Analysis of loadings

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1   Data

Description   The data in this report comes from four Wisconsin streams that were monitored (with some gaps in data collection) between 1989 and 2007. The streams and the period during which each was monitored are:

Stream	Monitored
Eagle	1991-1994, 2003-2007
Joos Valley	1990-1994, 2002-2007
Otter	1990-1997, 2000-2002
Brewery	1989, 1994-2002, 2004-2005

Each entry in our data set represents one loading event, which is defined based on the hydrograph - the event begins when the loading rises from a base level toward a peak, and ends when the loading falls back to its new base level. Two kinds of load are measured for each event - the sediment load and the phosphorus load. There are two typical ways that sediment and phosphorus get into streams - they can be carried either by runoff during a rainstorm or by melting snow.

Not all of the data can be collected for each event. For instance, rainfall is measured only when the ground is free of snow because snow interferes with the rain gauges. And the amount of snowmelt is estimated by multiplying the water content of the snow by the change in snow depth a warm snap, which is inaccurate when additional snow falls during that event. Broadly, there is one set of measurements that are made during rainfall-driven events and a different set of measurements that are made during snowmelt-driven events. Because of this, the two types of event are modeled separately.

Exploratory Analysis   Our anaysis targets the phosphorus and sediment loads carried by each stream. Using Rainmaker software, each load can be broken into two parts: base load and storm load. We will consider models of the storm load and of the total load.

The first task was to decide whether most of the annual load is produced by snowmelt-driven or rainfall-driven events. The total loads from each kind of event are tabulated in Tables 1 (sediment) and 2 (phosphorus). Figure 1 presents the same information as the tables, while Figure 2 also compares the load from indivudual snowmelt- and rainfall-driven events. About two thirds of the sediment loading comes from rainfall-driven events, along with about half of the phosphorus loading.

We investigated dividing the snow-free seasons into early and late subseasons, separating the two on May 15th of each year. If vegetation serves to hold the soil together, and to increase both evapotranspiration and infiltration, then erosion may be more common early in the spring before most of the summer's vegetation appears. If so, the relationship between rainfall and the stream's loading might change during the summer.

The investigation was done by making linear models to describe the sediment and phosphorus loading during the two subseasons and comparing them to a single model fit to the entire snow-free period. Because the split makes the model more flexible, it will certainly improve the model's fit - the question is whether that improvement is enough to justify making the model more complex. At all four streams, the model improvement was statistically significant but too small to matter (the split models explained about 1%-2% more of the loads). We will not use the split in the rest of the analysis.

	snowmelt-driven	rainfall-driven
eagle	27.0%	73.0%
joosvalley	26.9%	73.1%
otter	35.4%	64.6%
brewery	32.8%	67.2%

Table 1: Proportion of total suspended solids loading contributed by each type of event

	snowmelt-driven	rainfall-driven
eagle	32.8%	67.2%
joosvalley	36.4%	63.6%
otter	46.5%	53.5%
brewery	49.6%	50.4%

Table 2: Proportion of total phosphorus loading contributed by each type of event

Over the course of the monitoring period, the majority of the total load (both of sediment and of phosphorus) was carried during just a few major events. Just 10% of the events carried between 73.1% (at Otter) and 93.4% (at Joos) of the total sediment load; the same events produced between 64.6% (at Otter) and 88% (at Joos) of the total phosphorus load.

2   Analysis

2.1   Variable selection

In order to make a model of the load carried by the stream, we need to select the predictor variables that have explanatory power. We used stepwise regression with the Bayesian Information Criterion (BIC) to screen the potential predictor variables.

Rainfall-driven events   The predictors that survived the screening at each stream are listed in Table 4. The variables are listed in the order of their importance to the model.

In every case, the theisen rainfall is the most important predictor, followed usually by antecedent baseflow. Using just those two predictors results in an R^2 greater than 0.7 in most models (the exception is at Brewery Creek - see Table 3.) Since the antecedent baseflow is considered an indicator of how wet is in the watershed before each event, we conclude that the amount of sediment and phosphorus washed into a stream by each event is mainly a function of the quantity of water in the system. At Brewery Creek, the intensity of rainfall is a more important predictor than the total quantity of rain.

