Estimated effect of the permafrost carbon feedback on the zero emissions commitment to climate change

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Abstract. Zero Emissions Commitment (ZEC), the expected change in global temperature following the cessation of CO_2 emissions has recently been assessed by the Zero Emissions Commitment Model Intercomparison Project (ZECMIP). ZECMIP concluded that the component of ZEC from CO_2 emissions will likely be close to zero in the decades following the cessation of emissions. However, of the 18 Earth system models that participated in ZECMIP only two included a representation of the permafrost carbon feedback to climate change. To better assess the potential impact of permafrost carbon decay on ZEC a series of perturbed parameter experiments are here conducted with an Earth system model of intermediate complexity. The experiment suggest that the permafrost carbon cycle feedback will directly add 0.06 [0.02 to 0.14]°C to the benchmark ZEC value assesses 50 years after 1000 PgC of CO_2 has been emitted to the atmosphere. An additional 0.04 [0 to 0.06]°C is likely to been added relative to the benchmark ZEC value from the thaw-lag effect unaccounted for in the ZECMIP experiment design. Overall we assess that the permafrost carbon feedback is unlikely to change the assessment that ZEC is close to zero on decadal timescales, however the feedback is expected to become more important over the coming centuries.

1 Introduction

Zero Emissions Commitment (ZEC) is the change in global temperature expected to occur following the cessation of anthropogenic emissions of greenhouse gases and aerosols (Hare and Meinshausen, 2006; MacDougall et al., 2020). ZEC is one of five metrics needed to compute the 'remaining carbon budget', which in turn quantifies the total emissions compatible with meeting a given temperature change guardrail (e.g. Rogelj et al., 2018; Matthews et al., 2021), such as those set out in the Paris Agreement (United Nations, 2015). ZEC was recently the focus of a model intercomparison project organized through the Coupled Model Intercomparison Project phase 6 (CMIP6) (Jones et al., 2019). The project, formally called the Zero Emissions Commitment Model Intercomparison Project (ZECMIP) (Jones et al., 2019), gathered simulations from 18 Earth System Models of full and intermediate complexity (MacDougall et al., 2020) and assessed the CO_2 component of ZEC. For the tier-one idealized experiment where 1000 PgC of CO_2 were emitted to the atmosphere before cessation of emissions, ZEC ranged from -0.36 to 0.29 °C with a model ensemble mean of -0.07°C 50 years after emissions ceased. However, only two of the 18 models that participated in ZECMIP had a representation of the permafrost carbon feedback to climate change, a feedback process that is expected to release CO_2 and CH_4 into the atmosphere for centuries after emissions cease (Schuur et al., 2015; McGuire et al., 2018). Thus the effect of the permafrost carbon feedback on ZEC has yet to be well quantified.

The soils of the northern hemisphere permafrost region are estimated to contain between 1100 and 1500 PgC of organic matter (Hugelius et al., 2014), about half of which is held in the perennially frozen zone of these soils (Hugelius et al., 2014). As climate warms and permafrost thaws organic matter in permafrost affected soils is exposed to increased periods of time where temperature is above freezing, and hence to enhanced rate of decay releasing CH_4 and CO_2 to the atmosphere

30   (e.g. Schuur et al., 2015). A recent informal (non-CMIP) model intercomparison exercise quantifying the permafrost carbon feedback estimated a release of carbon of between 74 to 652 PgC by year 2300 under the high end Representative Concentration Pathway 8.5 scenario, with substantially lower release or even gain of soil carbon under mitigation scenarios (McGuire et al., 2018). Thus the permafrost carbon feedback to climate change has the potential the affect the value of ZEC in a fashion that was poorly quantified by ZECMIP.

35   Uncertainty in projections from Earth system models can be classified into three components: 1) structural uncertainty, 2) parameter uncertainty, and 3) scenario uncertainty (e.g. MacDougall and Knutti, 2016). Structural uncertainty is created from the discrepancy between the system the model is intended to represent and the system the model actually describes (Smith, 2007; Eyring et al., 2016). Examining different models of the same system with standardized forcings through model inter-comparison projects such as ZECMIP are the principle way of quantifying structural uncertainty in climate sciences Eyring

40   et al. (2016). Parameter uncertainty is uncertainty about the value model parameter should take on (Smith, 2007). As model parameters are sometimes quantities measurable in the natural world parameter uncertainty is some cases equivalent to measurement uncertainty. In other cases parameters represent an amalgam of natural processes, in such cases defining parameter uncertainty becomes more ambiguous (Smith, 2007). Parameter uncertainty can be quantified with perturbed parameter experiments, wherein ensembles of model variants with parameter values selected from defined probability distribution functions are

45   run under the same experiment conditions (e.g. Forest et al., 2002). Several such experiments have been conducted to assess uncertainty in the permafrost carbon feedback (Schneider von Deimling et al., 2015; MacDougall and Knutti, 2016; Gasser et al., 2018). Scenario uncertainty is created by uncertainty about what humans will do in the future, and is well explored by the coordinated scenario framework of CMIP (Eyring et al., 2016; O'Neill et al., 2017).

For the permafrost carbon feedback uncertainty ranges derived from structural uncertainty and parameter uncertainty assess-

50   ments have proven similar. The model intercomparison exercise of McGuire et al. (2018) found a range of 74 to 652 PgC with a mean of 341 PgC released from permafrost soils by year 2300 under representative concentration pathway 8.5, compared to 159 to 587 PgC with a mean of 376 PgC for the perturbed parameter experiment of MacDougall and Knutti (2016), forced under the same scenario.

In addition to 16 out of the 18 ZECMIP models having no permafrost carbon module, the experimental design of ZECMIP

55   is ill designed to quantify the permafrost carbon feedback. The top-tier idealized ZECMIP experiments branches from the idealized 1pctCO2 experiment where atmospheric CO_2 concentration rises at 1% a year compounded leading to a quadrupling of CO_2 concentration in 140 years (Jones et al., 2019; Eyring et al., 2016). Following this protocol CO_2 and hence global temperatures rise much faster than in the historical trajectory, and since it takes time for permafrost soil to thaw and organic mater within them to decay, the experimental protocol will tend to underestimate release of carbon from permafrost soils

60   (MacDougall, 2019).

Here we will use a perturbed parameter ensemble approach to estimate the contribution to CO_2-only ZEC from the release of carbon from permafrost soils, following the ZECMIP protocol. We will also conduct an experiment following a more realistic CO_2 emission trajectory in order to quantity the thaw-lag effect from the high emission rates of the ZECMIP protocol.

