  
694 0.005  
695 0.0005  
696 0.03  
697 0.001  
698 0.01  
699 0.00000001  
700 0.005  
701 0.01  
702 0.00000001  
703 0.01  
704 0.005  
705 0.00000001  
706 0.02  
707 0.01  
708 0.00000001  
709 0.001  
710 0.01  
711 0.00000001  
712 0.00000001  
713 0.01

714 0.01  
715 0.01  
716 0.001  
717 0.02  
718 0.0025  
719 0.025  
720 0.00000001  
721 0.02  
722 0.001  
723 0.001  
724 0.001  
725 0.00000001  
726 0.02  
727 0.00001  
728 0.003  
729 0.01  
730 0.03  
731 0.06  
732 0.01  
733 0.00000001  
734 0.03  
735 0.001  
736 0.00000001  
737 0.00000001  
738 0.00000001  
739 0.01  
740 0.005  
741 0.025  
742 0.00000001  
743 0.01  
744 0.05  
745 0.02  
746 0.01  
747 0.01  
748 0.005  
749 0.01  
750 0.002  
751 0.00001



752 0.005  
753 0.005  
754 0.05  
755 0.00000001  
756 0.00000001  
757 0.00000001  
758 0.02  
759 0.001  
760 0.01  
761 0.025  
762 0.025  
763 0.005  
764 0.00000001  
765 0.02  
766 0.00000001  
767 0.00000001  
768 0.02  
769 0.00000001  
770 0.00000001  
771 0.00000001  
772 0.01  
773 0.025  
774 0.001  
775 0.01  
776 0.00000001  
777 0.00000001  
778 0.01  
779 0.01  
780 0.08  
781 0.005  
782 0.04  
783 0.08  
784 0.00000001  
785 0.02  
786 0.01  
787 0.00000001  
788 0.01  
789 0.00000001

790 0.015  
791 0.008  
792 0.015  
793 0.008  
794 0.02  
795 0.001  
796 0.00000001  
797 0.00000001  
798 0.03  
799 0.03  
800 0.02  
801 0.00001  
802 0.00000001  
803 0.001  
804 0.005  
805 0.01  
806 0.001  
807 0.05  
808 0.001  
809 0.01  
810 0.00000001  
811 0.01  
812 0.01  
813 0.001  
814 0.00000001  
815 0.015  
816 0.001  
817 0.005  
818 0.00000001  
819 0.03  
820 0.001  
821 0.00000001  
822 0.03  
823 0.01  
824 0.005  
825 0.02  
826 0.00000001  
827 0.00000001

828 0.02  
829 0.00000001  
830 0.02  
831 0.00000001  
832 0.001  
833 0.08  
834 0.003  
835 0.03  
836 0.07  
837 0.001  
838 0.03  
839 0.02  
840 0.01  
841 0.01  
842 0.00000001  
843 0.00000001  
844 0.04  
845 0.02  
846 0.00001  
847 0.005  
848 0.00000001  
849 0.005  
850 0.0001  
851 0.0001  
852 0.03  
853 0.00001  
854 0.02  
855 0.01  
856 0.001  
857 0.00000001  
858 0.005  
859 0.06  
860 0.01  
861 0.02  
862 0.00000001  
863 0.001  
864 0.00000001  
865 0.00000001

866 0.02  
867 0.03  
868 0.001  
869 0.00000001  
870 0.01  
871 0.00000001  
872 0.005  
873 0.02  
874 0.00000001  
875 0.08  
876 0.005  
877 0.005  
878 0.01  
879 0.001  
880 0.003  
881 0.02  
882 0.012  
883 0.02  
884 0.01  
885 0.01  
886 0.00000001  
887 0.03  
888 0.00000001  
889 0.001  
890 0.03  
891 0.02  
892 0.00000001  
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940 1.00000010  
941 1.00000010  
942 1.00000010  
943 0.00000100  
944 2.00000000  
945 2.00000000  
946 2.00000000  
947 0.20000000  
948 1.50000000  
949 1.30000000  
950 0.00000100  
951 5.00000000  
952 2.50000000  
953 2.00000000  
954 2.00000000  
955 1.00000010  
956 0.20000000  
957 2.00000000  
958 1.00000010  
959 1.00000010  
960 1.00000010  
961 0.10000000  
962 1.50000000  
963 1.00000010  
964 1.00000010  
965 0.80000000