Snowmelt-driven events We had less success modeling the loading produced by the snowmelt-driven events. The predictors that survived the screening process were different from stream to stream and those variables that did survive at most sites weren't always selected in the same order (like they were for the rainfall-driven events). What's more, the models for snowmelt-driven events were less accurate than for rainfall-driven events, ranging in R^2 from 0.24 to 0.53, with most in the 0.45 range.

At most sites, the most important predictor was a temperature measurement, either the maximum or the mean temperature during the loading event. The antecedent baseflow also appears to be important at most sites. It seems likely that, as in the case of rainfall-driven events, the loading is driven by the quantity of water that moves through the watershed during the event.

3   Conclusions

We have learned that we can predict the loading that will result from a storm with good accuracy, based just on the base flow before the storm and on the amount of rain that falls during the storm. Antecedent base flow is a measurement of how much water is in the watershed before a storm and any new water comes as rainfall, so it seems that the sediment and phosphorus loads are driven mainly by the quantity of water moving through the watershed. We have not yet found an accurate way to model the amount of load during a snowmelt-driven event but we have seen that the air temperature (which drives snowmelt), the antecedent base flow, and the amount of additional precipitation are important predictors for those events.

Most of the annual loading seems to be produced by a few major events. Characterizing these events will be an important step in describing the distribution of loadings and in informing management practices.

4   Next steps

There are at least two more creeks to include in the analysis. We also need to decide if there is an effective way to predict whether any given event will be one of the major events that produce most of the loading. Figures 4 and 3 make it look like the majority of the ranfall-driven loading comes from storms that drop at least two inches of rain. Mitigating the effect of large storms will probably require slowing the

water's movement through the watershed - for instance, by impounding runoff before it can flow into the creeks. Our analysis will look at the frequency of big storms in order to get an idea of how quickly impounded water must be dealt with in order to be ready for the next event.

