

Energy Budget Constraints on the Time History of Aerosol Forcing and Climate Sensitivity

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Text S1. Non-aerosol radiative forcing timeseries

This supplementary text describes the non-aerosol forcing time series used in the simulations. Most forcing categories are similar to those used in version 1.3 of the Finite-amplitude Impulse Response model (FaIR; Smith, Forster, et al., 2018) with updates where appropriate.

Greenhouse gases

Greenhouse gas (GHG) forcings are calculated from concentration to radiative forcing (RF) relationships (Etminan et al., 2016; Hodnebrog et al., 2013; Myhre et al., 2013). Concentrations are taken from the observationally-based CMIP6 historical time series (Meinshausen et al., 2017) extended forwards to 2020 using the SSP2-4.5 pathway (Gidden et al., 2019). To move from RF to ERF, 5% is added to the Etminan et al. (2016) RF for CO₂ and 14% subtracted from CH₄ to account for land-surface warming and tropospheric adjustments respectively (Smith, Kramer, et al., 2018). For all other GHGs, ERF is assumed to be the same as RF. We apply a 20% relative uncertainty (5-95% range) on the present-day ERF from CO₂, N₂O and other GHGs except CH₄ for which we apply 28%; this follows AR5 (Myhre et al., 2013) with an increase in the uncertainty for CH₄ following (Etminan et al., 2016). CMIP6 model results suggest these ranges are

conservative and the ERF spread is possibly lower (Smith et al., 2020). We do not modify the shape of the historical greenhouse gas forcing as historical concentrations are known with low uncertainty. This definition of greenhouse gases does not include stratospheric water vapour from methane oxidation, for which we use the Smith, Forster, et al. (2018) value of 12% of the methane ERF based on AR5.

Ozone

Historical tropospheric and stratospheric ozone forcing is used from Checa-Garcia et al. (2018) which is calculated from the CMIP6 ozone prescribed concentrations for 1850 to 2014. For extrapolating tropospheric ozone before 1850 and after 2014, we use CMIP6 historical and SSP2-4.5 precursor emissions to forcing relationships for CO, NO_x, non-methane volatile organic compounds and CH₄ from Smith, Forster, et al. (2018), based on Stevenson et al. (2013), and updated coefficients to correspond to the CMIP6 emissions and Checa-Garcia et al. forcing. We use a 50% uncertainty on our best estimate of tropospheric ozone ERF of 0.36 W m⁻² for 1750-2019. Stratospheric ozone forcing is based on equivalent effective stratospheric chlorine (EESC) of halogenated compound concentrations (Newman et al., 2007) and has a 200% relative uncertainty following AR5 around the best estimate of -0.05 W m⁻².

Other anthropogenic forcings

Land use forcing from Ghimire et al. (2014) is used for the historical period, scaled to -0.12 W m⁻² for 1750-2011, and extrapolated forward using cumulative land-use CO₂ emissions to 2019 (Smith, Forster, et al., 2018). To this we apply a 75% uncertainty range. For aviation contrail forcing, we sum radiative forcing (RF) from linear contrails from Gettelman & Chen (2013) and radiative forcing from contrail cirrus from Bock & Burkhardt (2016), both based on the year 2006. This is converted to a timeseries of radiative forcing by scaling by aviation NO_x emissions, using aviation NO_x emissions as a proxy for aviation activity. The aviation ERF is derived from the RF by multiplying by 35% (Bickel et al., 2020) owing to strongly-opposing rapid adjustments, and is small (0.02 W m⁻²). Black carbon on snow forcing follows BC emissions in the SSP time series and is scaled to 0.08 W m⁻² in 2018 relative to 1750.