966 2.00000000  
967 1.80000000  
968 0.50000000  
969 2.00000000  
970 1.00000010  
971 0.20000000  
972 1.00000010  
973 2.00000000  
974 1.00000010  
975 1.50000000  
976 1.80000000  
977 1.00000010  
978 1.00000010  
979 5.00000000  
980 2.00000000  
981 0.50000000  
982 2.00000000  
983 1.50000000  
984 3.00000000  
985 1.00000010  
986 1.00000010  
987 2.00000000  
988 2.00000000  
989 1.00000010  
990 1.00000010  
991 1.00000010  
992 1.00000010  
993 0.50000000  
994 0.20000000  
995 2.00000000  
996 0.50000000  
997 0.70000000  
998 1.50000000  
999 1.00000010  
1000 2.00000000;

## (2) AMPL mod-file, Plausible Ranges

# save this as DICE\_2016\_RPE\_MonteCarlo.mod to be compatible with the run-file

# PARAMETERS

#Time horizon

param T default 100;

# number of runs for Monte Carlo analysis

param nruns;

# Preferences

param etas {1..nruns}; #I-EMUC consumption

param eta;

param rhos {1..nruns}; #time preference rate

param rho;

# relative prices additions

param zeta=min(Normal(-0.11,0.17),1);

#substitution parameter, first value is central calibration

param beta default 0.1; #share of non-market good in utility function

param EQbar=max(Normal(7.77,3.96),0); #subsistence level of non-market good

param cbar default 0; #subsistence level of consumption per capita

# Discount factor

param R {t in 0..T} = 1\*exp(-rho\*5\*t);

# Population and its dynamics

param L0:=7403; #initial world population 2015 (millions)

param gL0:=0.134; #growthrate to calibrate to 2050 pop projection

param L {t in 0..T}>=0;

let L[0]:=L0;

let {t in 1..T} L[t]:=L[t-1]\*((11500/L[t-1])^gL0);

# Technology and its dynamics

param gamma:=0.3; #capital elasticity in production function

param deltaK:=0.1; #depreciation rate on capital (per year)

```

param Qgross0:=105.5; #Initial world gross output 2015 (trill 2010 USD)
param K0:=223; #initial capital value 2015 (trillions 2010 USD)
param A0:=5.115; #initial level of total factor productivity
param gA0 :=0.076; #initial growth rate for TFP per 5 years
param deltaA=max(Normal(0.005,0.00255),0); #decline rate of TFP per 5 years

param gA {t in 0..T} := gA0*exp(-deltaA*5*t); # growth rate for TFP per period

param A {t in 0..T}>=0;
let A[0]:=A0;
let {t in 1..T} A[t]:=A[t-1]/(1-gA[t-1]);

# Emission parameters
param gsigma0:=-0.0152; #initial growth of sigma (coninuous per year )
param deltasigma:=-0.001; #decline rate of decarbonization per period
param ELand0:=2.6; #initial Carbon emissions from land 2015 (GtCO2 per period)
param deltaLand:=0.115; #decline rate of land emissions (per period)
param EInd0:=35.85; #Industrial emissions 2015 (GtCO2 per year)
param Ecum0:=400; #Initial cumulative emissions (GtCO2)
param mu0:=.03; #Initial emissions control rate for base year 2015
param Lambda0:=0; #Initial abatement costs
param sigma0:=EInd0/(Qgross0*(1-mu0));
#initial sigma (kgCO2 per output 2005 USD in 2010)

param gsigma {t in 0..T};
let gsigma[0]:=gsigma0;
let {t in 1..T} gsigma[t]:=gsigma[t-1]*((1+deltasigma)^5);

param sigma {t in 0..T}>=0;
let sigma[0]:=sigma0;
let {t in 1..T} sigma[t]:=sigma[t-1]*exp(gsigma[t-1]*5);

param ELand {t in 0..T}>=0;
let ELand[0]:=ELand0;
let {t in 1..T} ELand[t]:=ELand [t-1]*(1-deltaLand);

# Carbon cycle
param MAT0=851; # Initial Concentration in atmosphere 2015 (GtC)