2 Methods

2.1 Model Description

65 The University of Victoria Earth System Climate Model (UVic ESCM) is a climate model of intermediate complexity founded around a three dimensional ocean general circulation model coupled to a simplified moisture and energy balance atmosphere (Weaver et al., 2001). The version of the model used here (version 2.9pf) has full representation of the oceanic and terrestrial carbon cycles (Schmittner et al., 2008; Meissner et al., 2003). The oceanic carbon cycle has representations of ocean carbonate

70 chemistry (Weaver et al., 2001), phytoplankton-zooplankton-detritus ocean biology scheme (Schmittner et al., 2008), and interaction between ocean sediments and alkalinity (Archer, 1996). The terrestrial component is composed of the Top-down Representation of Interactive Foliage and Flora Including Dynamics (Triffid) dynamic vegetation model (Cox et al., 2001; Meissner et al., 2003), a multi-layer representation soil respiration (MacDougall et al., 2012), and a permafrost carbon module (MacDougall and Knutti, 2016).

75 The terrestrial subsurface of the model is composed of 14 layers, reaching a total depth of 250 m (Avis et al., 2011). The top 8 layers (10 m) are active and the hydraulic cycle and deeper layers are impermeable bedrock (Avis et al., 2011). The freeze-thaw physics of the soil accounts for the effect of soil valence forces on freezing point and the frozen and unfrozen fraction of the soil water is calculated using equations the minimize Gibbs free energy (Avis, 2012). The top 6 layers of the model (3.35 m) are active in the carbon cycle. Carbon is assigned to soil layers from Triffid based on the root density in each soil layer, with

80 remaining dead plant matter added to the top soil layer (MacDougall et al., 2012). Root density varies by plant function type and the temperature of the soil layer (roots do not grow in frozen soil) (MacDougall et al., 2012). In model grid cells where permafrost exists (where soil layers have been below 0°C for two or more consecutive years) a diffusion based cryoturbation scheme is used to redistribute soil carbon in the soil column. The scheme was originally developed by Koven et al. (2009) and modified for implementation in the UVic ESCM in MacDougall and Knutti (2016). The scheme allows for a permafrost carbon

85 pool to be generated in permafrost soils. The modifications made to the scheme by MacDougall and Knutti (2016) allow the permafrost carbon pool to come into equilibrium during the 5000 year model spin-up. The version of the UVic ESCM used here does not have a methane production module. Thus emissions of carbon from soils to the atmosphere happens only as CO_2.

In the version of the UVic ESCM used here (MacDougall and Knutti, 2016) permafrost carbon is a separate carbon pool. Permafrost carbon is created when carbon is advected across the permafrost table by the cryoturbation scheme and can only

90 be destroyed by being respired into CO_2. The pool is characterized by a decay rate constant (κ_p), a fraction of the pool that is available for decay (available fraction, A_f), and a passive pool transformation rate (κ_{tf}), which is the rate at which the passive permafrost carbon becomes part of the available fraction. The available fraction is essentially the combined size of the fast and slow carbon pools as conceptualized in incubation experiments (Schädel et al., 2014; MacDougall and Knutti, 2016).

This scheme accounts for the large fraction of permafrost carbon that is very resistant to decay (Schädel et al., 2014), while
95 still allowing the pool to decay over millennial time periods (MacDougall and Knutti, 2016). A fourth parameter, the saturation
factor (S) from the cryoturbation scheme allows the size of the permafrost carbon pool to be tuned (MacDougall and Knutti,
2016).

The model experiments here use the permafrost carbon variant of the UVic ESCM 2.9 detailed in MacDougall and Knutti
(2016). A newer version of the UVic ESCM (version 2.10) is now available (Mengis et al., 2020). We use the older version of
100 the model to allow for the use of legacy code and legacy model spin-ups from MacDougall and Knutti (2016). Note that the
terrestrial component of UVic ESCM 2.10 was taken from the version developed for MacDougall and Knutti (2016), and thus
the terrestrial components of the model versions are virtually identical (Mengis et al., 2020).

By changing the flow of outgoing longwave radiation to space as a function of global surface temperature anomaly the
climate sensitivity of the UVic ESCM can be altered (Zickfeld et al., 2009). Similarly by changing the meridional diffusivity
105 of the atmosphere within the model, arctic amplification can also be altered (Fyke et al., 2014).

2.2 Perturbed Parameter Experiments

To assess the uncertainty in the strength of the permafrost carbon cycle feedback to climate change MacDougall and Knutti
(2016) generated 250 variants of the UVic ESCM by perturbing six model parameters. Four of these parameters control the size
and susceptibility to decay of the permafrost carbon pool and two (climate sensitivity and arctic amplification) are physical
110 climate parameters. The four permafrost carbon parameters are: 1) the permafrost carbon decay constant; 2) the available
fraction; 3) the passive pool transformation rate; and 4) the permafrost carbon saturation factor – which controls the size of
the permafrost carbon pool. The permafrost carbon decay constant controls how fast available permafrost carbon can decay
given the temperature and moisture of the soil. The available fraction is the fraction of permafrost carbon that is allowed
to decay, effectively the fraction of permafrost carbon that is unprotected or weakly protected from decay. The passive pool
115 transformation rate is the rate at which highly protected permafrost carbon becomes weakly protected. The Probability Density
Functions (PDFs) for the permafrost carbon decay constant and the available fraction were taken from the meta-analysis
of permafrost carbon incubation experiments conducted by Schädel et al. (2014). The passive pool transformation rate is
constrained primarily by the non-existence of a remnant mid-latitude permafrost carbon pool from the last glacial maximum,
yielding an estimated value of 0.25×10^{-10} to 4×10^{-10} s^{-1}, with a best guess of 1×10^{-10}s^{-1} (Trumbore, 2000; MacDougall
120 and Knutti, 2016). Fortunately, the analysis of MacDougall and Knutti (2016) showed that the passive pool transformation rate
has only a weak effect of the permafrost carbon feedback on decadal and centennial time scales. The estimated uncertainty
in the size of the permafrost carbon pool was taken from Hugelius et al. (2014). For each model variant 5000 year model
spin-ups were conducted with year 1850 CE radiative forcing to bring the permafrost carbon into equilibrium, representing an
investment of 1.25 million model-years of simulation time. We have re-used these model spin-ups for the present study.

125 MacDougall and Knutti (2016) also perturbed two physical model parameters, climate sensitivity and arctic amplification.
These parameter do not affect the model spin-up as both effect deviations from the pre-industrial climate, thus they can be
changed for the present study. To my knowledge there has been no major update in the uncertainty range of arctic amplification

since 2016, however Sherwood et al. (2020) has substantially constrained the uncertainty in equilibrium climate sensitivity to a 5th to 95th percentile range of 2.3 to 4.7°C for a doubling of atmospheric CO_2 concentration. Thus we have computed

130   new climate sensitivity parameters for the 250 model variants. We use the same functional form for the climate sensitivity PDFs as previous papers (e.g. Olson et al., 2012; MacDougall and Knutti, 2016), a product of two Normal Inverse Gaussian functions. To get new parameter values for the climate sensitivity PDF a Monte-Carlo method was used to fit the function to the distribution outlined by Sherwood et al. (2020): a 5th to 95th percentile range of 2.3 to 4.7°C a 66% range of 2.6 to 3.9°C and a median value of 3.0°C. The new parameter values for the PDF are given in Appendix A.