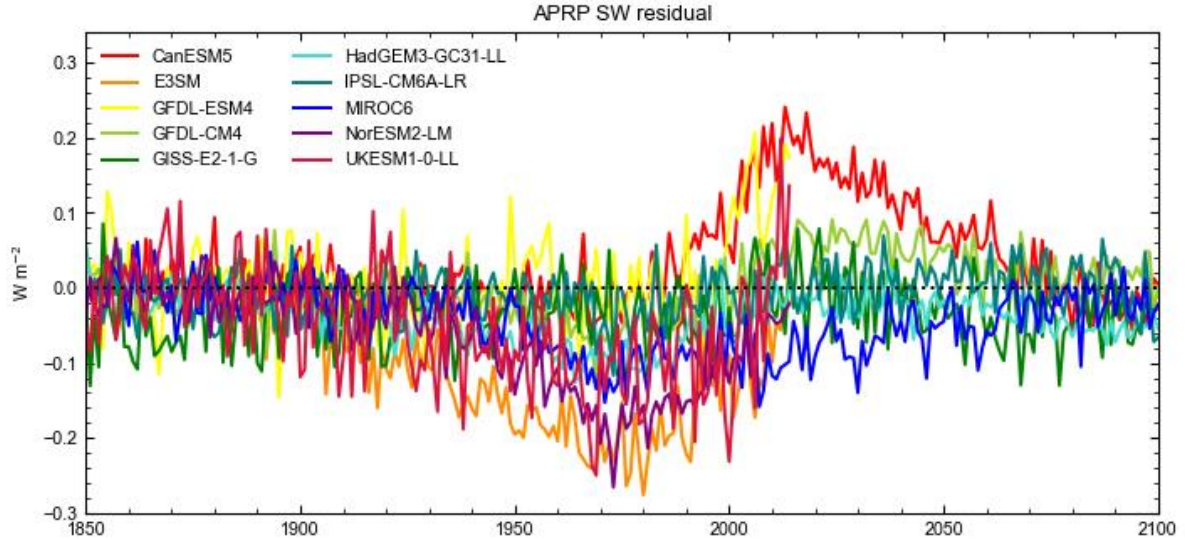
Volcanic forcing

Volcanic ERF is calculated using -18τ (Gregory et al., 2016), where τ is the stratospheric aerosol optical depth (SAOD) at 550 nm. We combine 3 overlapping datasets for SAOD: the eVolv database for the period 500 BCE to 1900 CE (Toohey & Sigl, 2017), the CMIP6 historical volcanic SAOD for 1850-2014, and the GloSSAC dataset for 1979-2018 (Kovilakam et al., 2020). The GloSSAC dataset is converted from SAOD at 525 nm to the target 550 nm using an Ångström exponent of -2.33 (Kovilakam et al., 2020). For 2019 we repeat 2018 noting no significant eruptions occurred (Global Volcanism Program, 2013). A linear transition between eVolv and CMIP6 is performed for 1850-1900 and for CMIP6 to GloSSAC for 1979-1989. The time-mean SAOD over the 2519 years of available data is defined to be the zero forcing, such that quiescent years have a small positive forcing. The rationale for this is that the long-term mean temperature anomaly should be zero when only volcanic forcing is present.

Solar forcing

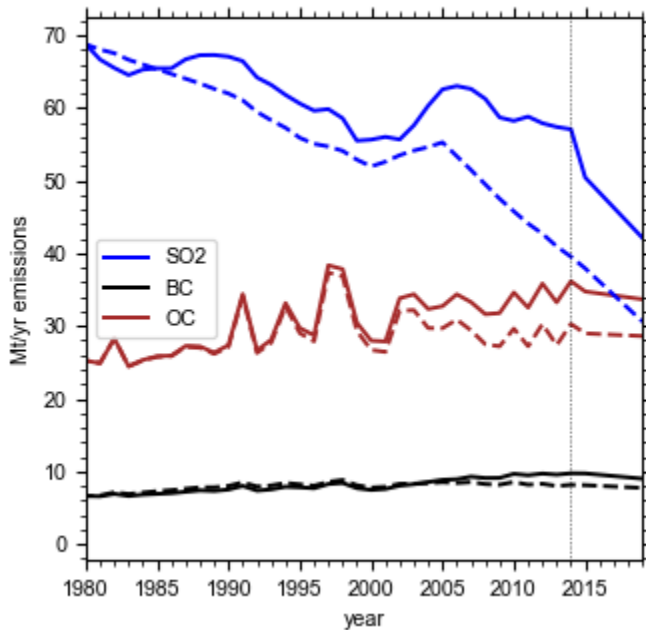
The solar forcing is taken from the derived SATIRE-M ^{14}C solar irradiance database for the 9000 years up to 2014 (Vieira et al., 2011) and the concurrent CMIP6 total solar irradiance from 2015 onwards (Matthes et al., 2017). The solar irradiance timeseries is referenced to the two solar cycles spanning 1745-1765, as a proxy for 1750 conditions. Effective radiative forcing from solar variability is then calculated from the annual total solar irradiance anomaly, multiplied by $\frac{1}{4}$ (geometric effect) $\times 0.71$ (planetary co-albedo) $\times 0.72$ (tropospheric and stratospheric adjustments (Gray et al., 2009; Smith, Kramer, et al., 2018)).

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100 **Figure S1:** Residual from shortwave APRP decomposition of aerosol ERF in RFMIP and
101 AerChemMIP models and E3SM (showing $ERF_{SW} - ERF_{aer,SW} - ERF_{fac,SW}$)
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104 **Figure S2:** Comparison of the CMIP6 emissions dataset (solid lines) with the ECLIPSE
105 v6b emissions dataset considered as an alternative in this study. Solid lines are the
106 CMIP6 emissions, dashed lines are the ECLIPSE v6b emissions. SO_2 is expressed in Mt

