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Contrasting CO₂ concentration discharge dynamics in headwater streams: a multi-catchment comparison

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1. Abstract

Aquatic CO$_2$ concentrations are highly variable and strongly linked to discharge but until recently measurements have been largely restricted to low-frequency manual sampling. Using new in-situ CO$_2$ sensors we present concurrent, high-frequency (<30-min resolution) CO$_2$ concentration and discharge data collected from five catchments across Canada, UK and Fennoscandia to explore concentration-discharge dynamics; we also consider the relative importance of high flows to lateral aquatic CO$_2$ export. The catchments encompassed a wide range of mean CO$_2$ concentrations (0.73 – 3.05 mg C L$^{-1}$) and hydrological flow regimes from flashy peatland streams to muted outflows within a Finnish lake-system. In three of the catchments CO$_2$ concentrations displayed clear bimodal distributions indicating distinct CO$_2$ sources. Concentration-discharge relationships were not consistent across sites with three of the catchments displaying a negative relationship and two catchments displaying a positive relationship. When individual high flow events were considered, we found a strong correlation between both the average magnitude of the hydrological and CO$_2$ response peaks, and the average response lag times. An analysis of lateral CO$_2$ export showed that in three of the catchments the top 30% of flow (i.e. flow that was exceeded only 30% of the time) had the greatest influence on total annual load. This indicates that an increase in precipitation extremes (greater high-flow contributions) may have a greater influence on the flushing of CO$_2$ from soils to surface waters than a long-term increase in mean annual precipitation, assuming source limitation does not occur.
Soils represent an important and dynamic store of global carbon which interacts with the atmospheric carbon pool either through direct soil-plant-atmosphere exchange, or transport to and subsequent loss from the surface drainage system. Until recently, much of the literature has focused on the first of these pathways often ignoring losses through the drainage system [e.g. Baldocchi et al., 2001; Bubier et al., 2002; Lafleur et al., 2003]. We currently have a relatively good understanding of the dynamics and hydrochemical processes that control concentrations and fluxes of dissolved organic carbon (DOC) in flowing surface waters [Clark et al., 2007; Hope et al., 1994; McDowell and Likens, 1988]. Whilst the relative importance of gaseous evasion from surface waters to total catchment budgets is now recognised [Butman and Raymond, 2011; Dinsmore et al., 2010; Huotari et al., 2011; Nilsson et al., 2008; Richey et al., 2002], the processes which control temporal and spatial variability of CO$_2$ concentration (and hence the magnitude of the flux) are still not fully understood. Gaseous evasion therefore represents a significant source of uncertainty in greenhouse gas accounting. To fully understand catchment carbon budgets, all flux pathways need to be accounted for. Although lateral CO$_2$ export is often of a smaller magnitude than vertical evasion [Dinsmore et al., 2010; Wallin et al., 2013], it represents a loss of C from the terrestrial to the aquatic system and is therefore an important term to both quantify and understand.

CO$_2$ supersaturation is common in most natural drainage networks across boreal, temperate and tropical systems [Aufdenkampe et al., 2011; Cole et al., 1994; Cole et al., 2007; Kling et al., 1991; Richey et al., 2002]. Estimates of CO$_2$ evasion rates from running waters (expressed per unit water surface area) cover a wide range of values, e.g. 1.65 ± 0.21 µg C m$^{-2}$ s$^{-1}$ from Arctic rivers [Kling et al., 1992] to 263 ± 76.1 µg C m$^{-2}$ s$^{-1}$ in Amazonian tropical...
forests [Richey et al., 2002], 21-806 µg C m\(^{-2}\) s\(^{-1}\) in Scottish peatland catchments [Dinsmore et al., 2010; Hope et al., 2001; Billett and Harvey, in press], and typically 10-300 µg C m\(^{-2}\) s\(^{-1}\) in boreal streams [Wallin et al., 2011]. Contributions of stream/river CO\(_2\) evasion to total catchment budgets have been estimated to equal up to 50% of net annual carbon accumulation in Arctic tundra [Kling et al., 1991], up to 70% in peatlands [Hope et al., 2001] and roughly equal to net annual carbon accumulation in the central floodplain region of the Amazon [Richey et al., 2002].

The rate of gaseous evasion from surface waters is dependent on the solubility of the gas, the physical rate at which molecules can exchange across the water-air interface (given by the gas transfer coefficient or gas transfer velocity), and the water-air concentration gradient. Understanding the temporal dynamics and controls on CO\(_2\) concentrations is therefore the essential first step in accurately quantifying and understanding evasion fluxes.

Aquatic CO\(_2\) can be derived from biogeochemical processes in the bedrock-soil system (weathering, decomposition of organic matter, root respiration) with dissolved CO\(_2\) transported directly to the stream channel by runoff [Billett et al., 2007; Johnson et al., 2007]. In-stream bacterial and photo-chemical degradation of either terrestrial or aquatic derived substrate is an additional source of CO\(_2\) [Köhler et al., 2002]. The relative importance of terrestrially-derived carbon is both seasonal and site-specific, with in-stream productivity restricted by cold temperatures, short in-stream residence times, low stream water pH and stream bed characteristics [Dubois et al., 2010; Raymond et al., 1997; Zeng and Masiello, 2010]. Both the rate of carbon transport into the drainage system and the source-contribution area within the catchment are also highly temporally variable over shorter timescales in response to precipitation events [Dinsmore and Billett, 2008; Rasilo et al., 2012].
The dominant hydrological flowpath through the catchment is dependent on the current and antecedent precipitation conditions. Hydrological flowpath dynamics control the areas within the catchment which are connected to the surface drainage network at any point in time. Due to the high degree of heterogeneity in biogeochemical processes within the terrestrial system and the resulting heterogeneity in carbon form and concentration, runoff chemistry is highly linked to source area and catchment flowpath dynamics [e.g. Chapman et al., 1999; Nyberg, 1995; Wolock et al., 1990]. The location of soil types within the catchment also significantly influences the degree to which they contribute to stream water concentrations. Stream water concentrations of biogenic CO₂ are generally higher where organic soils (rather than mineral soils) form alongside stream channels [Wallin et al., 2010]. The half life of in-stream CO₂ can be as little as a few hours [Öquist et al., 2009] and the contribution from upstream areas can drop quickly with distance from sampling location [Rasilo et al., 2012]. Catchment contributing area is therefore an important consideration when linking in-stream CO₂ concentrations to terrestrial sources, and an important consideration when choosing a sampling location.

Current climate predictions suggest a general increase in precipitation extremes across much of the globe, especially across tropical and northern regions where mean annual precipitation is also expected to increase [Pachauri and Reisinger, 2007]. Previous studies have shown that lateral aquatic carbon export is strongly biased towards high flow events [e.g. Dinsmore and Billett, 2008; Dyson et al., 2010; Ojala et al., 2011], so any increase in storm frequency or intensity is likely to impact the total CO₂ export from soils to surface waters. An understanding of stormflow CO₂ dynamics is therefore becoming increasingly important if we want to accurately quantify and predict catchment carbon losses via the aquatic pathway.
Up to now methodological limitations have restricted our knowledge of aquatic CO$_2$ dynamics, which is based on either low frequency manual sampling techniques [e.g. Billett and Moore, 2008; Billett et al., 2004; Dinsmore et al., 2010; Kling et al., 1991] or inferred indirectly from the speciation of dissolved inorganic carbon concentrations [Butman and Raymond, 2011; Maberly, 1996; Neal et al., 1998; Waldron et al., 2007]. Datasets based on manual sampling techniques are inevitably of low temporal resolution and often biased towards low flow conditions, limiting their use for stormflow analysis. Even the indirect methods, which can be used to produce continuous CO$_2$ datasets, rely on alkalinity which itself is often measured on a spot sampling basis and therefore unlikely to provide sufficient information to accurately analyse high flows. The recent adaptation of high-frequency non-dispersive infrared sensors for use in aquatic systems has, for the first time, allowed direct and continuous measurements of CO$_2$ concentrations to be made in flowing water [Johnson et al., 2010].