135   Figure 1 shows the PDFs of each parameter perturbed in this study. Note that all but equilibrium climate sensitivity are identical to MacDougall and Knutti (2016).
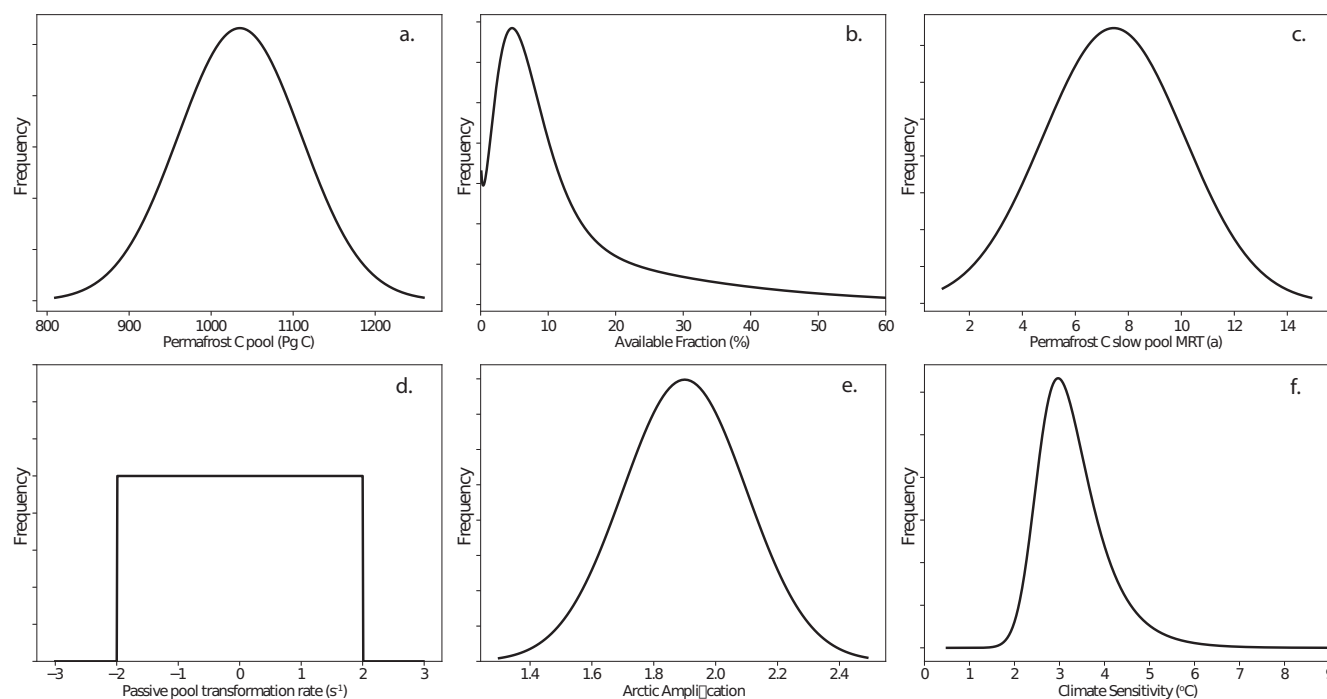
Figure 1. Probability distribution functions of the six parameters perturbed in this study. MRT is mean residence time. Panel (d) has a logarithmic scale. All parameter PDFs except climate sensitivity are that same as MacDougall and Knutti (2016).

2.3 Model Experiments

To quantify the effect of the permafrost carbon feedback on ZEC we have three key questions: 1) How much warming will the permafrost carbon feedback add to ZEC? 2) What is the magnitude of the thaw-lag permafrost effect from using the 1pctCO2

140   experiment to quantify ZEC? And 3) How sensitive is the permafrost carbon feedback contribution to ZEC to total CO_2 emitted before cessation of emissions?

To answer the first question we would ideally compare simulations with and without permafrost carbon that are otherwise identical. In the UVic ESCM framework we can create a version without permafrost carbon by setting the cryoturbation diffusion parameter to zero during model spin-up. Without cryoturbation there will be no permafrost carbon pool and active

145 layer carbon pool will also be reduced in size. However, the presence of carbon in soils subtly changes soil thermal and hydraulic properties in the UVic ESCM (Avis, 2012) such that the absence of a permafrost carbon pool could change the baseline climate conditions of the model. To test the magnitude of this effect simulations were conducted with the UVic ESCM with the cryoturbation diffusion parameter set to zero and all other model parameters held at their default settings. The model version was spun-up for 5000 years. Both the version of the model with cryoturbation set to zero and the default version of

150 the model were forced with the 1pctCO2 experiment (where CO_2 concentration rises at 1% per year compounded). Figure B1 shows that the difference between the two simulations is minimal with respect to global average temperature, with a $0.01^{\circ}C$ difference in baseline global temperature and a smaller difference by the time atmospheric CO_2 concentration doubled in year 70 of the experiment.

Thus to quantify the effect of permafrost carbon feedback on ZEC two parallel sets of experiments were conducted. In one

155 set of experiments model spin-ups from MacDougall and Knutti (2016) were used along with the 250 variants of the model to compute ZEC including permafrost. In a second set of experiments a single model spin-up with the cryoturbation diffusion parameter set to zero was used and 250 model variants were generated using just the climate sensitivity and arctic amplification parameters from the perturbed parameter sets. Thus each parallel variant pair will have the same climate sensitivity and arctic amplification parameters with only the existence of permafrost carbon different between the parallel variants. All model variants

160 were forced with the esm-1pct-brch-1000PgC (A1) ZEC experiment described in Jones et al. (2019) where the 1000 PgC of carbon is emitted following the 1pctCO2 experiment pathway and emission instantaneously go to zero once 1000 PgC is reached. All non-CO_2 forcings are held either at their 1850 values or their long-term mean for volcanic and solar forcing. The simulations are forced with CO_2 emissions diagnosed from the default version of the UVic ESCM 2.9pf such that all simulations are forced with the same CO_2 emissions pathway. Thus most model variants will only approximately follow the

165 1pctCO2 CO_2 trajectory but all variants have the same point in time that emissions cease, greatly simplifying analysis of the results. The difference between parallel variants with and without permafrost carbon quantifies the effect of the permafrost carbon pool on ZEC.