107 S. A linear transition is performed between 1980 and 1990, which is the first year of
108 ECLIPSE emissions.

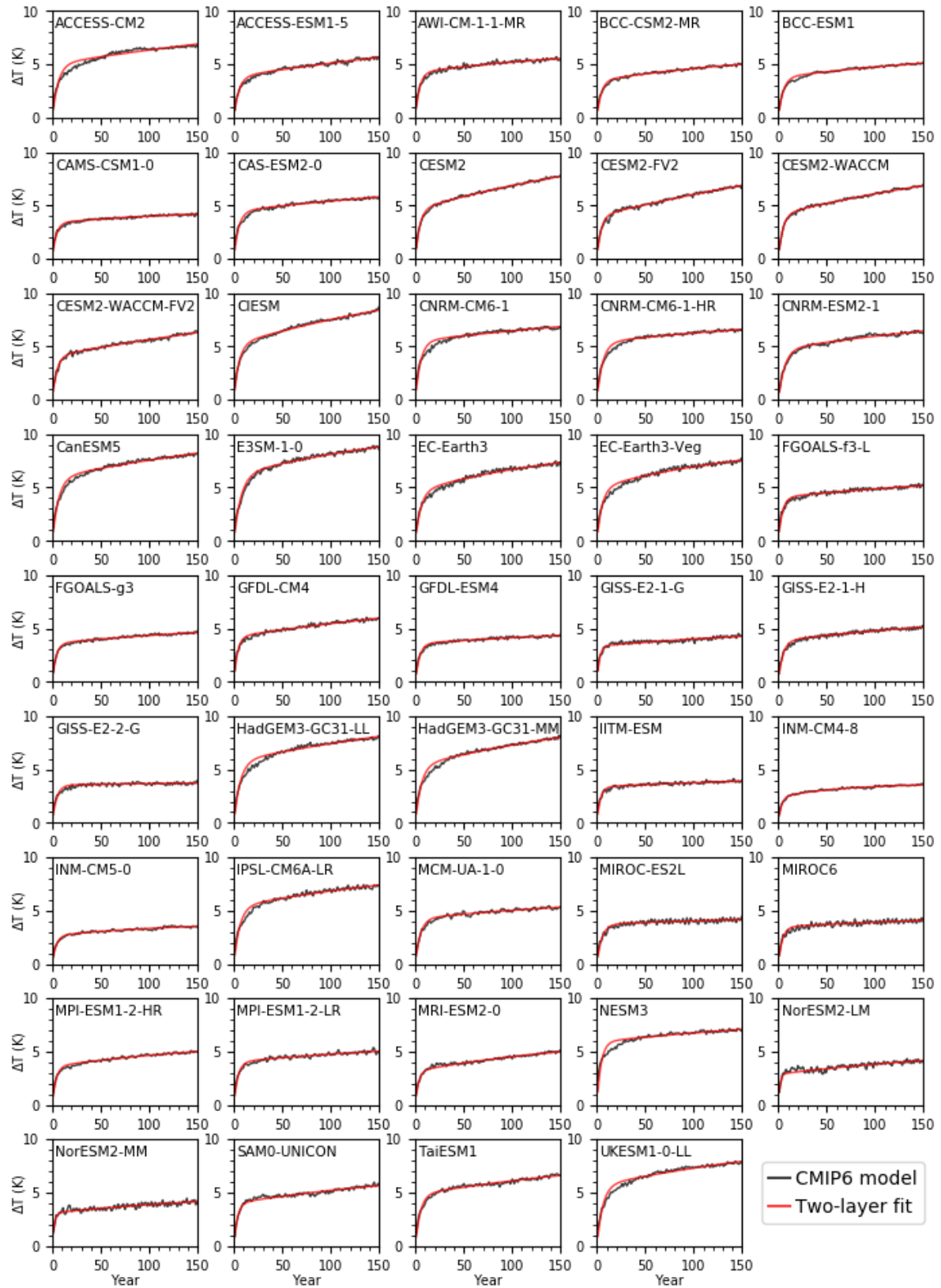


Figure S3: Two-layer model fits to CMIP6 abrupt 4xCO₂ global surface air temperature (GSAT).

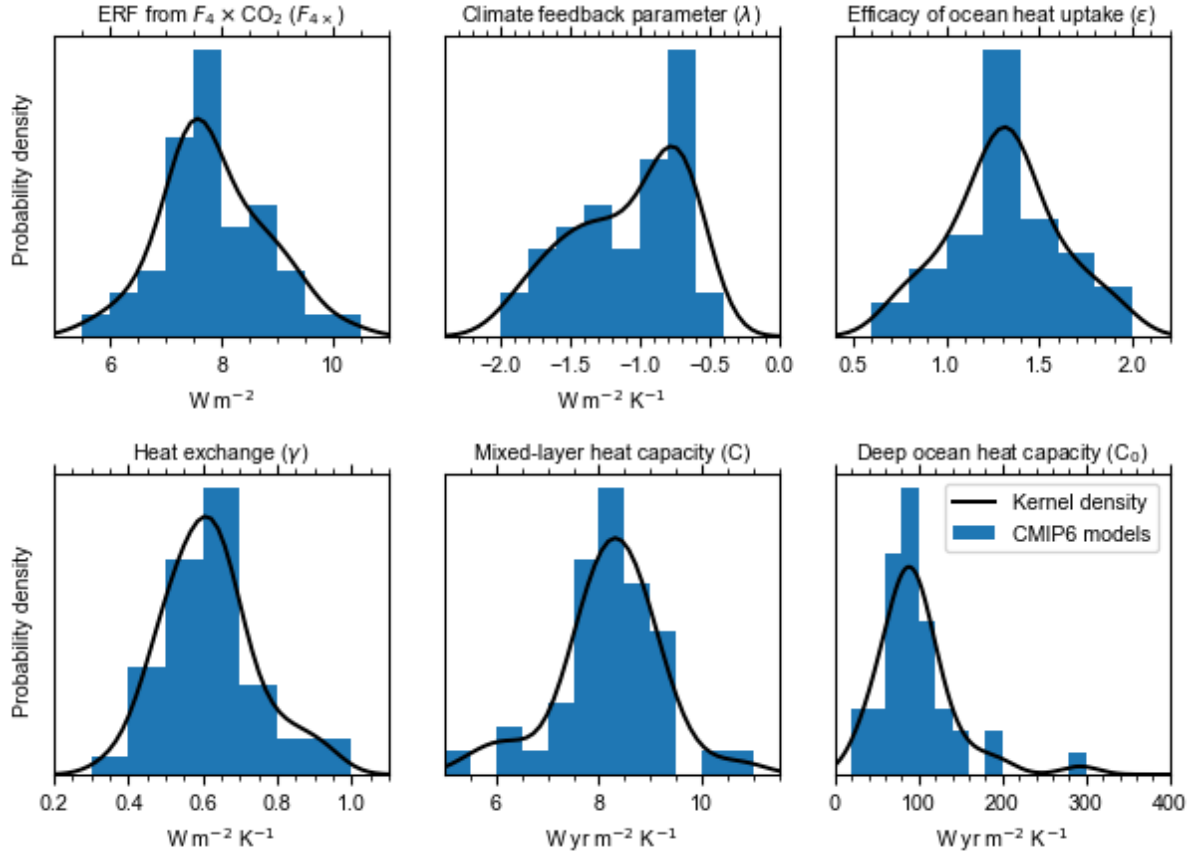


Figure S4: Histograms of Geoffroy model parameter emulations from CMIP6 model and kernel density estimate fit to distributions