Previous studies have shown that the relationship between CO$_2$ and dissolved inorganic carbon (DIC) concentrations and discharge are generally negative [Andrade et al., 2011; Billett et al., 2004; Dinsmore et al., 2010; Edwards, 1973; Edwards et al., 1984; Semiletov et al., 2011; Wallin et al., 2010], although the strength and nature of the relationship is highly variable. For example, the role of pH in controlling the speciation of DIC (ratio between free CO$_2$, bicarbonates, carbonates and carbonic acids) was suggested to counteract the reduction in CO$_2$ due to dilution in a number of streams monitored within the Krycklan catchment, Sweden [Wallin et al., 2010]. However, until the recent use of submerged sensor technology relatively few direct measurements have been made across a sufficient hydrograph range to understand specific CO$_2$ stormflow dynamics. Where sensor technology has been utilised the resulting chemographs show previously unseen intricacies, such as CO$_2$ pulses on the falling
limb of the hydrographs [Johnson et al., 2007] or at peak flow [Dinsmore and Billett, 2008]; they also allow the quantification of response lags and determination of total stormflow exports.

Here we combine aquatic time series data collected at 5 different northern hemisphere sites across northern Europe and Canada where CO₂ has been measured using submerged, in-situ, CO₂ sensors during a series of storm events. The aim of this study, which uses consistent methodology and sensor type, is to compare and contrast the CO₂ concentration-discharge dynamics in individual streams and identify whether consistent relationships can be identified across sites. Specifically we aim to test the following hypotheses:

1. CO₂ concentrations are diluted during high flow events resulting in negative concentration-discharge relationships.

2. The form of the concentration-discharge relationship is linked to measurable catchment characteristics such as soil type or flow-duration indices.

3. The magnitude of the CO₂ response during individual storm events is correlated to the magnitude of the runoff response; the CO₂ response can therefore be predicted from hydrograph characteristics.

4. The lateral export of aquatic CO₂ from the upstream catchment area is strongly influenced by stormflow events as the effect of increased runoff counteracts the decrease in stream concentrations.
3. Methods

3.1. Site descriptions

We use data from 5 different study sites (Figure 1 and Table 1): the Malcom Knapp Research Forest (MK) in Maple Ridge, BC, Canada, the Black Burn draining Auchencorth Moss (AM) peatland, Scotland, Cottage Hill Sike (CHS) in the Moorhouse Reserve, England, Svartberget (SV) in the Vindeln Experimental Forests, Sweden, and the northern inlet of Lake Kuivajärvi near Hyytiälä (HY) SMEAR II (Station for Measuring Forest Ecosystem – Atmosphere Relations) field station, Finland. Site abbreviations (in brackets) will be used henceforth to reference individual field sites.

The MK site is located within a 7 ha forested catchment in the coastal western hemlock climatic zone of BC, Canada. The site climatic conditions consist of mild, wet winters (mean January temperature 2.8°C) and warm dry summers (mean July temperature 17.2°C). The mean annual temperature and precipitation are 9.6°C and 2200 mm y⁻¹, respectively [Trubilowicz et al., 2009]. The catchment is dominated by western red cedar (Thuja plicata), Douglas-fir (Pseudotsuga menziesii) and western hemlock (Tsuga heterophylla). The catchment soil is a highly permeable humic podzol consisting of an upper horizon of organic matter (< 10 cm), a sandy loam subsoil, and is underlain by glacial till over granitic bedrock [Scordo and Moore, 2009].

Both AM (335 ha catchment) and CHS (17.4 ha catchment) are temperate oceanic peatland catchments within the UK. Mean annual air temperature and precipitation at AM are 8.1°C and 1155 mm, respectively (provided by M. Coyle, unpublished data, 2012). The mean annual temperature between 1931 and 2006 at Moor House weather station, 620 m from the CHS catchment, was 5.3°C. Mean annual precipitation was 2012 mm (records from 1951-
170 1980 and 1991-2006) [Holden and Rose., 2011]. A typical winter in both AM and CHS will see several snowfall and melt events. AM vegetation is a mix of *Deschampsia flexuosa*, *Eriophorum vaginatum* and *Juncus effusus* covering a base layer of *Sphagnum* mosses. The bedrock geology is Upper Carboniferous/Lower Devonian sandstones with occasional bands of limestone, mudstone, coal and clay overlain by a thick layer of glacial boulder clay. CHS is lithologically similar to AM (Lower Carboniferous limestone, sandstone and shale sequence overlain by glacial boulder clay) with vegetation consisting primarily of *Eriophorum vaginatum*, *Empetrum nigrum*, *Calluna vulgaris* and *Sphagnum capillifolium*.

Both catchments are dominated by histosols. A full carbon budget, including all aquatic carbon species, for AM has previously been published by Dinsmore et al. [2010]; aquatic carbon fluxes for CHS have been published in Holden et al. [2012].

The SV site is located within a 50 ha forested catchment in boreal Sweden with an 8 ha headwater mire, and another 2-3 ha of riparian peat in a 5-10 m wide strip adjacent to the stream. Mean annual air temperature (1980-2008) is 1.7°C with average temperatures in January and June of -9.6°C and 14.6°C, respectively. Mean annual precipitation (1981-2008) is 612 mm, with approximately 168 days of ground snow cover per year [Haei et al., 2010]. The forest vegetation is dominated by Norway spruce (*Picea abies*) and Scots pine (*Pinus sylvestris*) with an understory of *Calluna vulgaris*, *Vaccinium vitis-idaea* and *Vaccinium myrtillus*. The 8 ha mire area is dominated by *Sphagnum* mosses. Soils are primarily podzols on glacial till formed from biotite plagioclase schist and orthogneiss with histosols in the mire area and histic gleysols in the 5-10 m wide riparian zone on either side of the stream. Lateral and evasive fluxes of CO$_2$, DIC and DOC are presented in Wallin et al. [2013].

The HY site drains a large catchment of approximately 700-1000 ha (estimated from contour map) consisting of both forest and mire. Mean annual air temperature and precipitation are
2.9°C, and 692 mm [Ilvesniemi et al., 2010], respectively, with an average of 126 days per year of ground snow cover in 2009 and 2010 compared to an average of 147 days from 2006-2011. The monitoring site is located at the northern inlet of Lake Kuivajärvi (surface area ca. 1 km²), downstream of Lake Saarijärvi (area ca. 30 ha). The length of the stream between the two lakes is ca. 250 m. The vegetation consists of a mixture of Scots pine (Pinus sylvestris) and Norway spruce (Picea abies) with an understory of Vaccinium myrtillus, Vaccinium vitis-idaea and Rhododendron tomentosum in the riparian wetlands, mosses Dicranum polysetum, Hylocomium splendens and Pleurozium schreberi on the uplands, and Sphagnum spp. on the riparian wetlands. Haplic podzols overlie glacial till on granitic bedrock in the upland forests with histosols occurring in the riparian wetlands.

3.2. Methods

CO₂ concentrations at all sites were monitored using GMT220 series non-dispersive infra-red (NDIR) CO₂ transmitters manufactured by Vaisala (Helsinki, Finland), at a temporal frequency of < 30 minutes following the method described in Johnson et al. [2010]. Sensor accuracy is 1.5% of the calibrated range (0-1% CO₂) + 2% of the reading; this correlates to a maximum error of 0.33 mg C L⁻¹ based on the maximum sensor reading measured at the CHS catchment. Sensors were enclosed in water-tight, gas-permeable PTFE membranes, deployed under the water surface (typically within a perforated PVC sleeve for protection) and connected to a datalogger. Sensors were calibrated against known gas standards before and after deployment and corrected as appropriate; no signal drift was evident. Volume fraction outputs from the NDIR sensors were corrected for variations in temperature and pressure (atmospheric and water depth) using the method described in Johnson et al. [2010] and expressed in units of mg CO₂-C L⁻¹, hereafter annotated as mg C L⁻¹.
Discharge and water temperature were measured concurrently with CO₂ concentration at each site. At MK, stream discharge was measured using a recording capacitance probe (TruTrack model WT-HR; Christ Church, New Zealand) located adjacent to a 90° V-notch weir at the watershed outlet. The TruTrack WT-HR sensor was also used to record air and water temperature. Both AM and CHS utilised Level TROLL® water level and temperature sensors (In-Situ Inc.). Discharge was calculated from a curvilinear stage-discharge rating curve (AM $r^2 = 0.97$; CHS $r^2 = 0.99$) built from a series of dilution gauging measurements. During periods of over-banking at CHS (3% of study period), discharge was extrapolated from a correlation with discharge at the nearby Trout Beck gauging station ($r^2 = 0.77$) provided by the UK ‘Environmental Change Network’. Water level and temperature from SV was measured in a dam house with a V-notch weir using Campbell Scientific data loggers equipped with pressure transducers. Discharge was calculated from a stage-discharge rating curve based on a series of manual dilution gauging and bucket measurements ($r^2 > 0.90$). At HY, discharge was calculated using a relationship ($r^2 = 0.84$) between water level monitored continuously at half hour intervals using pressure sensors (Levelogger Gold, Solinst Canada ltd., Gergetown, ON) and manual flow rate measurements (portable water velocity meter, Global Water FP111, Xylem Inc., White Plains, NY). Concentration datasets were not collected simultaneously across sites and therefore vary in both season and length. Monitoring periods for individual sites were: MK April 2007 – October 2008 (463 days); AM October 2007 – February 2008 (133 days); CHS May – September 2009 (107 days); SV April – November 2007 (215 days) and April – October 2008 (182 days); HY March – October 2010 (211 days).