To quantify the permafrost thaw-lag effect a set of experiments were conducted with the 250 model variants with permafrost carbon. The model variants were forced with a CO_2 emissions trajectory that follows historical emissions until year 2019

170 Friedlingstein et al. (2020) and afterwards follows the CO_2 emissions trajectory of a Shared Socioeconomic Pathway (SSP) (O'Neill et al., 2017) until 1000 PgC has been emitted (Figure 2). Thereafter CO_2 emissions are set to zero. All non-CO_2 forcings are held either at their 1850 values or their long-term mean for volcanic and solar forcing. Eight SSPs were used by CMIP6 to quantify scenario uncertainty (O'Neill et al., 2017). For the experiment conducted here we selected an SSP based two criteria: that the SSP reaches 1000 PgC of CO_2 emissions, and that the 1000 PgC cumulative CO_2 emissions total is reached

175 before emissions begin to approach zero. The second criteria is needed as the ZECMIP bell shaped curve experiments showed that the Transient Climate Response to Cumulative CO_2 Emissions (TCRE) and ZEC effects become mixed as emissions

approach zero (MacDougall et al., 2020). Therefore a sudden cessations of emissions is needed to separate TCRE from ZEC. SSP4-6.0 is the lowest emission SSP that reaches 1000 PgC whilst maintaining near-peak CO_2 emissions. Thus using SSP4-6.0 maximizes the time need to reach 1000 PgC and therefore is optimal for assessing the thaw-lag effect. Under SSP4-6.0 CO_2

180 emissions 1000 PgC is reached in year 2067, allowing permafrost the time to thaw and the organic matter within it to decay.

The effect of total CO_2 emissions on the permafrost carbon contribution to ZEC is quantified by forcing each of the 250 parallel model variants with the esm-1pct-brch-2000PgC (A3) ZEC experiment from Jones et al. (2019) wherein 2000 PgC of CO_2 is emitted following 1pctCO2 experiment pathway and emission instantaneously go to zero once 2000 PgC is reached. Again the permafrost carbon effect on ZEC is computed from the difference between parallel variants with and without permafrost

185 carbon.
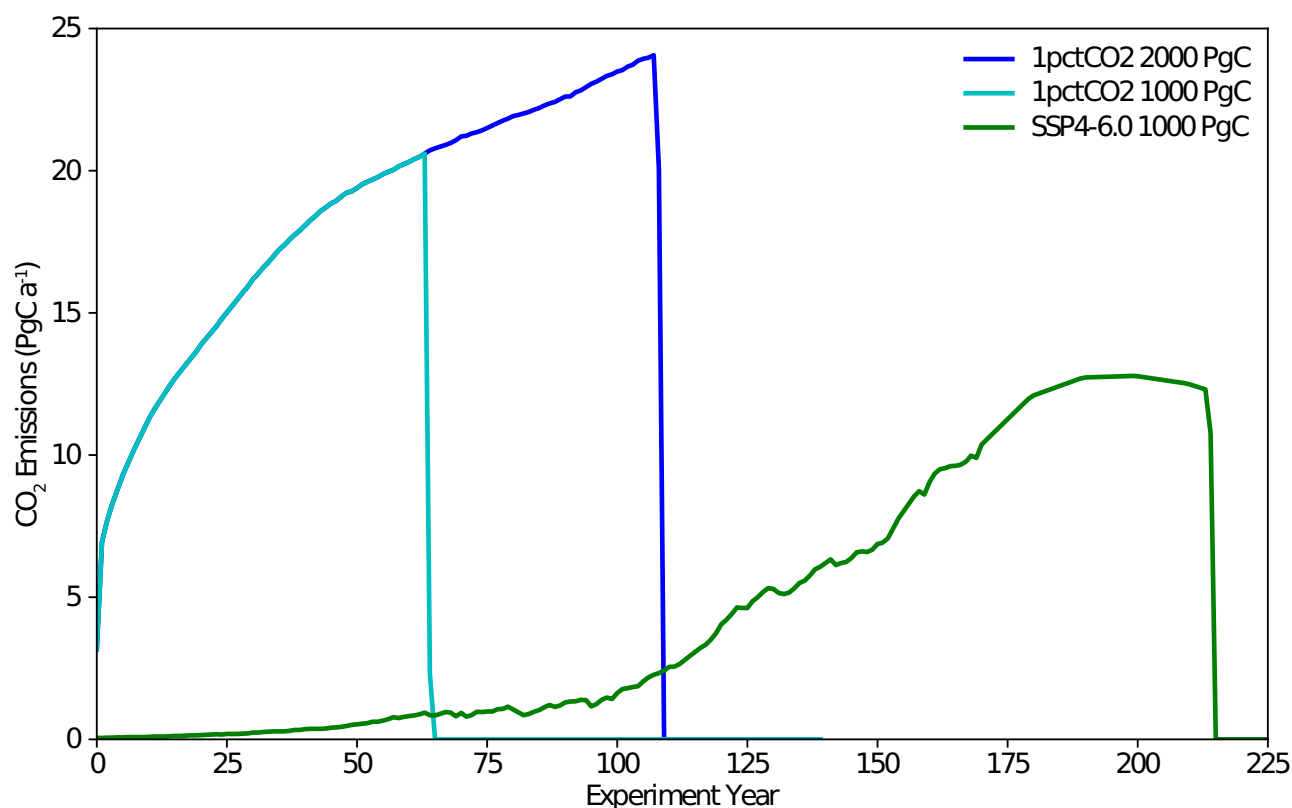
Figure 2. Emissions trajectories for the three experiment sets conducted for this study.

3   Results

Figure 3 shows ZEC for the A1 (1000 PgC) experiment for the model versions with and without permafrost carbon. The figure shows that over centennial timescales the model version with permafrost carbon has a higher ZEC. Consistent with a higher ZEC, the model version with permafrost carbon exhibits a slower decline in atmospheric CO_2 concentration after emissions

190   cease. Figure 4 a displays the difference in ZEC between the simulations with and without permafrost carbon for the A1 (1000 PgC) experiment. Fifty years after emissions cease the existence of a permafrost carbon pool has added 0.06 [0.02 to 0.14]°C to ZEC (median [5th to 95th percentile]), rising to an addition of 0.09 [0.04 to 0.21]°C 100 years after emissions cease, and 0.27 [0.12 to 0.49]°C 500 years after emissions cease. The additional warming is being driven by release of carbon from permafrost soils which totals 29 [10 to 90] PgC by the time emissions cease, 73 [32 to 190] PgC by 50 years after emissions cease, 100

195   [46 to 222] PgC 100 years after emissions cease, and 178 [70 to 346] PgC 500 years after emissions cease (Figure 5 a). The carbon released from permafrost soils by the time emissions ceases causes 0.04 [0.01 to 0.12]°C of additional warming in the model versions with permafrost carbon.