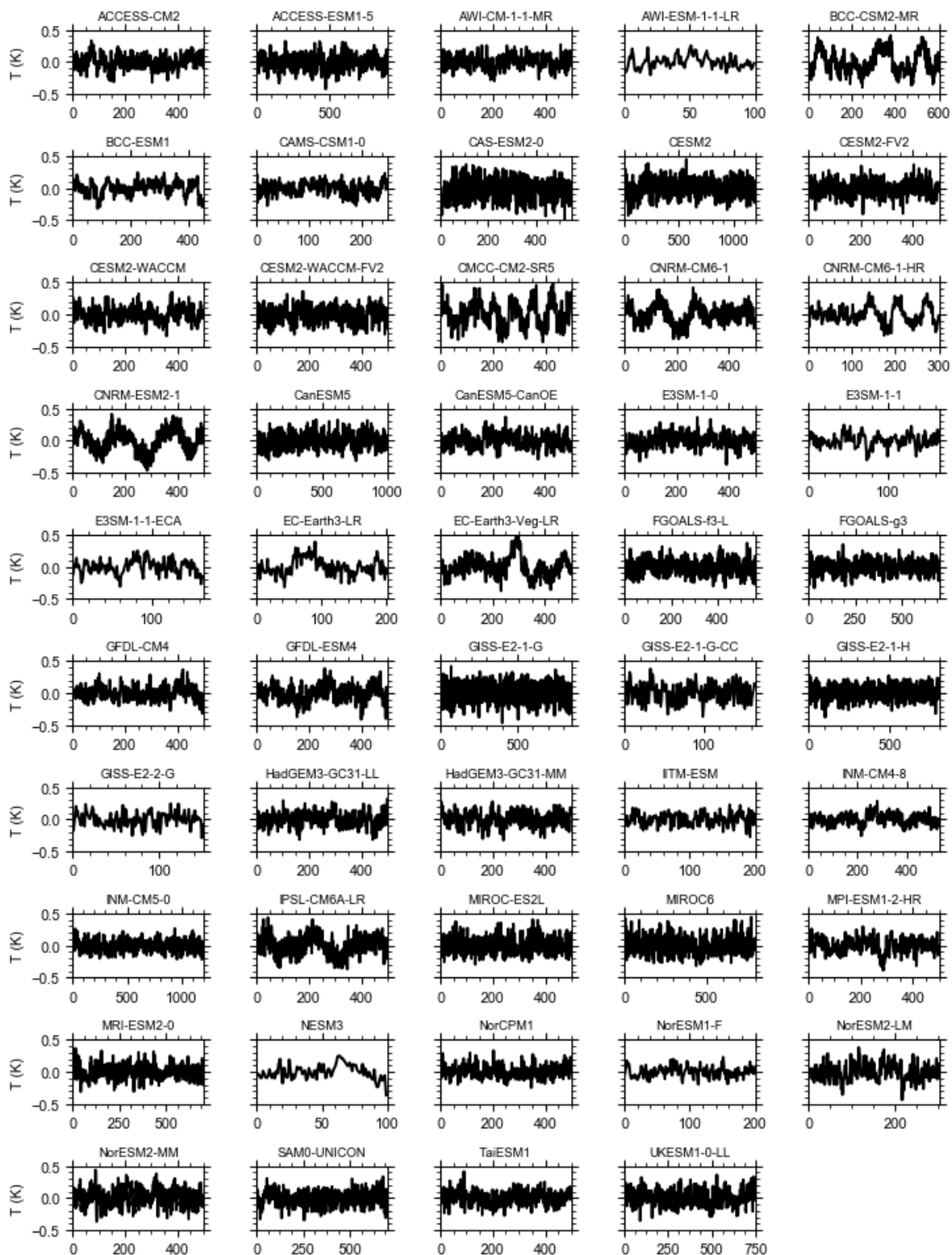
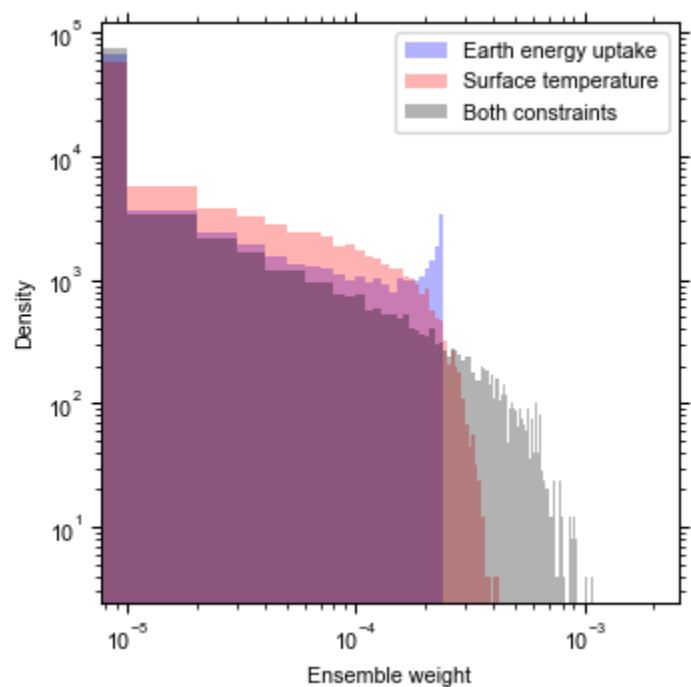


Figure S5: De-trended CMIP6 pre-industrial control global mean temperature anomalies used for characterisation of internal variability.