Continuous pH measurements were made alongside CO₂ at both MK (YSI 6000 multi-parameter sonde) and AM (Campbell Scientific CSIM11 pH-probe). A dataset of weekly pH
measurements from 1993 and 2007 at the CHS catchment was provided by the UK environmental change network (ECN). pH at HY was measured weekly in 2010 and pH at SV weekly throughout 2007-2008 [Wallin et al., 2010].

3.3. Data Analysis

All concentrations are expressed in units of mg C L\(^{-1}\) and discharge in L s\(^{-1}\). Site specific export values represent lateral downstream transport calculated as the mean of hourly exports from instantaneous concentration multiplied by instantaneous discharge. Export values are given in units of g C per m\(^2\) of catchment area per year (g C m\(^{-2}\) yr\(^{-1}\)), allowing for comparison between catchments of different size. Vertical CO\(_2\) evasion is not estimated in this study.

Hydrograph characteristic descriptions were based on daily mean discharges over the measurement period; hence they do not necessarily represent long-term flow statistics. Descriptors include Q\(_{50}\), Q\(_{90}\), Q\(_{95}\), Q\(_{10}\) and Q\(_{5}\) defined as the daily mean flow exceeded or equalled 50% (i.e. the median), 90%, 95%, 10% or 5% of the time, respectively. Q\(_{90}:Q_{50}\) ratios were calculated as a measure of low flow characteristics and Q\(_{10}:Q_{50}\) as an additional normalised indication of high flow characteristics.

Concentration-discharge relationships were examined using box plots of CO\(_2\) concentrations within specified discharge exceedence limits (flow sectors), and the ratio of ‘flow weighted mean concentration’ (FWMC) to unweighted CO\(_2\) concentration. FWMC was calculated using equation 1 where \(c_i\) is the instantaneous concentration, \(q_i\) is the instantaneous discharge and \(t_i\) is the time step between subsequent concentration measurements.

\[
\text{FWMC} = \frac{\sum(c_i \times t_i \times q_i)}{\sum(t_i \times q_i)}
\]

(1)
In the box-plot diagrams, the box represents the interquartile range with a line showing median CO₂ concentration. Whiskers extend to the highest/lowest data values within the upper/lower limit defined as 1.5 times the interquartile range. Outliers are defined as any data point beyond the upper/lower whisker limit; only the maximum and minimum outlier values are plotted. Hysteresis was examined by calculating the mean CO₂ concentration within the same discharge exceedence intervals as the box plots, separated into rising and falling limb data points. The statistical significance of the hysteresis was tested using paired T-tests on these discharge exceedence class means.

CO₂ concentrations were modelled at each site utilising discharge and stream water temperature (the only two parameters that were available at the same temporal resolution as CO₂ concentration at all sites). Datasets were Ln-transformed where required to achieve a normal distribution. Lag terms of 1-10 hours were applied to both temperature and discharge datasets; the lagged values were correlated with CO₂ concentrations and the best fit parameter taken forward to multiple regression models. Temperature was included as a model parameter at three temporal scales; a) instantaneous, b) mean over preceding day, and c) mean over preceding week. Interaction terms between discharge and temperature were applied to all three temperature parameters. Model statistics represent the Pearson product moment correlation coefficient (Minitab version 16) comparing modelled vs. measured CO₂ concentrations in a subset of randomly selected data points within the full time series. The size of the subset was defined by the degree of autocorrelation so that when ordered by date no autocorrelation was present. For all sites except HY, a subset of 500 data points was randomly selected, for HY where autocorrelation was particularly strong, a smaller subset of 100 data points was selected.
The classification of individual storm ‘events’ was based on 30-day moving average hourly $Q_{60}$ and $Q_{20}$ values calculated for each stream individually. An event was classified as such if peak discharge exceeded the 30-day average $Q_{20}$. The start and end of the event were classified as the points at which the discharge exceeded and dropped below the 30-day average $Q_{60}$. This classification was devised through a process of adjustment (changing the threshold values) until events identified on visual inspection of all 5 datasets were suitably captured. Rising and falling hydrograph limbs are classified as time steps within the event classifications which are before or after the event hydrograph peaks, respectively.

The following parameters were calculated for all individual events within the 5 discharge datasets: event duration, rising and falling limb durations, peak discharge and ‘time since last’ defined as the time between the start of the current event and end of previous event. The individual CO$_2$ time series within each hydrologically defined event was then examined to identify any significant peaks or troughs. The following parameters were calculated from the CO$_2$ time series: CO$_2$ response defined as either the highest or lowest CO$_2$ concentration depending on whether a peak or trough was identified, relative CO$_2$ response defined as the CO$_2$ response divided by the mean CO$_2$ concentration over the full measurement period, and the concentration-discharge response lag defined as the time between discharge peak and peak CO$_2$ response. Relationships between individual event parameters within each catchment were examined using Pearson’s product moment correlation analysis (Minitab® 16) on datasets transformed to fit a normal distribution. The mean of each event parameter was also calculated for each catchment and the catchment means compared again using correlation analysis.

DIC speciation was calculated using equation 2 where $\alpha$H$_2$CO$_3$ is the proportion of dissolved carbonic acid (including both hydrated and dissolved CO$_2$). $K_1$ and $K_2$ are temperature
dependent dissociation constants calculated from equations 3 and 4 where T is temperature measured in units of Kelvin; constants a, b and c are taken from [Harnard and Davis, 1943; Harnard and Scholes, 1941].

\[ H_2CO_3 = \frac{[H^+]^2}{[H^+]^2 + K_1 + K_2} \]  
\[ K = 10^{-pK} \]  
\[ pK = \frac{a}{T} + bT + c \]

4. Results

Sites differed considerably in their discharge ranges (Table 2; Figure 2), the lowest mean discharge values were recorded at MK and CHS (\(<1\) L s\(^{-1}\)); the highest (97.9 L s\(^{-1}\)) at HY. MK was the only site where discharge \(<0.01\) L s\(^{-1}\) was recorded, this occurred during an extended precipitation-free summer when the stream dried out completely. The highest mean CO\(_2\) concentration was measured at CHS (3.05 ± 0.95 mg C L\(^{-1}\)), it was statistically similar to both AM (2.13 ± 0.17 mg C L\(^{-1}\)) and the SV catchment in 2007 (0.93 ± 1.40 mg C L\(^{-1}\)) (Table 2). Although the mean CO\(_2\) at SV during 2007 was lower than both MK and SV 2008, variability was high (Figure 3). The lowest mean CO\(_2\) concentration was recorded in HY (0.73 ± 0.54 mg C L\(^{-1}\)).

The calculation of DIC speciation based on temperature and pH showed that in all catchments CO\(_2\) was the major form of inorganic carbon (Figure 4). With the exception of HY, CO\(_2\) represented a median proportion of \(>86\)% of all inorganic C species. The proportion of inorganic C represented by CO\(_2\) in HY ranged from 31% to 78%. The greatest interquartile range was seen in AM, which despite a median of 94% CO\(_2\), had a minimum of only 16%
CO₂. Based on the pH and temperature speciation, CO₃²⁻ was not present at any of the sample sites.