Figure 6 shows the difference in ZEC between simulations following the A1 (1000 PgC) tier-1 ZECMIP experiment protocol and the experiment where emissions follow historical and SSP4-6.0 emissions until 1000 PgC of CO_2 has been emitted. In the

200   SSP4-6.0 based experiment ZEC is 0.04 [0 to 0.06]°C, 0.03 [-0.01 to 0.05]°C and 0.03 [-0.02 to 0.08]°C warmer 50, 100, and 500 years after emission cease. The slightly larger ZEC is being driven by additional carbon being released from permafrost soils under the SSP4-6.0 based experiment with 51 [22 to 132] PgC released when emissions cease compared to 29 [10 to 90] PgC in the A1 experiment. 50, 100 and 500 years after emission cease 84 [39 to 204] PgC, 107 [50 to 230] PgC, and 180 [71 to 354] PgC have been released under the SSP4-6.0 based experiment compared to 73 [32 to 190] PgC, 100 [46 to 222] PgC, and

205   178 [70 to 346] PgC under the A1 experiment. Notably the effect of the thaw-lag diminishes with time after emissions cease.

The ZEC for model variants without and with permafrost carbon for the A3 (2000 PgC) experiments are shown in Figure 7. Similar to the A1 experiment the existence of a permafrost carbon pool adds to the magnitude of ZEC. The difference between the model versions with and without permafrost carbon is shown in Figure 4 b for the A3 experiment. The difference in ZEC 50, 100, and 500 years after emissions cease is 0.06 [0.03 to 0.12]°C, 0.09 [0.05 to 0.18]°C, and 0.24 [0.11 to 0.50]°C,

210   corresponding to a release of carbon from permafrost soils of 84 [40 to 213] PgC when emissions cease and 159 [85 to 300] PgC, 205 [114 to 354] PgC, and 312 [148 to 505] PgC 50, 100 and 500 years after emissions cease (Figure 5 b). Consistent with previous results, that examined representative concentration pathway scenarios, the temperature effect of the permafrost carbon cycle feedback is not strongly effected by the total cumulative emissions (e.g. MacDougall et al., 2012; Schneider von Deimling et al., 2015).