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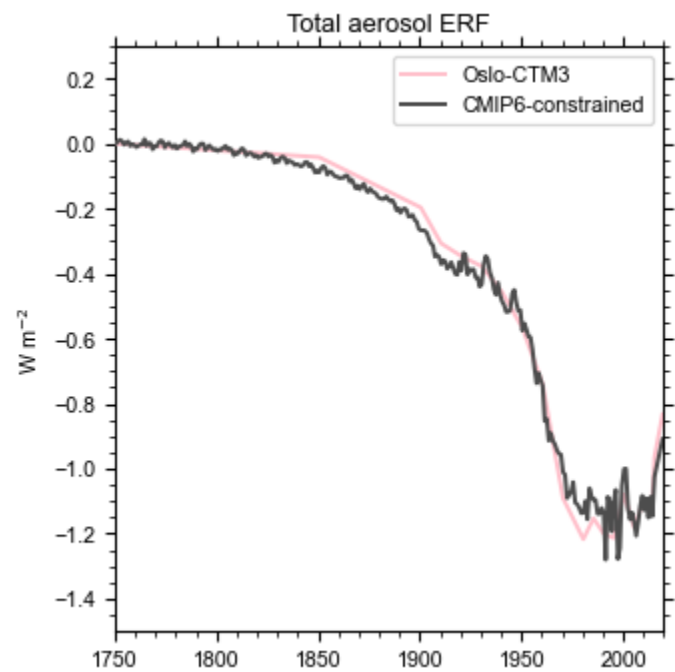
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Figure S6: log-log histogram of ensemble member density using ocean heat uptake only (blue), surface temperature only (red) and both constraints (grey). Most ensemble members have a low weight and near-zero contribution to the total ensemble.



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Figure S7: Comparison of best-estimate aerosol ERF from the CMIP6-constrained time-series and Lund et al., (2018, 2019) time series from Oslo-CTM3 scaled based on observational constraints.

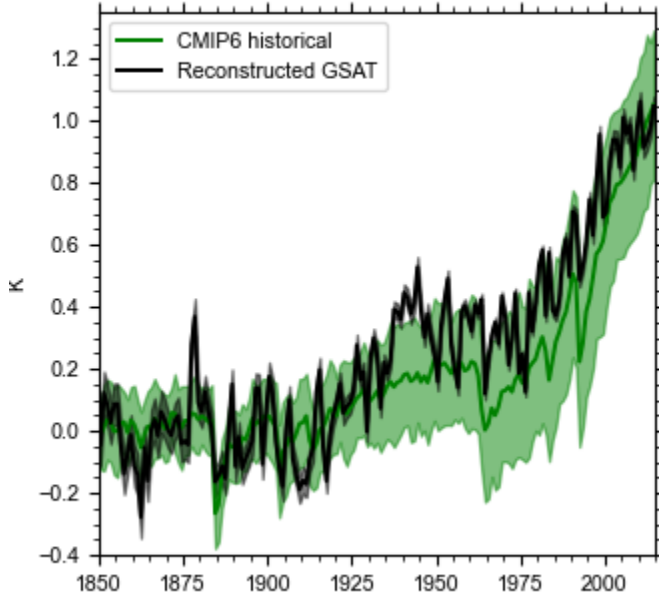


Figure S8: Historical warming simulated by 47 CMIP6 models (r1i1p1f1 used where multiple realisations exist) compared with reconstructed GSAT observations.

Table S1: Two-layer model parameter fits for CMIP6 models.

Model	F_{4x}	λ	C_{mix}	C_{deep}	γ	ϵ	ECS
ACCESS-CM2	7.66	-0.69	8.82	97.46	0.53	1.49	5.57
ACCESS-ESM1-5	6.97	-0.72	9.02	96.79	0.61	1.71	4.83
AWI-CM-1-1-MR	8.41	-1.30	8.17	54.70	0.49	1.30	3.24
BCC-CSM2-MR	6.89	-1.06	8.51	73.69	0.64	1.32	3.25
BCC-ESM1	6.68	-0.94	8.47	91.72	0.58	1.33	3.57
CAMS-CSM1-0	8.88	-1.88	10.01	62.41	0.53	1.34	2.37
CAS-ESM2-0	7.13	-0.93	7.57	72.83	0.45	1.43	3.84
CESM2	8.84	-0.72	8.33	70.64	0.64	1.73	6.15
CESM2-FV2	7.94	-0.56	7.96	91.10	0.70	1.91	7.10
CESM2-WACCM	8.28	-0.73	8.72	84.86	0.72	1.63	5.64
CESM2-WACCM-FV2	7.13	-0.59	7.60	111.71	0.71	1.54	6.00
CIesm	8.94	-0.72	8.59	73.23	0.67	1.38	6.25
CNRM-CM6-1	7.51	-0.77	7.05	121.14	0.56	1.03	4.88
CNRM-CM6-1-HR	7.53	-0.94	8.07	90.43	0.58	0.75	3.99
CNRM-ESM2-1	5.70	-0.63	7.48	94.94	0.61	0.87	4.51
CanESM5	7.61	-0.66	7.86	78.83	0.54	1.09	5.78