```

```

param MUP0:=460; # Initial Concentration in upper strata 2015 (GtC)
param ML00:=1740; # Initial Concentration in lower strata 2015 (GtC)
param MATEQ:=588; # Equilibrium concentration in atmosphere (GtC)
param MUPEQ:=360; # Equilibrium concentration in upper strata (GtC)
param MLOEQ:=1720; # Equilibrium concentration in lower strata (GtC)

# Flow parameters (carbon cycle transition matrix)

param phi12:=0.12;
param phi23:=0.007;
param phi11=1-phi12;
param phi21=phi12*MATEQ/MUPEQ;
param phi22=1-phi21-phi23;
param phi32=phi23*MUPEQ/MLOEQ;
param phi33=1-phi32;

# Climate model parameters
param nu:=3.1; # Equilibrium temperature impact (°C per doubling CO2)
param Fex0:=0.5; # 2015 forcings of non-CO2 GHG (Wm-2)
param Fex1:=1.0; # 2100 forcings of non-CO2 GHG (Wm-2)
param TL00:=0.0068; # Initial lower stratum temperature change (°C from 1900)
param TAT0:=0.85; # Initial atmospheric temp change (°C from 1900)
param xi1:=0.1005; # Speed of adjustment parameter for atmospheric temperature
param xi3:=0.088; # Coefficient of heat loss from atmosphere to oceans
param xi4:=0.025; # Coefficient of heat gain by deep oceans
param kappa:=3.6813; # Forcings of equilibrium CO2 doubling (Wm-2)
param xi2=kappa/nu; # climate model parameter

# external forcing (Wm-2)
# is assumed to be constant and equal to Fex1 from 2100 onward
# see e.g. Traeger (2014, Fig.1)
param Fex {t in 0..T}>=0;
let {t in 0..18} Fex[t]:=Fex0+1/18*(Fex1-Fex0)*(t);
let {t in 19..T} Fex[t]:=Fex1;

# Climate damage parameters
param Psi default 0.00181; # damage term without 25% adjustment;
# damage quadratic term with 25% adjustment is 0.00236

```

```

param MD default 0.0163; # market damages for 3°C warming (Nordhaus (2017))
param NMD=max(Normal(0.01646,0.0415),0);# non-market damages for 3°C warming
param TD=MD+NMD; # total damages
param TATlim default 12; # upper bound on atm. temperature change

# Abatement cost
param Theta:=2.6; # Exponent of control cost function
param pback0:=550; # Cost of backstop 2010 $ per tCO2 2015
param gback:=0.025; # Initial cost decline backstop cost per period

param pback {t in 0..T}>=0;
let pback[0]:=pback0;
let {t in 1..T} pback[t]:=pback[t-1]*(1-gback);

param phead {t in 0..T}=pback[t]*sigma[t]/Theta/1000;

# VARIABLES

# capital (trillions 2010 USD)
var K {t in 0..T}>=1;

# Gross output (trillions 2010 USD)
var Qgross {t in 0..T}=A[t]*((L[t]/1000)^(1-gamma))*(K[t]^gamma);

# carbon reservoir atmosphere (GtC)
var MAT {t in 0..T}>=10;

# carbon reservoir upper ocean (GtC)
var MUP {t in 0..T}>=100;

# carbon reservoir lower ocean (GtC)
var MLO {t in 0..T}>=1000;

# total radiative forcing (Wm-2)
var F {t in 0..T}=kappa*((log(MAT[t]/MATEQ))/log(2))+Fex[t];

# atmospheric temperature change (°C from 1750)
var TAT {t in 0..T}>=0, <=TATlim;

```

```

# ocean temperature (°C from 1750)
var TLO {t in 0..T}>=-1, <=20;

# damage fraction
var Omega {t in 0..T}=Psi*(TAT[t])^2;

# damages (trillions 2010 USD)
var damage {t in 0..T}=Omega[t]*Qgross[t];

# emission control
var mu {t in 0..T}>=0;

# abatement costs (fraction of output)
var Lambda {t in 0..T}=Qgross[t]*phead[t]*(mu[t]^Theta);

# industrial emissions
var EInd {t in 0..T}=sigma[t]*Qgross[t]*(1-mu[t]);

# total emissions
var E {t in 0..T};

# maximum cumulative extraction fossil fuels (GtC)
var Ecum {t in 0..T}<=6000;

# Marginal cost of abatement (carbon price)
var cprice {t in 0..T}=pback[t]*mu[t]^(Theta-1);

# output net of damages and abatement(trillions 2010 USD)
var Q {t in 0..T}=(Qgross[t]*(1-Omega[t]))-Lambda[t];

# per capita consumption (1000s 2010 USD)
var c {t in 0..T} >= .1;

# aggregate consumption
var C {t in 0..T} = L[t]*c[t]/1000;

# Investment(trillions 2005 USD)