215   ## 4   Discussion

ZECMIP found that the inter-model range of ZEC 50 years after emissions cease for the A1 (1000 PgC) experiment is -0.36 to 0.29°C with a median value of -0.05°C (MacDougall et al., 2020). Thus the additional warming expected from the permafrost carbon cycle feedback of 0.06 [0.02 to 0.14]°C directly and 0.04 [0 to 0.06]°C from the thaw-lag effect will not substantially
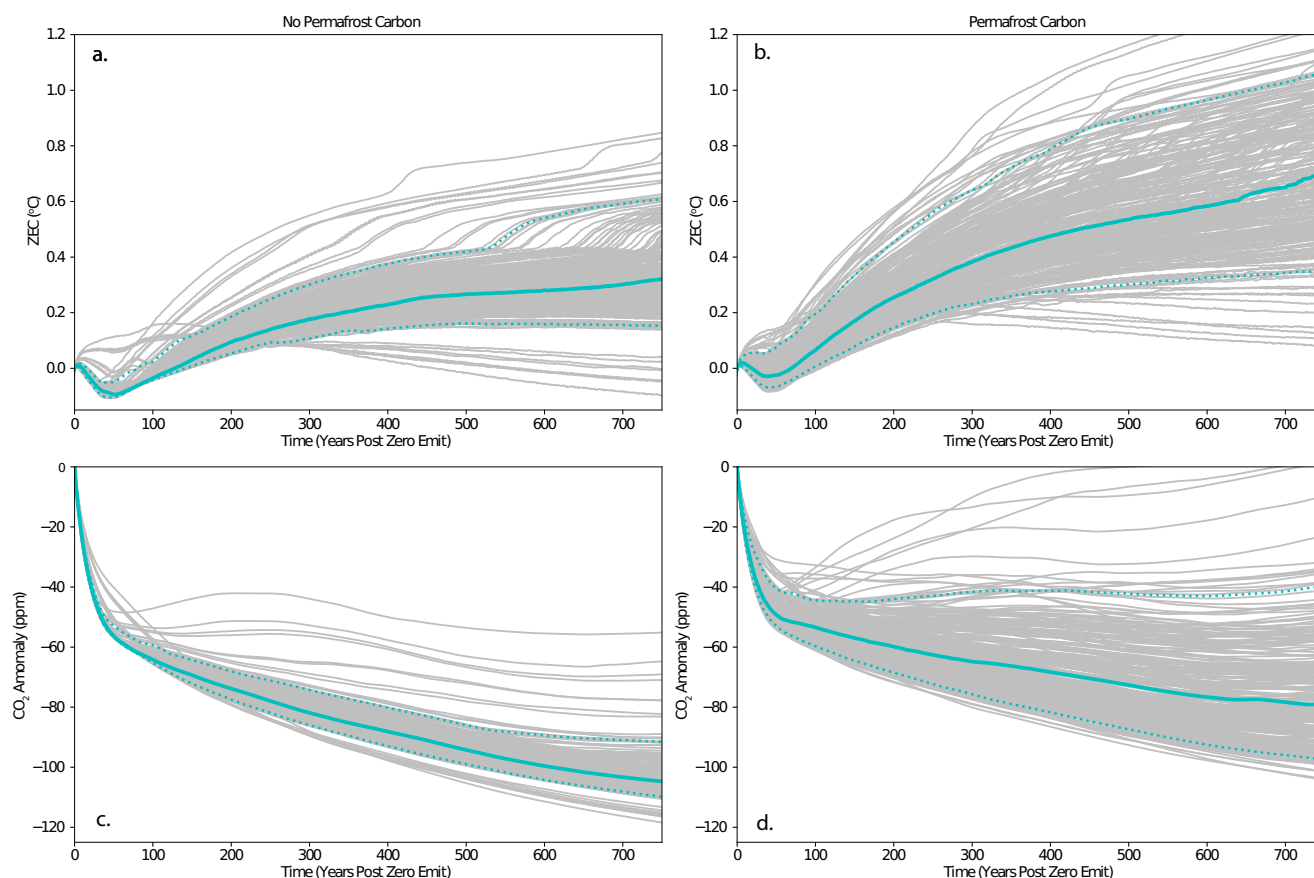
Figure 3. (a,b) Zero Emission Commitment (ZEC) for model versions without (a) and with (b) a permafrost carbon pool forced by the A1 (1000 PgC) experiment. ZEC is temperature change relative to the year emissions cease. (c,d) Change in atmospheric CO_2 concentration relative to year emissions cease for model versions without (c) and with (d) a permafrost carbon pool, forced by the A1 (1000 PgC) experiment. Grey lines are individual model variants, solid line is the median of the variants and dashed lines are the fifth and ninety-fifth percentile. Sudden increase in warming rate seen in many warmer model variants is associated with the disappearance of perennial sea-ice in the Weddell and Ross seas and concurrent changes in overturning circulation.
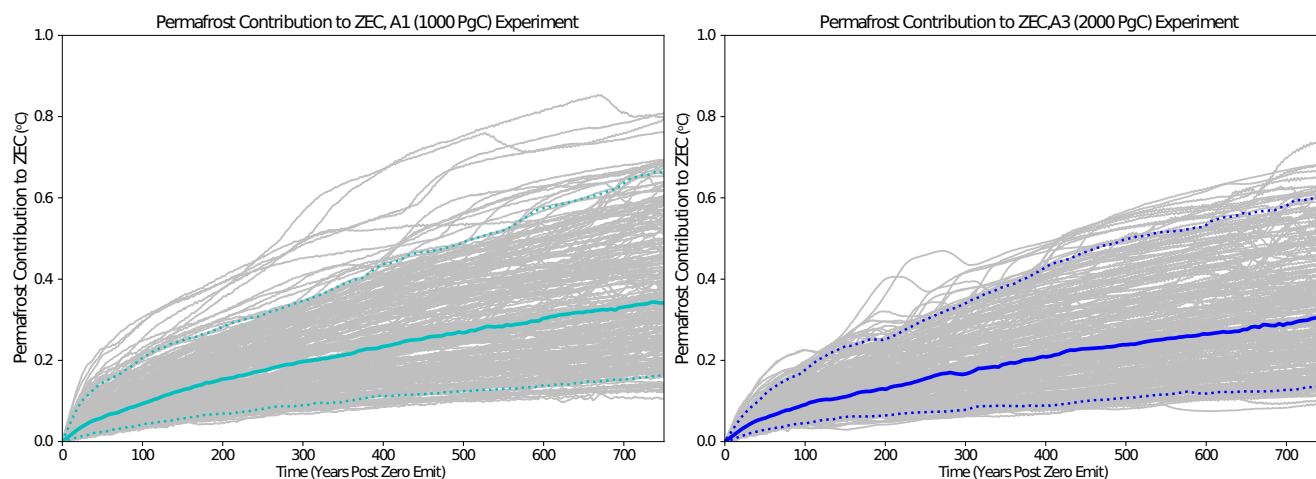
Figure 4. Difference in ZEC between experiments with and without permafrost carbon. (a) Anomalies for the A1 (1000 PgC) experiment. (b) Anomalies for the A2 (2000 PgC) experiment. Grey lines are individual model variants, solid line is the median of the variants and dashed lines are the fifth and ninety-fifth percentile.
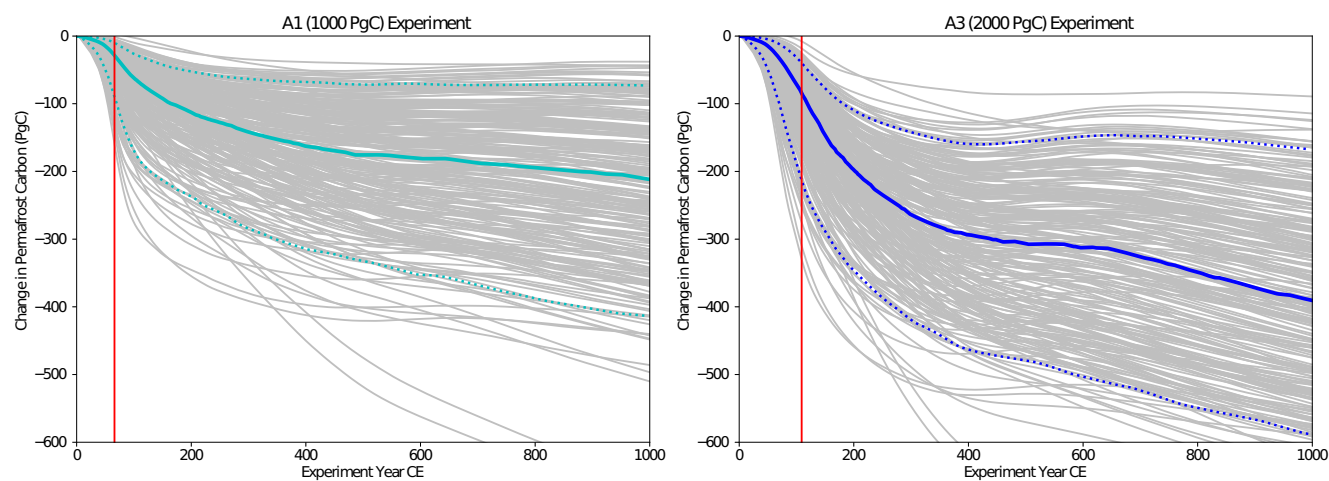
Figure 5. Change in the soil carbon held in permafrost regions relative to pre-industrial size.(a) Change for the A1 (1000 PgC) experiment. (b) Change for the A2 (2000 PgC) experiment. Grey lines are individual model variants, solid line is the median of the variants and dashed lines are the fifth and ninety-fifth percentile. Vertical red line marks the time emissions cease.
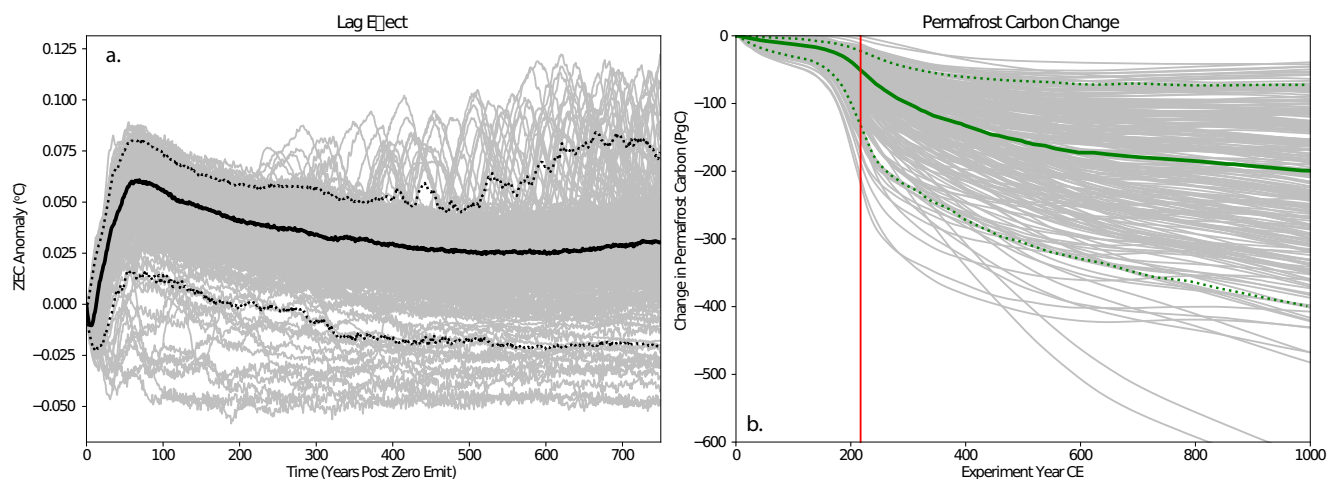
Figure 6. (a) Difference in ZEC between the SSP4-6.0 based 1000 PgC experiment and the standard ZECMIP A1 (1000 PgC) experiment. (b) Change in the soil carbon held in permafrost regions relative to pre-industrial size under the SSP4-6.0 based experiment. Vertical red line marks the time emissions cease. Grey lines are individual model variants, solid line is the median of the variants and dashed lines are the fifth and ninety-fifth percentile.

change the expected value of ZEC of decal time-scales. Thus, the overall conclusion that ZEC will be close to zero in the

220 decades following cessation of emissions remains unchanged.