4.1. Hydrology

Hydrographs over the full data collection periods are given in Figure 2 with hydrograph characteristics summarised in Table 3. HY had both the highest Q₉₀:Q₅₀ and lowest Q₁₀:Q₅₀ ratios indicating a relatively unresponsive catchment with a high base-flow contribution. The Q₁₀:Q₅₀ ratio suggested CHS is the most responsive/flashiest catchment.

The number of high flow events recorded in the time series ranged from 8 in HY to 23 across the 2 year dataset from SV. However, when corrected for the different time periods, the greatest frequency of events was observed in CHS, followed by AM, SV, MK and HY, respectively (Table 4). The average event (using the classification given above) lasted from 3.1 days in CHS to 12.8 days in HY. The average event duration was directly correlated with the event frequency, following a negative power function (r² = 0.98, P < 0.01). Similarly, both rising and falling limb durations were negatively correlated with event frequency following power functions (Rising r² = 0.93, P = 0.02; Falling r² = 0.98, P < 0.01). In all catchments the falling hydrograph limb was consistently >2 times longer than the rising limb.

4.2. Variability in CO₂ concentrations

CO₂ concentrations were highly variable across all time series (Figure 3); many of the catchments displayed not only large scale spikes and troughs throughout the time series but also higher frequency variability. This high frequency variability was greatest in CHS and HY datasets collected in the summer season and lowest in AM (October to February). Frequency plots of CO₂ concentrations show clear bimodal distributions in MK (frequency
peaks: 1.13 mg C L\(^{-1}\) and 2.42 mg C L\(^{-1}\)), SV (frequency peaks: 0.80 mg C L\(^{-1}\) and 1.02 mg C L\(^{-1}\)) and HY (frequency peaks: 0.59 mg C L\(^{-1}\) and 1.72 mg C L\(^{-1}\)) compared to the positively skewed distributions seen in both AM and CHS (Figure 5). AM and CHS also showed clear clustering of storm concentrations at the lower CO\(_2\) range.

4.2.1. Concentration-discharge relationships

To summarise concentration discharge relationships, the concentration dataset was split into flow sectors based on discharge exceedence levels (Figure 6). The 0-10 discharge classes in Figure 6 represent the highest 0-10% of flow, and the 90-100 class the lowest 0-10% of flow. Hence the median CO\(_2\) concentration decreases with increasing discharge classes in MK, AM and CHS indicating a negative concentration-discharge relationship. The relationship is less clear in MK and CHS due to the large number of outliers. In SV although there are outliers, a clear increase in median CO\(_2\) concentration is evident in response to increasing discharge class indicating an overall positive concentration-discharge relationship. A general increase in CO\(_2\) concentrations with increasing discharge class was observed in HY. However, removal of the 0-10 class, which relates almost entirely to snowmelt, significantly weakened this trend. The high CO\(_2\) anomalies in the HY plot all represent points between 30 March 2010 and 01 May 2010, i.e. during the snowmelt period. CO\(_2\) concentrations were high on the rising limb of the snowmelt event (Figure 3), hence high concentrations from this period were evident across the full range of percentile groups.

Similar discharge dependent relationships were seen when the FWMC was compared to the unweighted mean concentrations (Table 2). A FWMC > unweighted mean indicates a positive concentration-discharge relationship; ratios of >1 were seen in SV (1.14) and HY (1.24), compared to ratios of <1 in MK (0.72), AM (0.68) and CHS (0.53).
When datasets were split into rising or falling hydrograph limbs and the mean concentration within each percentile range plotted, hysteresis was evident (inferred from paired T-tests comparing rising and falling limbs) in 4 of the 5 catchments (Figure 6). Concentrations were significantly greater on the rising compared to the falling hydrograph limbs at the 4 sites. CHS was the only catchment in which hysteresis was not observed.

CO$_2$ concentration models were fitted based on the observed discharge relationships and stream water temperature (Figure 3, Table 5). Discharge was a significant parameter in models across all sites. In both AM and CHS discharge alone explained >70% of variability in the CO$_2$ concentration based on a power function; all other catchments were best modelled using multiple linear regression models. CO$_2$ concentrations at MK were relatively slow to respond to environmental parameters and were best modelled using a negative discharge function with a 9 hour time lag and positive temperature averaged over the preceding week. Both SV and HY displayed positive discharge relationships alongside a significant discharge-temperature interaction. However, whilst in SV the coincidence of high temperatures and high discharges resulted in high CO$_2$ concentrations, the opposite was true for HY.

### 4.2.2. Event Analysis

Table 4 summarises the CO$_2$ responses to individual storm events. The main CO$_2$ responses in MK, AM and CHS were identified as troughs in the CO$_2$ time series; the main responses in SV and HY were identified as peaks. In both MK and AM the response sometimes consisted of a trough with a small additional peak (events 3, 7, 9, 16, 17, 18 in MK and events 2, 3, 4, 8 and 9 in AM Figure 3). In MK the peak occurred just prior to the CO$_2$ dilution response, whereas in AM the peak occurred at the base of the trough; examples of both are given in Figure 7. In some catchments the number of CO$_2$ responses (n) was less than the number of
identified events. Troughs and peaks were identified by visual inspection of the time-series data. For some events, particularly in the HY time series, the response was either masked by consistent diurnal variability or did not exist and could therefore not be included in the analysis. In all catchments except the lake inlet site HY (which was based on a sample size of only 4 events), the magnitude of the CO₂ response (peak/trough) was significantly correlated with the magnitude of the discharge peak (MK $r = 0.55$, $P = 0.03$; AM $r = 0.60$, $P < 0.01$; CHS $r = 0.28$, $P = 0.02$; SV $r = 0.46$, $P < 0.01$; HY $r = 0.38$, $P = 0.24$).

The peak/trough concentration relative to the mean CO₂ concentration (Table 2) was calculated to enable comparison between catchments (Table 4). The greatest relative CO₂ responses were seen in CHS and AM, followed by MK, SV and HY, respectively, collectively showing a negative linear relationship with mean event duration ($r = 0.95$, $P = 0.01$). Within individual catchments, no relationship was seen between individual event duration and the magnitude of the CO₂ response. Furthermore, no relationships were found in any of the catchments between ‘time since last event’ and CO₂ response.

The lag between hydrological response and CO₂ response was greatest in HY, which took an average of ~2 days to react. In contrast the fastest peak CO₂ response occurred in both AM and CHS ~20 min before the hydrograph peak. The variability in lag response time was extremely large among events with all catchments displaying both positive and negative lags; i.e. maximum CO₂ responses before and after peak discharge. Although within individual catchments there was no relationship between event duration and concentration-discharge response lag, when the mean concentration-discharge response lag of the 5 catchments was plotted against mean event duration for each catchment, there was a strong ($r = 0.91$, $P = 0.03$) positive logarithmic relationship. The relationship between concentration-discharge response lag was also strong with rising limb duration ($r = 0.96$, $P < 0.01$).
4.3. High Flow Contribution to Total CO₂ export

Plotting the percent of total CO₂ export against the discharge exceedence probability (the probability that discharge at a randomly selected time point will exceed a specified magnitude) provides a way of assessing the relative importance of different flow sectors (Figure 8a). The point where the plot shows the greatest deviation from the 1:1 line indicates the proportion of flow which is most significant to total CO₂ export. In HY the greatest deviation occurred at approximately 70% equalled or exceeded discharge, and in SV approximately 50%, indicating low flow was proportionally more important to total CO₂ export than high flow. However in the three first order streams (AM, CHS and MK), the greatest deviation and therefore the proportionally most important discharge sector was the top 30% of flow.

CO₂ export is controlled by both CO₂ availability (both external source concentration and in-situ production) and runoff, i.e. the volume of water available to transport the gas from one location to another. By plotting the percentage of total CO₂ export against the percentage of total runoff we can distinguish between the influence of increased water flow and other contributing factors (Figure 8b). The 1:1 line indicates CO₂ export within all flow sectors is controlled solely by discharge. Both HY and SV lie almost completely on top of one another and very slightly above the 1:1 line indicating that runoff was the primary factor controlling export. MK also follows the 1:1 line closely though sits beneath it. Both AM and CHS display concave curves, close to one another but below the 1:1 line.