E3SM-1-0	7.42	-0.64	8.43	43.77	0.36	1.41	5.81
EC-Earth3	7.37	-0.82	8.46	38.41	0.48	1.43	4.48
EC-Earth3-Veg	7.81	-0.85	8.12	38.54	0.45	1.42	4.57
FGOALS-f3-L	9.54	-1.43	9.29	87.98	0.53	1.63	3.33
FGOALS-g3	7.87	-1.28	8.21	112.65	0.67	1.30	3.07
GFDL-CM4	8.45	-0.89	7.36	96.54	0.58	1.85	4.75
GFDL-ESM4	7.34	-1.27	8.44	129.94	0.60	1.23	2.88
GISS-E2-1-G	8.11	-1.46	6.52	145.03	0.85	1.09	2.78
GISS-E2-1-H	7.58	-1.17	8.76	83.36	0.63	1.21	3.25
GISS-E2-2-G	7.25	-1.63	8.63	293.05	0.57	0.72	2.23
HadGEM3-GC31-LL	7.46	-0.62	8.07	78.11	0.51	1.23	5.98
HadGEM3-GC31-MM	7.37	-0.67	7.85	70.75	0.61	1.07	5.53
IITM-ESM	9.25	-1.93	9.38	159.19	0.73	1.07	2.39
INM-CM4-8	6.25	-1.69	7.84	29.20	0.64	1.21	1.85
INM-CM5-0	6.35	-1.59	9.20	51.95	0.55	1.38	2.00
IPSL-CM6A-LR	7.52	-0.76	8.21	60.03	0.44	1.35	4.93
MCM-UA-1-0	7.12	-1.04	8.48	129.93	0.74	0.80	3.41
MIROC-ES2L	7.98	-1.54	10.82	199.60	0.66	0.89	2.59
MIROC6	7.73	-1.36	9.00	182.38	0.64	1.31	2.84
MPI-ESM1-2-HR	8.63	-1.34	8.89	84.13	0.66	1.49	3.23
MPI-ESM1-2-LR	9.28	-1.46	9.25	104.63	0.63	1.29	3.18
MRI-ESM2-0	8.03	-1.20	7.89	92.84	0.94	1.34	3.34
NESM3	7.72	-0.83	5.50	103.69	0.47	0.97	4.62
NorESM2-LM	10.21	-1.77	6.01	115.66	0.90	1.92	2.89
NorESM2-MM	9.39	-1.69	6.13	116.97	0.79	1.66	2.77
SAM0-UNICON	8.69	-1.14	7.67	106.86	0.82	1.21	3.83
TaiESM1	8.51	-0.92	8.72	97.26	0.63	1.27	4.64
UKESM1-0-LL	7.61	-0.68	7.77	77.15	0.54	1.14	5.57

Table S2: Correlation coefficients used for six-dimensional joint kernel density estimate for obtaining random “meta-models” derived from CMIP6 model fits.

	F_{4x}	λ	C	C_0	γ	ε
F_{4x}	1.0000	-0.4649	0.0134	0.0533	0.3159	0.4487
λ		1.0000	-0.3400	-0.3828	-0.1569	0.1990
C			1.0000	0.1686	-0.4411	-0.4355
C_0				1.0000	0.1046	-0.2542
γ					1.0000	0.2857
ε						1.0000

Table S3: Best estimates, 68% and 90% ranges for aerosol forcing, ECS and TCR from each scaled historical aerosol time series based on weighted percentiles of the distribution after applying constraints. Aerosol ERFs are for 2019 relative to 1750 and for the SSP2-4.5 scenario except for ECLIPSE-constrained.

Time series	Variable	5%	16%	50%	84%	95%
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CMIP6-constrained	ECS	1.93	2.35	3.13	4.44	5.99
	TCR	1.34	1.53	1.85	2.29	2.68
	ERFaer	-1.49	-1.25	-0.89	-0.56	-0.37
	ERFari	-0.6	-0.45	-0.27	-0.14	-0.07
	ERFaci	-1.17	-0.93	-0.59	-0.29	-0.13
ECLIPSE-constrained	ECS	1.88	2.29	3.03	4.16	5.49
	TCR	1.31	1.5	1.8	2.23	2.56
	ERFaer	-1.37	-1.07	-0.72	-0.43	-0.26
	ERFari	-0.66	-0.44	-0.23	-0.1	-0.04
	ERFaci	-0.99	-0.76	-0.46	-0.21	-0.08
CanESM5	ECS	1.75	2.1	2.82	3.76	4.74
	TCR	1.23	1.42	1.72	2.12	2.39
	ERFaer	-0.89	-0.66	-0.34	-0.06	0.08
	ERFari	0.05	0.08	0.13	0.18	0.22
	ERFaci	-0.98	-0.77	-0.46	-0.21	-0.08
E3SM	ECS	1.94	2.36	3.13	4.45	5.95
	TCR	1.35	1.53	1.84	2.28	2.65
	ERFaer	-1.32	-1.18	-0.91	-0.63	-0.47
	ERFari	-0.68	-0.55	-0.38	-0.24	-0.16
	ERFaci	-0.91	-0.77	-0.51	-0.26	-0.13
GFDL-CM4	ECS	1.88	2.29	3.01	4.16	5.4
	TCR	1.3	1.5	1.8	2.21	2.51
	ERFaer	-1.04	-0.86	-0.58	-0.32	-0.18
	ERFari	-0.17	-0.14	-0.1	-0.06	-0.04
	ERFaci	-0.97	-0.77	-0.48	-0.21	-0.08
GFDL-ESM4	ECS	1.88	2.27	2.96	4.09	5.37
	TCR	1.3	1.49	1.77	2.15	2.44
	ERFaer	-1.22	-1.06	-0.76	-0.47	-0.31
	ERFari	-0.31	-0.25	-0.17	-0.11	-0.07
	ERFaci	-1.07	-0.88	-0.57	-0.29	-0.13
GISS-E2-1-G	ECS	1.97	2.42	3.23	4.71	6.38
	TCR	1.36	1.55	1.89	2.36	2.75
	ERFaer	-1.56	-1.34	-0.96	-0.58	-0.36
	ERFari	-0.26	-0.21	-0.14	-0.09	-0.06
	ERFaci	-1.43	-1.21	-0.81	-0.43	-0.21
HadGEM3-GC31-LL	ECS	1.89	2.32	3.04	4.23	5.46
	TCR	1.31	1.5	1.82	2.24	2.57
	ERFaer	-1.32	-1.15	-0.83	-0.53	-0.36
	ERFari	-0.37	-0.3	-0.2	-0.13	-0.08
	ERFaci	-1.13	-0.93	-0.61	-0.31	-0.14
IPSL-CM6A-LR	ECS	1.93	2.36	3.13	4.44	5.82
	TCR	1.34	1.53	1.85	2.3	2.66
	ERFaer	-1.2	-1.06	-0.82	-0.57	-0.42
	ERFari	-0.62	-0.49	-0.34	-0.21	-0.14
	ERFaci	-0.82	-0.69	-0.47	-0.25	-0.12