Here we found that the permafrost carbon feedback contribution to ZEC was insensitive to cumulative CO_2 emissions, despite a larger release of CO_2 to the atmosphere from permafrost soils in the A3 (2000 PgC) experiment. The linear relationship between cumulative emissions of CO_2 and global temperature change is generated by an atmosphere-ocean phenomena (MacDougall, 2017), thus the atmosphere-ocean system does not distinguish between emissions from the terrestrial biosphere and

225 fossil fuel emissions (Simmons and Matthews, 2016). Therefore this insensitivity of temperature to carbon released from permafrost soils appears anomalous. However, for intermediate complexity models like the UVic ESCM the cumulative emissions of CO_2 versus global temperature change curve is only approximate linear and the change in temperature with a unit of CO_2 emitted declines at high cumulative emission totals (MacDougall and Friedlingstein, 2015). Thus the emissions of CO_2 from permafrost soil is less effective at warming after 2000 PgC of CO_2 has been emitted than when 1000 PgC has been emitted

230 in the UVic ESCM. However, it has been shown the full Earth system models do not have non-linear cumulative emissions of CO_2 versus global temperature change curves (Tokarska et al., 2016). Therefore the insensitivity of the permafrost carbon feedback contribution to ZEC to cumulative CO_2 emissions found here should be treated with caution.

To date no Earth system model (McGuire et al., 2018) accounts for abrupt thaw processes in permafrost systems. These processes including thermokarst production, active hill slope erosion, and coastal erosion, could accelerate thaw processes by

235 40% over the next few centuries (Turetsky et al., 2020). Additionally the 2.9pf version of the UVic ESCM (used here) and publicly available 2.10 version of the UVic ESCM do not account enhanced methane production from permafrost thaw. A
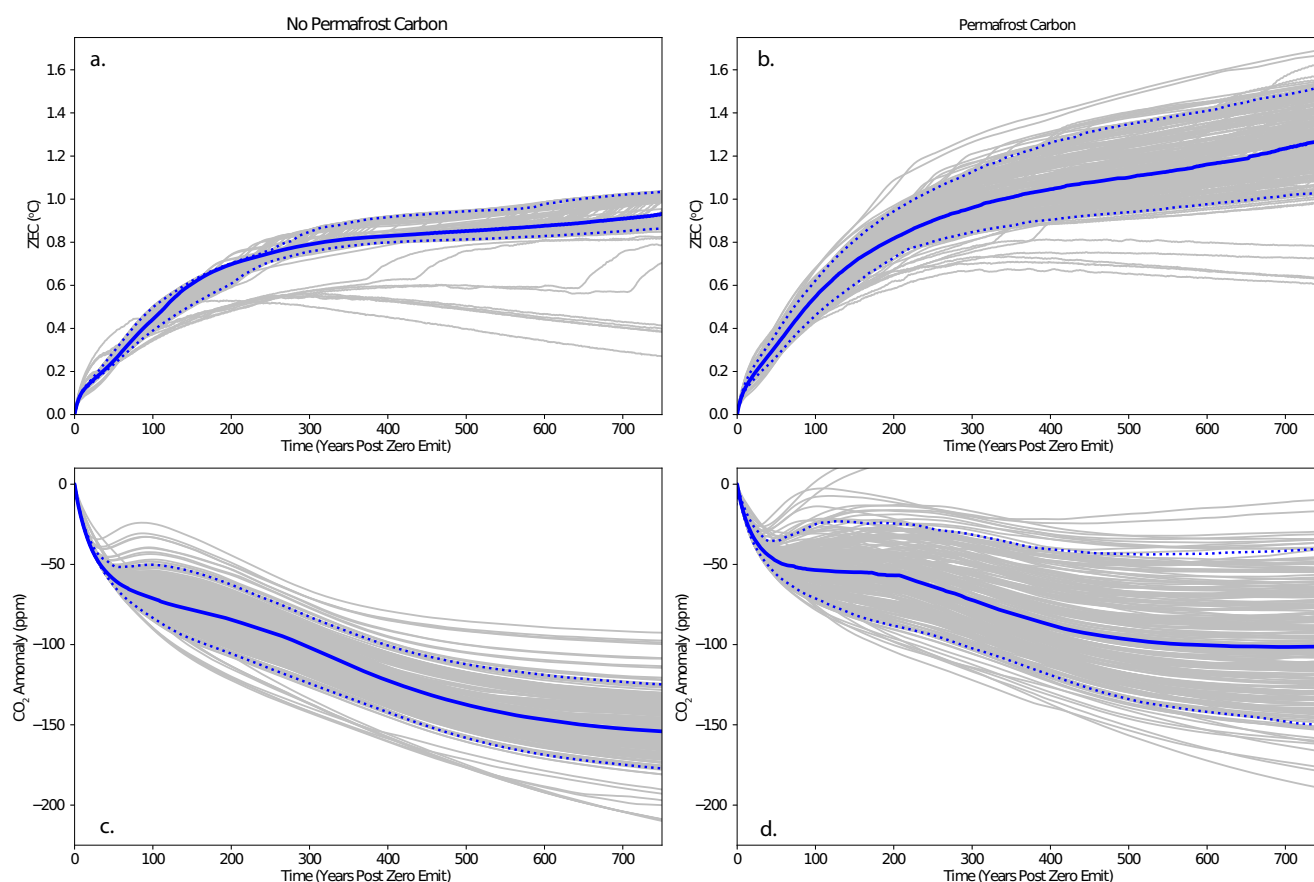
Figure 7. (a,b) Zero Emission Commitment (ZEC) for model versions without (a) and with (b) a permafrost carbon pool forced by the 2000 PgC A3 experiment. ZEC is temperature change relative to the year emissions cease. (c,d) Change in atmospheric CO_2 concentration relative to year emissions cease for model versions without (c) and with (d) a permafrost carbon pool forced by the A3 (2000 PgC) experiment. Grey lines are individual model variants, solid line is the median of the variants and dashed lines are the fifth and ninety-fifth percentile.

methane production scheme has recently been added to a newly developed thread of the model which preliminarily suggests a warming effect from CH_4 production from permafrost soils of 0 to 0.24°C, depending on parameter values and scenario followed (Nzotungicimpaye, 2021). Such values are consistent with expert assessment of (Schuur et al., 2015). Accounting

240   for these processes will likely increase the estimated effect of the permafrost carbon cycle feedback on ZEC and therefore the effect of the permafrost carbon feedback on ZEC should be reassessed when these processes are better accounted for in Earth system models.