5. Discussion

We carried out this study in three first order (MK, AM and CHS) and two second order (SV and HY) streams covering a range of flow regimes. The most distinct flow regime was seen
in HY, characterised by the greatest base flow component \( Q_{90}:Q_{50} = 0.55 \) and the smallest
\( Q_{10}\) in relation to median flow \( Q_{10}:Q_{50} = 1.76 \). This muted hydrological response is common
in streams draining lake systems [e.g. Spence, 2006]. HY was also distinct in its DIC
speciation pattern with the lowest proportion of inorganic carbon in the form of free CO\(_2\).
Both AM and CHS (UK peatland catchments) also had relatively high base flow components
but were much more responsive to precipitation events reflecting the high water holding
capacity of peat and the dominance of quick runoff pathways during rainfall events. MK was
the only catchment to record zero flows (due to a highly seasonal precipitation regime), but
again was highly responsive with an important high flow runoff contribution. SV had low
base flow inputs and a \( Q_{10}:Q_{50} \) ratio of only 3.72 (Table 3). Stream flashiness is influenced by
a number of catchment parameters including catchment size, slope, soil type but most notably
in the comparison between the flashier UK streams (AM and CHS) and SV, is the presence of
tree cover in the latter which affects both evapotranspiration and interception, slowing the
runoff response [e.g. Bosch and Hewlett, 1982].

Mean CO\(_2\) concentrations ranged from 0.73 mg C L\(^{-1}\) in HY to 3.05 mg C L\(^{-1}\) in CHS (Table
2); flow-weighted means ranged from 0.91 mg C L\(^{-1}\) in HY to 1.61 mg C L\(^{-1}\) in CHS. The
only distinct grouping based on CO\(_2\) concentrations was the two UK peatland sites, AM and
CHS, where the highest CO\(_2\) concentrations were measured. SV, which displayed the next
highest FWMC, comprised approximately 16% mire [Köhler et al., 2008]. This supports
work from Wallin et al. [2010] who found that peatland coverage was the most important
predictor of aquatic CO\(_2\) concentrations across 14 streams within the 67 km\(^2\) Krycklan
catchment. Although the HY catchment also contained peatland areas it displayed the lowest
FWMC (lower than the peat free MK catchment) possibly as a result of in-lake CO\(_2\)
consumption, which has previously been reported by Kling et al. [2000]. This suggests that
the presence of an upstream lentic environment was more important in controlling aquatic
CO₂ concentrations than catchment characteristics at this site. However the rate at which the
lake CO₂ signal dissipates downstream, and therefore the strength of the signal at a particular
sampling point, is likely to change seasonally in response to catchment inputs, aquatic carbon
cycling and discharge levels. Clearly further work is required to generalise the influence of
lentic systems on downstream CO₂ concentration dynamics.

The greatest interquartile range, and therefore the ‘flashiest’ CO₂ response, was seen in CHS,
followed by AM, MK, HY and SV, respectively (Table 2). Although not a statistically clear
relationship, there appeared to be a link between hydrology and CO₂ variability, with CHS
having the flashiest response for both. Furthermore, SV and HY, the two second order
streams with the most muted response to precipitation events, also displayed the least
variability in CO₂ concentration.

The CO₂ concentrations in HY, MK and SV all showed distinctive bimodal frequency
distributions (Figure 5) suggesting distinct CO₂ sources. The higher CO₂ frequency peak in
HY, centred around 1.72 mg C L⁻¹, could be isolated almost exclusively to the period prior to
1 May 2010 suggesting a linkage to snow melt runoff. This may reflect either CO₂ built up
over the winter period under ice and snow or the direct input of snow melt-water which has
previously been shown to contain high concentrations of CO₂ [Dinsmore et al., 2011a]. Again
the higher CO₂ frequency peak in MK can be linked to a specific time period, in this case
high concentrations relate to a period of low flow during summer 2007 (Figures 2 and 3),
suggesting a significant CO₂-rich base flow component with a median CO₂ concentration of
~2.42 mg C L⁻¹. No clear time period can be linked to the high concentration frequency peak
in SV although there appears to be a dominance of event flow associated with the peak
(Figure 5). We suggest this represents a source area within the catchment that has greater
hydrological connectivity at periods of high flow. Köhler et al. [2008] describes a tributary 20 m upstream of the monitored SV site which became ephemeral during summer low flow and had a median CO₂ concentration more than double that measured at the monitoring site described in this study [Wallin et al., 2010]. The high concentration CO₂ source (Figure 5) may therefore be the forested catchment drained by this tributary. Both AM and CHS have only one frequency peak despite the positively skewed distributions suggesting a more homogenous contributing source area.

5.1. Concentration-discharge relationships

Although DOC concentrations display significant positive concentration-discharge relationships in many northern Hemisphere catchments associated with organo-mineral soils [Hope et al., 1994]; the relationship does not hold true for many peatland streams [e.g. Clark et al., 2008]. In contrast, DIC tends to show negative relationships with discharge [Hope et al., 1994] linked to changing groundwater (weathering) contributions. Here we found that CO₂ concentrations in three of the five catchments also exhibited a negative concentration-discharge relationship.

The negative concentration-discharge relationships in MK, AM and CHS (Figure 6) all suggest a dilution of aquatic CO₂ concentrations at high flow in accordance with previous literature based on both CO₂ and DIC concentrations [Andrade et al., 2011; Billett et al., 2004; Dinsmore et al., 2010; Edwards, 1973; Edwards et al., 1984; Semiletov et al., 2011; Wallin et al., 2010]. The strength of the relationship (as indicated by the ratio of FWMC to unweighted mean concentration) appears to correlate negatively with mean stream water pH (Table 1); although with only 3 catchments this relationship cannot be tested statistically. A similar result was seen by Wallin et al. [2010] within the Krycklan catchment (Sweden) with
the strongest negative relationships observed in the streams with the lowest pH. Streamwater CO₂ concentrations usually represent an integrated signal of multiple hydrological sources. The streamwater concentration is dependent on both the CO₂ concentrations in the source areas and, due to its influence on the carbonate equilibrium, the pH of the source water. Previous studies at AM have shown a strong negative pH-discharge relationship [Dinsmore and Billett, 2008], indicating a greater proportion of DIC is present as HCO₃⁻ during low flow. Despite this, the CO₂ concentration in the streamwater increases at low flow indicating that the CO₂ content of the low flow water source is high enough to mask the influence of speciation. No significant correlation was observed between CO₂ concentration and pH in the MK time series (based on a non-autocorrelated random subsample of 500 data points).

CO₂ concentrations at both AM and CHS were best explained using discharge only (i.e. no temperature parameter), and reacted quickly to changes in discharge levels. In contrast, MK was slower to respond to discharge changes with the best concentration-discharge relationship achieved using a nine hour lag suggesting an initial input of high CO₂ concentration water, prior to dilution by recent event water. The concentration model for MK also included a positive term describing the average temperature over the preceding week. This suggests a link to recent biological activity rather than an influence of temperature dependent solubility, which would result in a negative relationship to instantaneous temperature such as that seen in SV.

Distinct water sources can be either vertically distributed throughout the soil/bedrock profile or represent varying tributary discharge contributions in 2nd order streams such as SV. As previously discussed, the hydrological contribution from the CO₂-rich tributary upstream of the SV monitoring site is minor during summer low flows [Köhler et al., 2008]. The greater proportion of total runoff originating from the CO₂-rich tributary during high flow may
explain the positive concentration-discharge relationship (Figures 4 and 6). This highlights the importance of varying specific discharges from upstream tributaries in understanding the hydrochemistry of higher order streams. It also highlights the importance of understanding water sources during site selection if a specific landscape interaction is being studied.