MIROC6	ECS	2	2.43	3.26	4.77	6.44
	TCR	1.38	1.56	1.89	2.36	2.75
	ERFaer	-1.48	-1.29	-0.96	-0.61	-0.41
	ERFari	-0.43	-0.35	-0.24	-0.15	-0.1
	ERFaci	-1.26	-1.06	-0.7	-0.35	-0.16
NorESM2-LM	ECS	1.93	2.36	3.12	4.41	5.86
	TCR	1.34	1.53	1.85	2.28	2.64
	ERFaer	-1.27	-1.13	-0.86	-0.58	-0.42
	ERFari	-0.57	-0.45	-0.31	-0.19	-0.13
	ERFaci	-0.96	-0.8	-0.52	-0.27	-0.12
Oslo-CTM3	ECS	1.93	2.38	3.17	4.52	6
	TCR	1.34	1.53	1.87	2.33	2.7
	ERFaer	-1.31	-1.13	-0.83	-0.54	-0.37
	ERFari	-0.43	-0.36	-0.25	-0.15	-0.1
	ERFaci	-1.09	-0.89	-0.56	-0.27	-0.11
UKESM1-0-LL	ECS	1.85	2.25	2.95	3.96	5.09
	TCR	1.29	1.48	1.78	2.17	2.46
	ERFaer	-1.1	-0.96	-0.72	-0.48	-0.34
	ERFari	-0.45	-0.36	-0.25	-0.16	-0.1
	ERFaci	-0.85	-0.7	-0.46	-0.23	-0.1

Table S4: As Table S3, using only GSAT as the constraint.

Time series	Variable	5%	16%	50%	84%	95%
CMIP6-constrained	ECS	1.95	2.36	3.15	4.57	6.11
	TCR	1.37	1.54	1.85	2.29	2.66
	ERFaer	-1.49	-1.22	-0.82	-0.48	-0.29
	ERFari	-0.6	-0.44	-0.26	-0.13	-0.06
	ERFaci	-1.15	-0.89	-0.53	-0.23	-0.08
ECLIPSE-constrained	ECS	1.94	2.34	3.11	4.42	5.86
	TCR	1.37	1.53	1.82	2.24	2.57
	ERFaer	-1.31	-1	-0.64	-0.35	-0.19
	ERFari	-0.61	-0.41	-0.21	-0.09	-0.02
	ERFaci	-0.93	-0.7	-0.4	-0.16	-0.03
CanESM5	ECS	1.91	2.31	3	4.25	5.52
	TCR	1.34	1.51	1.79	2.17	2.46
	ERFaer	-0.81	-0.58	-0.25	0.01	0.15
	ERFari	0.05	0.08	0.13	0.19	0.23
	ERFaci	-0.9	-0.68	-0.39	-0.15	-0.03
E3SM	ECS	1.93	2.34	3.11	4.49	6.02
	TCR	1.36	1.53	1.83	2.27	2.64
	ERFaer	-1.33	-1.16	-0.87	-0.58	-0.41
	ERFari	-0.66	-0.54	-0.37	-0.22	-0.14
	ERFaci	-0.91	-0.75	-0.48	-0.23	-0.09
GFDL-CM4	ECS	1.93	2.33	3.07	4.36	5.76

	TCR	1.36	1.52	1.82	2.23	2.54
	ERFaer	-1	-0.81	-0.52	-0.27	-0.13
	ERFari	-0.16	-0.14	-0.09	-0.06	-0.04
	ERFaci	-0.93	-0.72	-0.42	-0.16	-0.03
GFDL-ESM4	ECS	1.89	2.26	2.95	4.09	5.39
	TCR	1.33	1.5	1.76	2.14	2.44
	ERFaer	-1.26	-1.04	-0.7	-0.4	-0.24
	ERFari	-0.31	-0.25	-0.17	-0.1	-0.06
	ERFaci	-1.1	-0.87	-0.52	-0.23	-0.07
GISS-E2-1-G	ECS	1.96	2.38	3.21	4.69	6.37
	TCR	1.38	1.55	1.87	2.34	2.73
	ERFaer	-1.58	-1.32	-0.89	-0.49	-0.28
	ERFari	-0.26	-0.21	-0.14	-0.09	-0.05
	ERFaci	-1.45	-1.18	-0.74	-0.35	-0.13
HadGEM3-GC31-LL	ECS	1.93	2.34	3.11	4.44	5.91
	TCR	1.36	1.53	1.83	2.25	2.59
	ERFaer	-1.28	-1.09	-0.76	-0.46	-0.29
	ERFari	-0.36	-0.29	-0.2	-0.12	-0.07
	ERFaci	-1.1	-0.89	-0.55	-0.25	-0.08
IPSL-CM6A-LR	ECS	1.94	2.36	3.15	4.57	6.13
	TCR	1.37	1.54	1.85	2.3	2.68
	ERFaer	-1.18	-1.03	-0.77	-0.52	-0.36
	ERFari	-0.6	-0.48	-0.33	-0.19	-0.12
	ERFaci	-0.8	-0.66	-0.43	-0.21	-0.08
MIROC6	ECS	1.97	2.39	3.22	4.7	6.37
	TCR	1.38	1.55	1.88	2.34	2.73
	ERFaer	-1.52	-1.29	-0.9	-0.54	-0.34
	ERFari	-0.42	-0.34	-0.23	-0.14	-0.09
	ERFaci	-1.29	-1.05	-0.65	-0.3	-0.1
NorESM2-LM	ECS	1.94	2.35	3.13	4.48	5.99
	TCR	1.36	1.54	1.84	2.27	2.63
	ERFaer	-1.27	-1.1	-0.8	-0.52	-0.36
	ERFari	-0.55	-0.45	-0.3	-0.18	-0.12
	ERFaci	-0.95	-0.77	-0.48	-0.22	-0.07
Oslo-CTM3	ECS	1.96	2.39	3.2	4.62	6.24
	TCR	1.38	1.55	1.87	2.32	2.69
	ERFaer	-1.29	-1.09	-0.75	-0.46	-0.3
	ERFari	-0.43	-0.35	-0.24	-0.15	-0.1
	ERFaci	-1.07	-0.84	-0.5	-0.21	-0.06
UKESM1-0-LL	ECS	1.92	2.31	3.03	4.27	5.65
	TCR	1.35	1.51	1.8	2.2	2.51
	ERFaer	-1.06	-0.92	-0.67	-0.43	-0.29
	ERFari	-0.44	-0.36	-0.24	-0.14	-0.09
	ERFaci	-0.82	-0.66	-0.41	-0.19	-0.06

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Table S5: As Supplementary Table 3, using only EEU as the constraint.