UVic ESCM 2.10 was one of the two models that participated in ZECMIP that included a permafrost carbon scheme (the other was CESM). The ZEC 50 years after CO_2 emissions cease for the A1 experiment (1000 PgC) for the model version with

245   permafrost carbon used here is -0.02 [-0.07 to 0.08]°C. The equivalent metric for UVic ESCM 2.10 from ZECMIP was 0.03°C (MacDougall et al., 2020). 750 years after emission cease ZEC is 0.70 [0.35 to 1.06]°C for the A1 experiment for UVic ESCM 2.9pf and was 0.20°C for UVic ESCM 2.10 in MacDougall et al. (2020). Evidently the two model versions have similar decadal ZEC values but diverge substantially on centennial time scales. The main difference between UVic ESCM versions 2.9pf and 2.10 is the representation of the ocean (Mengis et al., 2020), with the newer version of the model having substantially improved

250   ocean dynamics and a state-of-the-art representation of ocean biogeochemistry (Mengis et al., 2020). Ocean heat and carbon uptake are two of the processes the determine the value of ZEC (MacDougall et al., 2020), therefore it is not unexpected that differences in the representation of the ocean would change the ZEC value of a model.

5   Conclusions

Here we have used a perturbed parameter ensemble with the UVic ESCM to estimate the impact of the permafrost carbon

255   cycle feedback on the value of the CO_2 component of the Zero Emissions Commitment. We find that assessed 50 years after emissions cease in an experiment where 1000 PgC of CO_2 is emitted to the atmosphere that the permafrost carbon feedback adds 0.06 [0.02 to 0.14]°C to ZEC, rising to 0.27 [0.12 to 0.49]°C 500 years after emissions cease. Additionally following a more realistic emissions trajectory based on historical and SSP4-6.0 emissions adds 0.04 [0 to 0.06]°C to ZEC 50 years after emissions cease. This thaw-lag effect diminishes with time after emissions cease. Overall accounting for the permafrost carbon

260   feedback is does not change the conclusions the ZEC will be close to zero on decadal time scales (MacDougall et al., 2020), though the effect of abrupt thaw remains unaccounted for and the feedback is of greater concern over longer timeframes.

Code availability. UVic ESCM 2.9 and 2.10 are available from: http://terra.seos.uvic.ca/model/

Data availability. Model output produced for this study is available at: https://doi.org/10.5683/SP2/I75BZ0

Appendix A: Probability Distribution Function for Climate Sensitivity

265 The functional form of the climate sensitivity PDF was taken to be the product of two Normal Inverse Gaussian functions following Olson et al. (2012). The Normal Inverse Gaussian function is:

$$PDF = \frac{\alpha \delta K_1(\alpha \sqrt{\delta^2 + (x-\mu)^2})}{\pi \sqrt{\delta^2 + (x-\mu)^2}} e^{\delta \sqrt{\alpha^2 - \beta^2} + \beta(x-\mu)}, \tag{A1}$$

where μ is location, α is tail heaviness, β is an asymmetry parameter, δ is a scale parameter and K_1 is a modified Bessel function of the third kind (Olson et al., 2012). The Python $scipy.stats$ software package was used to compute the PDF.

270 The new parameter values fitting the PDF to the Sherwood et al. (2020) constraints on equilibrium climate sensitivity are shown in Table A1 below.

Table A1. Parameter values for Climate sensitivity PDF composed of the product of two Normal Inverse Gaussian (NIG) functions. An additional scaling parameter of 5.9047 is needed to make the integral of the PDF 1.

Parameter	NIG –1	NIG – 2
α	2.97	2.94
β	2.90	2.00
δ	1.11	1.08
μ	1.78	2.38

Appendix B: Effect of existence of permafrost carbon pool on simulated global climate
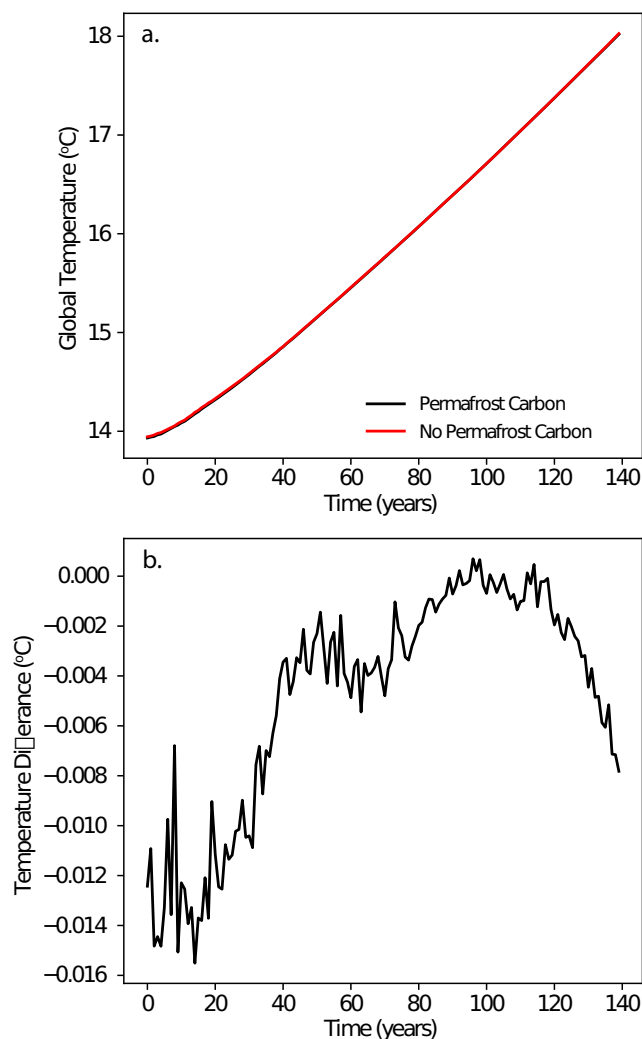
Figure B1. Simulations of the 1pctCO2 experiment for two versions of the UVic ESCM one with default settings (permafrost carbon) and a second starting from a model spin-up where cryoturbation diffusion parameter has been set to zero and hence there is no permafrost carbon pool. Since the 1pctCO2 is a concentration driven scenario the presence or absence of permafrost carbon does not have an effect on the atmospheric carbon pool.

Competing interests. I declare no competing interests.

Acknowledgements. I am grateful for support from the Natural Sciences and Engineering Research Council of Canada Discovery Grant
275 program, and for computational support from Compute Canada. C. Jones and C. Koven provided helpful critiques of an early draft of the
manuscript.

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