The weak discharge relationship in HY is most likely explained by increased water residence time in the upstream lake system, which has previously been shown to obscure temporal signals in outlet nutrient dynamics [e.g. Brown et al., 2008]. This causes a disconnect between the outflow stream water concentration and the catchment flowpath dynamics, muting the overall concentration response. Furthermore, the presence of photosynthesis within the upstream lakes can decrease CO$_2$ concentrations [Kling et al., 2000] and cause diurnal oscillations [Hari et al., 2008], which may be a more important source of CO$_2$ variability than discharge responses in these systems. The temperature-discharge interaction model term complicates the interpretation further by suggesting that the co-occurrence of high temperature and high discharge results in lower streamwater CO$_2$ concentrations. Clearly more parameters are needed to fully understand the CO$_2$ concentrations in the lake-stream continuum.

All catchments with the exception of CHS displayed significant hysteresis in concentration-discharge relationships with rising limb concentrations greater than falling limb concentrations; this relationship holds true regardless of the sign of the concentration-discharge relationship. The interaction between CO$_2$ and pH, and CO$_2$ loss through turbulence-dependent evasion, makes the interpretation of CO$_2$ patterns complicated. Furthermore Dinsmore and Billett [2008] showed that even within AM the hysteresis patterns across individual events were variable. Despite the aforementioned complexity, the presence of a significant relationship averaged over the full measurement period suggests that flushing
and source depletion [Johnson et al., 2007] were important controls on CO₂ dynamics in four
of our study catchments.

5.2. ‘Event’ responses

When specific CO₂ responses to individual events were analysed within each catchment the
only clear relationship that emerged was a correlation between discharge peak height and
magnitude of CO₂ response. There was significant variability in CO₂ response among events
which could not be explained with the available data. The classification of an event in this
study was based on parameters that could be applied equally across all catchments. In some
circumstances a single event under our classification actually contained a double hydrograph
peak (particularly in MK, AM and CHS where annual precipitation was high). This may have
led to a break in the linkage between hydrology and CO₂ response and explain the lack of any
clear correlations between event parameters. We used ‘time since last event’ as a surrogate
measure of antecedent conditions but found no correlation with CO₂ response. It is likely that
a more precise measure of antecedent precipitation or soil moisture parameters may show a
better relationship with CO₂ response. Despite the lack of clear relationships between events
within a single catchment, when average event responses were considered across catchments,
significant correlations to hydrological parameters were evident. The relative magnitude of
the main CO₂ response and the concentration-discharge response lag were greatest in
catchments with the shortest overall event duration, i.e. catchments with a flashier hydrology
also showed the flashiest CO₂ responses.

The lag between hydrological response and the response in CO₂ concentrations varied greatly
between individual events. In all catchments both positive and negative lags were seen
indicating a peak CO₂ response on both rising and falling hydrograph limbs. Between
catchments however a strong relationship existed between the response lag and both the
overall event duration and particularly the rising limb duration. In AM and CHS where quick
discharge responses (i.e. short rising limbs) were common, lowest CO$_2$ concentration tended
to occur before peak discharge. The quick runoff response indicated an initial dominance of
surface or near-surface flow pathways. Due to the low catchment residence time of this event
water, interaction with soil CO$_2$ stores is likely to be limited and result in a dilute source to
the stream. As the event water begins to infiltrate, the dominance moves towards a slightly
deeper through-flow component. This slower moving water has a greater opportunity to
incorporate soil-derived CO$_2$, transporting it to the stream channel and raising stream
concentrations. The timing of the switch between surface runoff pathways and deeper
through-flow of infiltrated water will be catchment specific and control the concentration-
discharge response lag.

MK had a much longer rising limb duration suggesting a slower runoff response; this was
also indicated by the 9-hour lag in the model discharge parameter (Table 5). The soil type,
forest cover and catchment shape (long and narrow with the stream channel only in the lower
half of the catchment [Trubilowicz et al., 2009]) are all characteristics that are likely to lead
to greater rates of event water infiltration. Previous studies have shown that infiltrating event
water can cause the displacement of CO$_2$-rich pre-event water [Carey and Quinton, 2004;
Inamdar et al., 2004; Johnson et al., 2007]. In this scenario, the first water to reach the
stream channel would be soil water displaced by piston flow which may explain the presence
of a CO$_2$ peak prior to the dilution response. Once this soil water had been flushed out, the
infiltrated event water would contain lower CO$_2$ concentrations and result in the observed
dilution pattern. We would therefore expect to see a correlation between the
presence/magnitude of the pre-dilution peak and antecedent soil moisture conditions; data was not available to test this hypothesis.

Some of the events within AM displayed secondary CO$_2$ response peaks occurring at the base of the dilution trough (Figure 7). The AM response was previously explained by Dinsmore and Billett [2008] as an indication of variable contributing source areas and linkage to deeper peat at the catchment perimeter. A similar effect may have been occurring in SV which also contains an area of mire, likely to be a high CO$_2$ source, located ~1.1 km upstream from the main stream sampling point. CO$_2$ is also lost along the stream length through vertical evasion, with an estimated ‘half life’ of 5.5 hrs in a headwater stream [Öquist et al., 2009; Wallin et al., 2011]. Therefore the influence of upstream catchment areas and the likelihood of such a secondary peak arising is largely dependent on water residence time.

5.3. Contribution of high flow to total lateral CO$_2$ export

Due to CO$_2$ source heterogeneity within catchment soil profiles, and the changing dominance of runoff flow-pathways across the hydrograph range, the relative importance of high versus low flow to total lateral CO$_2$ export varied between catchments. As current climate models predict not only future changes in total precipitation but changes in precipitation patterns i.e. a move towards fewer rainfall events of greater intensities and magnitudes across northern regions [Pachauri and Reisinger, 2007], the need to understand the impact of changing flow regimes is increasing. Here we show that the proportion of flow with the greatest influence on total lateral export was the top 30% in MK, AM and CHS. These three catchments displayed negative concentration-discharge relationships but had the highest $Q_{10}$:$Q_{50}$ hydrological ratios. At these 3 sites an increase in precipitation extremes is likely to lead to a greater lateral export of aquatic CO$_2$ (within the limits of source CO$_2$ availability) than would be expected
from a simple increase in mean annual precipitation alone. As CO$_2$ in headwater streams is
primarily allochthonous in origin, lateral transport can be considered as a surrogate for the
transport of CO$_2$ from the terrestrial to the aquatic system. Hence an increase in precipitation
extremes is likely to lead to a greater loss of terrestrial CO$_2$ to the aquatic system than an
increase in mean annual precipitation alone. The relative contribution of the different flow
sectors in SV was evenly distributed (Figure 8a), whereas in HY it was clearly the low flow
period that was most important for lateral CO$_2$ export (70% equalled or exceeded).

If we only considered concentration-discharge relationships we would expect high flow to be
proportionally more important where a positive concentration-discharge relationship existed.
Figure 8b illustrates that because of the concentration-discharge relationship, lateral CO$_2$
export across all flow sectors in MK, AM and CHS was less than would have been expected
due to increased runoff alone, the opposite was true in SV and HY. However, we find that
hydrograph characteristics (and the associated catchment characteristics that define the
hydrograph) appear to be more important than the concentration-discharge relationship in
defining the specific flow sector contribution to total lateral export. Although it is difficult to
draw numeric conclusions that can be usefully upscaled from only 5 catchments, we have
identified trends that could be explored empirically in other systems and potentially
modelled. This could lead to a much better understanding of the influence of precipitation
patterns on lateral CO$_2$ export, and more importantly, losses of CO$_2$ from the soil to the
stream system where vertical evasion is the predominant flux pathway.

Current methodological limitations prevent the measurement of gas transfer coefficients at a
temporal resolution suitable to carry out a similar flow sector analysis on CO$_2$ losses through
vertical evasion. As many previous studies have shown that the vertical evasion flux is often
greater than lateral CO$_2$ losses [e.g. Dinsmore et al., 2010; Wallin et al., 2013], increasing the
resolution of gas transfer coefficient measurements should be a future research priority and
would enhance flux estimates particularly at high flow extremes.