References

[1] M.E. Danz, S.R. Corsi, D.J. Graczyk, and R.T. Bannerman. Characterization of suspended solids and total phosphorus loadings from small watersheds in wisconsin. Scientific Investigations Report 2010-5039, United States Geological Survey, 2010.
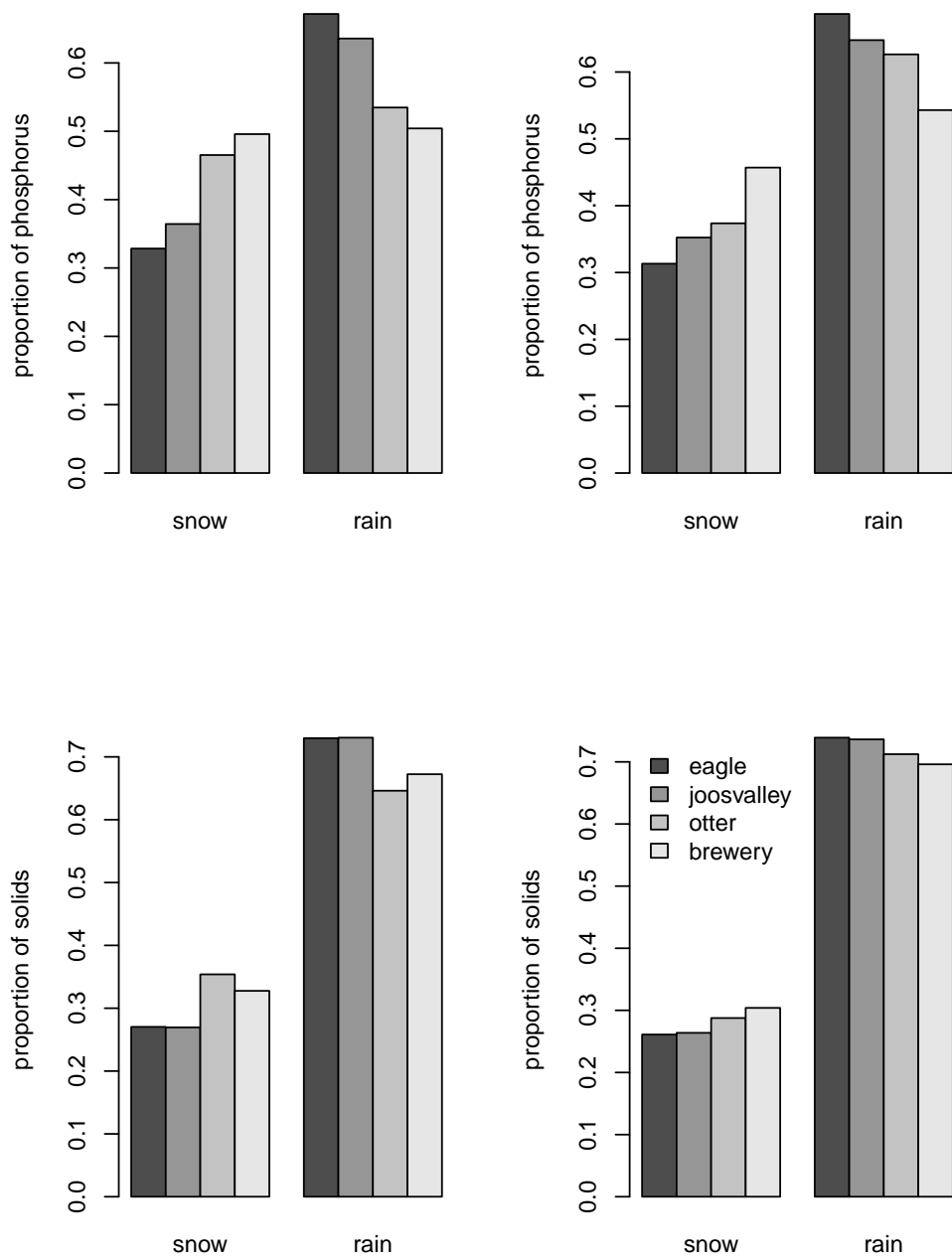
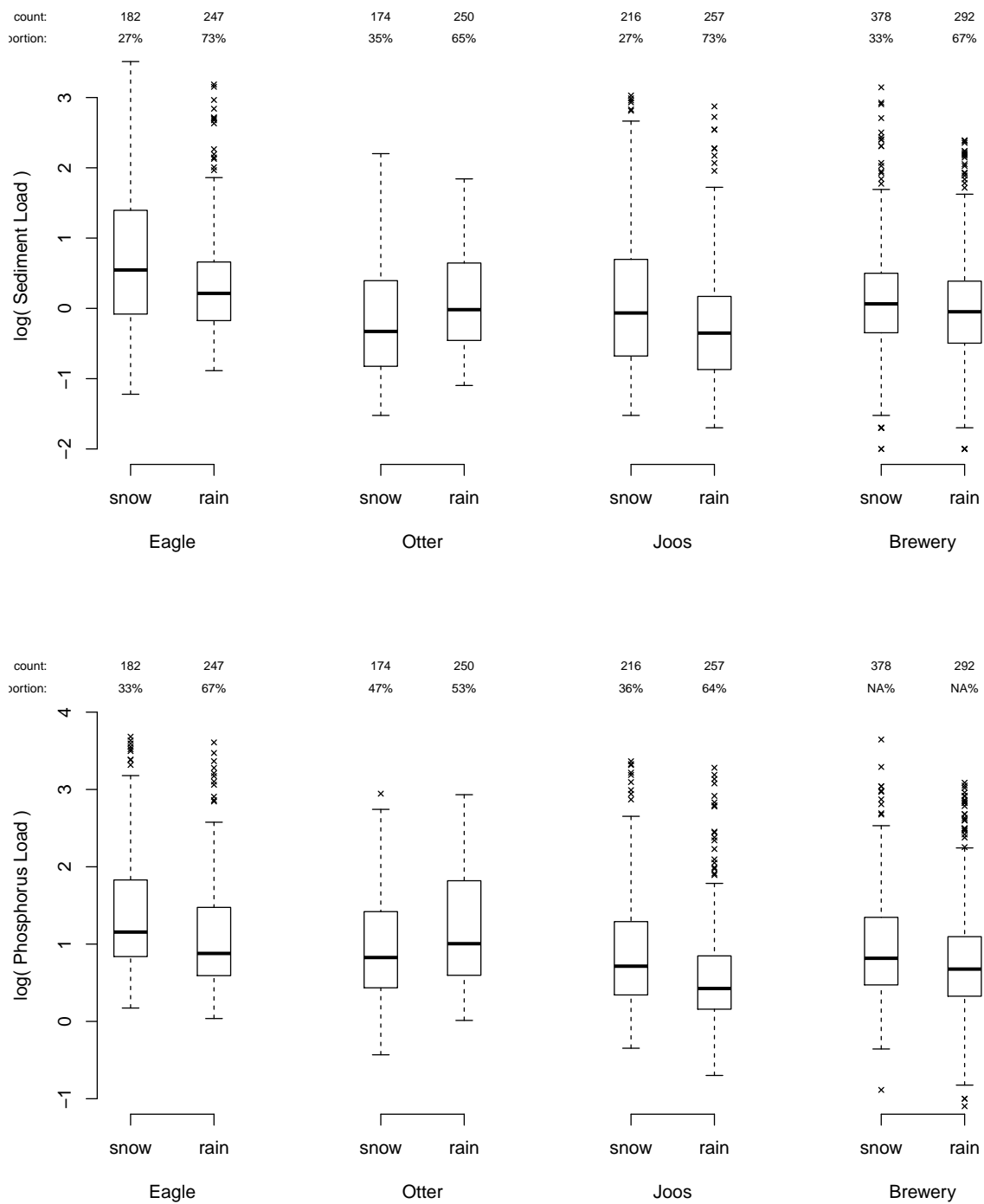Figure 1: Cumulative storm loadings at the four creeks.