```

```

var I {t in 0..T}>=0;

# Non-market good
var EQ {t in 0..T}>=0.0000001 <=1000;

# Non-market damages scaling parameter including subsistence requirement

# including sub
var a {t in 0..T} =(1/(nu^2))*(EQ[0]*(EQbar+((EQ[0]-EQbar)^(zeta)
+((1-beta)/beta)*(((1-TD)*C[0])^(zeta)-((1-MD)*C[0])^(zeta))))^(1/zeta))^(-1)-1);

# growth rate of market good

var g_C {t in 0..T-1} = (C[t+1]-C[t])/C[t];

# growth rate of non-market good

var g_EQ {t in 0..T-1} = ((EQ[t+1]-EQ[t])/EQ[t]);

# relative price effect

var RPE {t in 0..T-1} =(1-zeta)*(g_C[t]-((EQ[t]/(EQ[t]-EQbar))*g_EQ[t]));

# utility

var U {t in 0..T}= (((1-beta)*(c[t])^(zeta)
+beta*((EQ[t]-EQbar)*1000/L[t])^(zeta))^(1-eta))/(1-eta);

# welfare/objective function
var W=sum{t in 0..T} L[t]*U[t]*R[t];

maximize objective_function: W;
subject to initial_consumption: c[0]=10.4893;
subject to constr_accounting {t in 0..T}: C[t]=Q[t]-I[t];
subject to constr_emissions {t in 0..T}: E[t]=EInd[t]+ELand[t];
subject to constr_capital_dynamics {t in 1..T}:
K[t]=(1-deltaK)^5*K[t-1]+5*I[t-1];
subject to constr_cumulativeemissions {t in 1..T}:

```

```

Ecum[t]=Ecum[t-1]+(EInd[t-1]*5/3.666);
subject to constr_atmosphere {t in 1..T}:
MAT[t]=E[t]*(5/3.666)+phi11*MAT[t-1]+phi21*MUP[t-1];
subject to constr_upper_ocean {t in 1..T}:
MUP[t]=phi12*MAT[t-1]+phi22*MUP[t-1]+phi32*MLO[t-1];
subject to constr_lower_ocean {t in 1..T}:
MLO[t]=phi23*MUP[t-1]+phi33*MLO[t-1];
subject to constr_atmospheric_temp {t in 1..T}:
TAT[t]=TAT[t-1]+xi1*((F[t]-xi2*TAT[t-1])-(xi3*(TAT[t-1]-TLO[t-1])));
subject to constr_ocean_temp {t in 1..T}:
TLO[t]=TLO[t-1]+xi4*(TAT[t-1]-TLO[t-1]);

# Initial conditions
subject to initial_capital: K[0] = K0;
subject to initial_Ecum: Ecum[0]=Ecum0;
subject to initial_MAT: MAT[0]=MAT0;
subject to initial_MUP: MUP[0]=MUP0;
subject to initial_MLO: MLO[0]=MLO0;
subject to initial_TLO: TLO[0]=TLO0;
subject to initial_TAT: TAT[0]=TAT0;
subject to initial_control: mu[0]=mu0;
subject to control1 {t in 1..28}: mu[t]<=1;
subject to control2 {t in 29..T}: mu[t]<=1.2; # from 2150

subject to initial_EQ: EQ[0]=C[0];
subject to constr_EQ {t in 1..T}: EQ[t]=(EQ[0]/(1+a[t]*(TAT[t]^2)));

```



## AMPL-run file, Plausible Ranges

```
# DICE_2016_RPE_MonteCarlo.run
reset;
model DICE_2016_RPE_MonteCarlo.mod;
data random_delta_eta.dat;
option solver knitroampl;
for {i in 1..nruns} {
reset data zeta,deltaA,EQbar,NMD;
let eta:=etas[i];
let rho:=rhos[i];
solve;
display eta,rho,zeta,deltaA,EQbar,NMD;
# Produce csv-file with overview of parameters
printf "%f\t", eta>Results_figure5_parameters.csv;
printf "%f\t", rho>Results_figure5_parameters.csv;
printf "%f\t", zeta>Results_figure5_parameters.csv;
printf "%f\t", deltaA>Results_figure5_parameters.csv;
printf "%f\t", EQbar>Results_figure5_parameters.csv;
printf "%f\n", NMD>Results_figure5_parameters.csv;
# Produce csv-file with data for figure 5
for {t in 0..T-2}{
printf "%f\t", EInd[t]>Results_figure5_Emissions.csv;}
printf "%f\n", EInd[T-1]>Results_figure5_Emissions.csv;;
for {t in 0..T-2}{
printf "%f\t", TAT[t]>Results_figure5_Temperature.csv;}
printf "%f\n", TAT[T-1]>Results_figure5_Temperature.csv;
for {t in 0..T-2}{
printf "%f\t", -1000*constr_emissions[t]/constr_accounting[t]>
Results_figure5_SCC.csv;}
printf "%f\n", -1000*constr_emissions[T-1]/constr_accounting[T-1]>
Results_figure5_SCC.csv;
for {t in 0..T-2}{
printf "%f\t", (((RPE[t]+1)^(1/5))-1)*100>Results_figure5_RPE.csv;}
printf "%f\n", (((RPE[T-1]+1)^(1/5))-1)*100>Results_figure5_RPE.csv;
}
end
```