5.4. Conclusions

The highest CO₂ concentrations were measured at the two UK peatland dominated sites, AM
and CHS, with the next highest FWMC at SV which contained ~16% mire. This supports
previous work which concluded that peatland coverage was the most important predictor of
aquatic CO₂ concentrations [Wallin et al., 2010, Aitkenhead et al., 1999]. Distinct bimodal
frequency distributions in CO₂ concentration were observed in HY, MK and SV. These
suggest distinct CO₂ sources associated with (1) the snow melt period, (2) well defined parts
of the soil profile such as deep horizons which produce a stronger CO₂ signal at low flow,
and (3) spatially separated sub-catchments with variable tributary discharge contributions.

Previous studies of aquatic C dynamics have shown significant positive DOC concentration-
discharge relationships in organo-mineral soils and negative DIC concentration-discharge
relationships associated with changing groundwater (weathering) contributions [e.g. Hope et
al., 1994]. Here we found inconsistent CO₂ concentration-discharge relationships across sites.
Negative relationships (Hypothesis 1) were seen in MK, AM and CHS suggesting dilution of
aquatic CO₂ at high flow. In contrast, SV displayed a positive concentration-discharge
relationship. No clear relationship was seen in HY which was located within a lake system
and where diurnal oscillations appeared to dominate CO₂ variability. The strength of the
concentration-discharge relationship appeared to correlate negatively to stream water pH in
AM, CHS and MK, however SV and HY illustrate that a more complex range of parameters
is required to accurately predict the form of the concentration-discharge relationship. Since
the range of processes that effect aquatic CO₂ in streams is clearly large, the ability to
measure concentrations *in-situ* and at high-frequency is greatly improving our understanding of its source, transport and delivery to the stream system.

CO₂ variability was strongly linked to hydrological variability with the ‘flashiest’ response in both CO₂ concentration and the hydrograph occurring in CHS, and the least ‘flashy’ responses in CO₂ occurring in SV and HY, which also displayed the most muted stream flow response to precipitation events. Correlations between hydrological and CO₂ responses in both response magnitude and lag were identified. Since concentration-discharge response lags were shortest in catchments with quick run-off responses, we suggest that the timing of the switch between surface runoff pathways (where present) and deeper through-flow of infiltrated water contributes to the concentration-discharge response lag.

We show that the proportion of flow with the greatest influence on total lateral CO₂ export was the top 30% in MK, AM and CHS, indicating that an increase in event flow as a result of an increase in precipitation ‘extremes’ will result in greater transport of terrestrial CO₂ to surface waters than would occur as a result of a simple increase in mean annual precipitation. The disparity in lateral export response between increasing mean precipitation and increasing extremes has important implications for carbon flux predictions based on future climate scenarios. The increase in lateral CO₂ export was linked to event flow runoff proportions rather than concentration-discharge relationships. It may therefore be possible in the future to use hydrographs to predict the relative importance of precipitation ‘extremes’ to CO₂ export from individual catchments.
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Figure 1. Location of study sites

Figure 2. Hydrographs for individual sites during CO₂ monitoring period. Dashed lines refer to upper and lower storm classification thresholds. Numbers refer to individual storms.

Figure 3. CO₂ time series for individual sites. Dashed lines and numbers refer to storm peaks identified in Figure 2.

Figure 4. Box plots showing range DIC speciation ranges across individual sites. The box represents median and interquartile range; the whiskers represent range.

Figure 5. Frequency distributions of CO₂ concentrations identified as either storm (black bars) or non-storm (grey bars) data points.

Figure 6. Box plots of concentrations split into discharge percentile classes with mean rising and falling limb concentrations shown to illustrate hysteresis. T and P-values represent statistical test for hysteresis.

Figure 7. Example of secondary CO₂ peaks from a) MK and b) AM

Figure 8. Plots of percent of total export against a) percent of discharge displayed as exceedence probabilities and b) percent of total runoff.
### Table 1. Site characteristics

<table>
<thead>
<tr>
<th>Location</th>
<th>Vancouver Island (MK)</th>
<th>Auchencorth Moss (AM)</th>
<th>Cottage Hill Sike (CHS)</th>
<th>Svartberget (SV)</th>
<th>Hyytiälä (HY)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Country</td>
<td>Canada</td>
<td>UK</td>
<td>UK</td>
<td>Sweden</td>
<td>Finland</td>
</tr>
<tr>
<td>Lat/Long</td>
<td>49°26' N; 122°55'W</td>
<td>55°47'N; 3°14'W</td>
<td>54°41'N; 2°23'W</td>
<td>64°14'N, 19°46'E</td>
<td>61°50'N; 24°17'E</td>
</tr>
<tr>
<td>Catchment characteristics</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Cathment size (ha)</td>
<td>7</td>
<td>335</td>
<td>17</td>
<td>50</td>
<td>700-1000</td>
</tr>
<tr>
<td>Ecosystem type</td>
<td>Coastal Forest</td>
<td>Peatland</td>
<td>Peatland</td>
<td>Forest/Peat</td>
<td>Forest/Peat</td>
</tr>
<tr>
<td>Primary soil type</td>
<td>Humic Podzol</td>
<td>Histosol (85%)</td>
<td>Histosol</td>
<td>Podzol/Histosol</td>
<td>Haplic Podzol</td>
</tr>
<tr>
<td>Hydrology/Hydrochemistry</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Stream Order</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>Mean Annual Runoff Ratio</td>
<td>44%</td>
<td>77%</td>
<td>81%</td>
<td>50%</td>
<td>36-53%</td>
</tr>
<tr>
<td>Mean pH (range)</td>
<td>5.7 (5.4-6.0)</td>
<td>5.5 (4.3-7.3)</td>
<td>4.3 (3.8-7.2)</td>
<td>5.4 (4.2-6.0)</td>
<td>6.5 (6.0-6.7)</td>
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<tr>
<td>Mean Conductivity (μS cm⁻¹)</td>
<td>22.2</td>
<td>87.8</td>
<td>41.1</td>
<td>28.8</td>
<td>31.6</td>
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<tr>
<td>Climate variables</td>
<td></td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Climate zone</td>
<td>Coastal Western Hemlock</td>
<td>Sub-arctic oceanic</td>
<td>Sub-arctic oceanic</td>
<td>Boreal zone</td>
<td>Boreal zone</td>
</tr>
<tr>
<td>Mean Annual Temp. (°C)</td>
<td>9.6</td>
<td>8.1</td>
<td>5.3</td>
<td>1.7</td>
<td>2.0</td>
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<tr>
<td>Mean Annual Precip. (mm)</td>
<td>2200</td>
<td>1155</td>
<td>2012</td>
<td>612</td>
<td>692</td>
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<tr>
<td>Annual snow cover</td>
<td>NA*</td>
<td>NA</td>
<td>41 days yr⁻¹</td>
<td>168 days yr⁻¹</td>
<td>126</td>
</tr>
</tbody>
</table>

* Snow cover data not available for MK. Snowfall represents about 5% of annual precipitation at MK
Table 2. Summary of monitored stream variables over measurement period. Discharge values represent median and range, and both temperature and CO$_2$ are displayed as mean ± stdev. Groupings a, b and c indicate groups where CO$_2$ concentrations overlap ± 1 standard deviation.