Time series	Variable	5%	16%	50%	84%	95%
CMIP6-constrained	ECS	2.03	2.46	3.39	5.17	7.31
	TCR	1.38	1.57	1.96	2.59	3.16
	ERFaer	-1.6	-1.34	-0.97	-0.56	-0.27
	ERFari	-0.64	-0.46	-0.27	-0.11	0.08
	ERFaci	-1.26	-1.04	-0.68	-0.32	-0.14
ECLIPSE-constrained	ECS	2.03	2.47	3.44	5.25	7.22
	TCR	1.39	1.58	1.98	2.62	3.19
	ERFaer	-1.46	-1.19	-0.82	-0.45	-0.12
	ERFari	-0.67	-0.42	-0.22	-0.05	0.32
	ERFaci	-1.13	-0.91	-0.58	-0.28	-0.12
CanESM5	ECS	1.97	2.4	3.33	5.07	7.22
	TCR	1.36	1.55	1.94	2.57	3.14
	ERFaer	-1.02	-0.81	-0.45	-0.12	0.07
	ERFari	0.08	0.12	0.2	0.29	0.36
	ERFaci	-1.18	-0.99	-0.67	-0.34	-0.15
E3SM	ECS	2.05	2.48	3.4	5.19	7.23
	TCR	1.39	1.57	1.95	2.59	3.14
	ERFaer	-1.44	-1.28	-1	-0.67	-0.47
	ERFari	-0.72	-0.58	-0.4	-0.25	-0.16
	ERFaci	-1.03	-0.85	-0.56	-0.27	-0.11
GFDL-CM4	ECS	2.06	2.5	3.47	5.37	7.5
	TCR	1.4	1.59	1.99	2.63	3.19
	ERFaer	-1.23	-1.05	-0.73	-0.42	-0.24
	ERFari	-0.2	-0.16	-0.11	-0.07	-0.05
	ERFaci	-1.13	-0.94	-0.61	-0.3	-0.12
GFDL-ESM4	ECS	2.04	2.48	3.41	5.28	7.38
	TCR	1.38	1.57	1.96	2.6	3.15
	ERFaer	-1.36	-1.19	-0.9	-0.56	-0.36
	ERFari	-0.38	-0.31	-0.21	-0.13	-0.08
	ERFaci	-1.17	-0.98	-0.67	-0.34	-0.15
GISS-E2-1-G	ECS	2.04	2.47	3.38	5.18	7.22
	TCR	1.38	1.57	1.95	2.58	3.13
	ERFaer	-1.66	-1.42	-1.02	-0.59	-0.34
	ERFari	-0.27	-0.22	-0.15	-0.09	-0.06
	ERFaci	-1.52	-1.27	-0.86	-0.43	-0.19
HadGEM3-GC31-LL	ECS	2.03	2.47	3.38	5.16	7.18
	TCR	1.38	1.57	1.95	2.59	3.14
	ERFaer	-1.6	-1.39	-1.02	-0.62	-0.39
	ERFari	-0.4	-0.32	-0.22	-0.14	-0.09
	ERFaci	-1.4	-1.17	-0.78	-0.39	-0.17
IPSL-CM6A-LR	ECS	2.05	2.48	3.39	5.17	7.19

	TCR	1.38	1.57	1.95	2.59	3.14
	ERFaer	-1.34	-1.2	-0.93	-0.61	-0.42
	ERFari	-0.61	-0.49	-0.34	-0.21	-0.14
	ERFaci	-0.99	-0.83	-0.56	-0.28	-0.12
MIROC6	ECS	2.06	2.49	3.43	5.27	7.3
	TCR	1.39	1.58	1.96	2.6	3.15
	ERFaer	-1.56	-1.35	-1	-0.62	-0.39
	ERFari	-0.44	-0.36	-0.24	-0.15	-0.1
	ERFaci	-1.34	-1.11	-0.73	-0.36	-0.15
NorESM2-LM	ECS	2.04	2.47	3.38	5.16	7.17
	TCR	1.38	1.57	1.95	2.58	3.11
	ERFaer	-1.43	-1.26	-0.97	-0.62	-0.43
	ERFari	-0.59	-0.47	-0.32	-0.2	-0.13
	ERFaci	-1.11	-0.92	-0.61	-0.3	-0.13
Oslo-CTM3	ECS	2.05	2.49	3.41	5.23	7.3
	TCR	1.39	1.58	1.96	2.59	3.14
	ERFaer	-1.44	-1.26	-0.94	-0.6	-0.39
	ERFari	-0.49	-0.39	-0.27	-0.17	-0.11
	ERFaci	-1.2	-0.98	-0.64	-0.31	-0.13
UKESM1-0-LL	ECS	2.02	2.46	3.36	5.11	7.12
	TCR	1.38	1.56	1.95	2.59	3.15
	ERFaer	-1.4	-1.23	-0.93	-0.58	-0.39
	ERFari	-0.49	-0.39	-0.27	-0.17	-0.11
	ERFaci	-1.14	-0.95	-0.63	-0.31	-0.13

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