<table>
<thead>
<tr>
<th>Site</th>
<th>Mean Discharge (L s$^{-1}$)</th>
<th>Mean Temp ($^\circ$C)</th>
<th>Mean CO$_2$ (mg C L$^{-1}$)</th>
<th>IQR (mg C L$^{-1}$)</th>
<th>FWMC (mg C L$^{-1}$)</th>
<th>Ratio Mean:FWMC</th>
</tr>
</thead>
<tbody>
<tr>
<td>MK</td>
<td>0.71 (0.51.5)</td>
<td>8.17 ± 2.71</td>
<td>1.47 ± 0.19$^a$</td>
<td>0.65</td>
<td>1.06</td>
<td>0.72</td>
</tr>
<tr>
<td>AM</td>
<td>19.9 (41.5 - 632)</td>
<td>4.36 ± 2.75</td>
<td>2.13 ± 0.17$^b$</td>
<td>1.48</td>
<td>1.45</td>
<td>0.68</td>
</tr>
<tr>
<td>CHS</td>
<td>0.92 (0.67 - 371)</td>
<td>11.3 ± 2.26</td>
<td>3.05 ± 0.95$^b$</td>
<td>1.95</td>
<td>1.61</td>
<td>0.53</td>
</tr>
<tr>
<td>SV</td>
<td>1.45 (0.48 - 137)</td>
<td>5.89 ± 3.19</td>
<td>0.96 ± 0.94$^c$</td>
<td>0.26</td>
<td>1.10</td>
<td>1.14</td>
</tr>
<tr>
<td>HY</td>
<td>97.9 (53.0 - 492)</td>
<td>13.2 ± 7.33</td>
<td>0.73 ± 0.54$^c$</td>
<td>0.35</td>
<td>0.91</td>
<td>1.24</td>
</tr>
</tbody>
</table>
Table 3. Median discharge for all catchments alongside high and low flow hydrograph descriptors described in text

<table>
<thead>
<tr>
<th></th>
<th>MK</th>
<th>AM</th>
<th>CHS</th>
<th>SV</th>
<th>HY</th>
</tr>
</thead>
<tbody>
<tr>
<td>Median (L s$^{-1}$)</td>
<td>0.73</td>
<td>36.0</td>
<td>2.49</td>
<td>2.79</td>
<td>159</td>
</tr>
<tr>
<td><strong>Low flow indices</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$Q_{90}$</td>
<td>0.00</td>
<td>16.7</td>
<td>0.81</td>
<td>0.74</td>
<td>86.6</td>
</tr>
<tr>
<td>$Q_{95}$</td>
<td>0.00</td>
<td>15.3</td>
<td>0.79</td>
<td>0.66</td>
<td>79.4</td>
</tr>
<tr>
<td>$Q_{90}:Q_{50}$</td>
<td>N/A</td>
<td>0.46</td>
<td>0.32</td>
<td>0.27</td>
<td>0.55</td>
</tr>
<tr>
<td><strong>High flow indices</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$Q_{10}$</td>
<td>4.13</td>
<td>156</td>
<td>21.5</td>
<td>10.4</td>
<td>279</td>
</tr>
<tr>
<td>$Q_{5}$</td>
<td>6.24</td>
<td>203</td>
<td>45.4</td>
<td>17.4</td>
<td>380</td>
</tr>
<tr>
<td>$Q_{10}:Q_{50}$</td>
<td>5.68</td>
<td>4.33</td>
<td>8.61</td>
<td>3.72</td>
<td>1.76</td>
</tr>
</tbody>
</table>
Table 4. Event characteristics. CO$_2$ response is defined as either the highest or lowest CO$_2$ concentration depending on whether a peak or trough was identified. Relative CO$_2$ response is defined as the CO$_2$ response divided by the mean CO$_2$ concentration over the full measurement period. The concentration-discharge response lag is defined as the time between discharge peak and peak CO$_2$ response.

<table>
<thead>
<tr>
<th>Hydrological Characteristics</th>
<th>MK</th>
<th>AM</th>
<th>CHS</th>
<th>SV</th>
<th>HY</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total Number Events</td>
<td>18</td>
<td>9</td>
<td>15</td>
<td>23</td>
<td>8</td>
</tr>
<tr>
<td>Events per Month</td>
<td>1.19</td>
<td>2.07</td>
<td>4.27</td>
<td>1.77</td>
<td>1.16</td>
</tr>
<tr>
<td>Event Duration (days)</td>
<td>10.7 ± 1.58</td>
<td>6.24 ± 1.21</td>
<td>3.10 ± 0.50</td>
<td>8.34 ± 4.56</td>
<td>12.8 ± 3.42</td>
</tr>
<tr>
<td>Rising Limb Duration (days)</td>
<td>2.71 ± 0.6</td>
<td>1.64 ± 0.54</td>
<td>0.84 ± 0.23</td>
<td>1.88 ± 1.48</td>
<td>4.01 ± 1.42</td>
</tr>
<tr>
<td>Falling Limb Duration (days)</td>
<td>7.96 ± 1.29</td>
<td>4.6 ± 1.35</td>
<td>2.26 ± 0.37</td>
<td>6.54 ± 4.42</td>
<td>8.81 ± 2.15</td>
</tr>
<tr>
<td>Peak Height (L s$^{-1}$)</td>
<td>12.0 ± 3.48</td>
<td>6.76 ± 0.58</td>
<td>36.4 ± 9.15</td>
<td>10.9 ± 21.8</td>
<td>1.33 ± 0.06</td>
</tr>
</tbody>
</table>

| CO$_2$ Response Characteristics | | | | | |
| Number of responses identified | 16 | 9 | 15 | 23 | 4 |
| Mean Event CO$_2$ (mg L$^{-1}$) | 1.29 ± 0.15 | 2.04 ± 0.18 | 3.04 ± 0.03 | 0.94 ± 0.06 | 0.94 ± 0.08 |
| 1º CO$_2$ peak/trough concentration (mg C L$^{-1}$) | 1.16 ± 0.13 | 1.07 ± 0.05 | 1.37 ± 0.11 | 1.24 ± 0.06 | 0.77 ± 0.07 |
| 2º CO$_2$ peak/trough concentration (mg C L$^{-1}$) | 1.74 ± 0.226 | 1.16 ± 0.06 | --- | --- | --- |
| Relative peak/trough height (mg C L$^{-1}$) | 0.31 | 1.06 | 1.68 | 0.28 | 0.04 |
| Concentration-discharge response lag (hr) | 17.9 ± 9.31 | -0.33 ± 0.8 | -0.33 ± 0.33 | 2.34 ± 7.02 | 47.8 ± 55.5 |


Table 5. Best fit model results for CO₂ concentration based on discharge and water temperature.

<table>
<thead>
<tr>
<th>Best fit model</th>
<th>Parameters/Equation</th>
<th>Coefficient</th>
</tr>
</thead>
<tbody>
<tr>
<td>MK</td>
<td>Linear</td>
<td>Intercept</td>
</tr>
<tr>
<td></td>
<td>(r = 0.86; p &lt;0.01)</td>
<td>0.44</td>
</tr>
<tr>
<td></td>
<td>Ln(Discharge lag 9 hours)</td>
<td>-0.13</td>
</tr>
<tr>
<td></td>
<td>Mean Weekly Temperature</td>
<td>0.12</td>
</tr>
<tr>
<td>AM</td>
<td>Power*</td>
<td>Ln(Discharge)</td>
</tr>
<tr>
<td></td>
<td>(r = 0.96; p &lt;0.01)</td>
<td>α</td>
</tr>
<tr>
<td></td>
<td></td>
<td>37.6</td>
</tr>
<tr>
<td></td>
<td></td>
<td>β</td>
</tr>
<tr>
<td></td>
<td></td>
<td>-2.32</td>
</tr>
<tr>
<td>CHS</td>
<td>Power*</td>
<td>(LnDischarge + 1)</td>
</tr>
<tr>
<td></td>
<td>(r = 0.71; p &lt;0.01)</td>
<td>α</td>
</tr>
<tr>
<td></td>
<td></td>
<td>4</td>
</tr>
<tr>
<td></td>
<td></td>
<td>β</td>
</tr>
<tr>
<td></td>
<td></td>
<td>-0.64</td>
</tr>
<tr>
<td>SV</td>
<td>Linear</td>
<td>Intercept</td>
</tr>
<tr>
<td></td>
<td>(r = 0.67; p &lt;0.01)</td>
<td>1.07</td>
</tr>
<tr>
<td></td>
<td>Ln(Discharge)</td>
<td>0.02</td>
</tr>
<tr>
<td></td>
<td>Temperature</td>
<td>-0.03</td>
</tr>
<tr>
<td></td>
<td>Interaction [Ln(Discharge) * Temperature]</td>
<td>0.01</td>
</tr>
<tr>
<td>HY</td>
<td>Linear</td>
<td>Intercept</td>
</tr>
<tr>
<td></td>
<td>(r = 0.84; p &lt;0.01)</td>
<td>0.58</td>
</tr>
<tr>
<td></td>
<td>Discharge</td>
<td>0.0035</td>
</tr>
<tr>
<td></td>
<td>Interaction [Discharge * Temperature]</td>
<td>-0.00023</td>
</tr>
</tbody>
</table>

* Power function written in format $y = \alpha.x^\beta$ where $\alpha$ and $\beta$ are model specific constants