Figure 2: Boxplots showing the sediment and phosphorus load produced by individual events at all four streams.

Sediment	R^2	Model terms
	0.503	theisen
	0.755	theisen + antecedent_qbase
Eagle	0.756	theisen + antecedent_qbase + tmean
	0.762	theisen + antecedent_qbase + tmean + tmax
	0.762	theisen + antecedent_qbase + tmean + tmax + p15max
	0.762	theisen + antecedent_qbase + tmean + tmax + p15max + p30max
	0.49	theisen
	0.665	theisen + antecedent_qbase
Joos	0.692	theisen + antecedent_qbase + p15max
	0.701	theisen + antecedent_qbase + p15max + cos_julian
	0.701	theisen + antecedent_qbase + p15max + cos_julian + antecedent_tmean
	0.701	theisen + antecedent_qbase + p15max + cos_julian + antecedent_tmean + ap_2day
	0.486	theisen
	0.738	theisen + antecedent_qbase
Otter	0.762	theisen + antecedent_qbase + sin_julian
	0.781	theisen + antecedent_qbase + sin_julian + antecedent_tmean
	0.781	theisen + antecedent_qbase + sin_julian + antecedent_tmean + ap_3day
	0.433	theisen
Brewery	0.459	theisen + p30max
	0.511	theisen + p30max + tmean

Phosphorus	R^2	Model terms
	0.579	theisen
	0.783	theisen + antecedent_qbase
Eagle	0.782	theisen + antecedent_qbase + tmean
	0.792	theisen + antecedent_qbase + tmean + tmax
	0.792	theisen + antecedent_qbase + tmean + p5max
	0.792	theisen + antecedent_qbase + tmean + tmax + p5max + p30max
	0.543	theisen
	0.715	theisen + antecedent_qbase
Joos	0.733	theisen + antecedent_qbase + p15max
	0.755	theisen + antecedent_qbase + p15max + ap_2day
	0.483	theisen
	0.737	theisen + antecedent_qbase
Otter	0.762	theisen + antecedent_qbase + tmin
	0.776	theisen + antecedent_qbase + tmin + sin_julian
	0.776	theisen + antecedent_qbase + tmin + sin_julian + ap_3day
	0.602	theisen
Brewery	0.636	theisen + ap_3day
	0.654	theisen + ap_3day + tmin + p60max

Table 3

Solids

Eagle:	theisen, antecedent_qbase, tmean, tmax, p15max, p30max
Joos:	theisen, antecedent_qbase, p15max, cos_julian, antecedent_tmean, ap_2day
Otter:	theisen, antecedent_qbase, sin_julian, antecedent_tmean, ap_3day
Brewery:	theisen, p30max, tmean

Phosphorus

Eagle:	theisen, antecedent_qbase, tmean, tmax, p5max, p30max
Joos:	theisen, antecedent_qbase, p15max, ap_2day
Otter:	theisen, antecedent_qbase, tmin, sin_julian, ap_3day
Brewery:	theisen, ap_3day, tmin, p60max

Table 4: The most important variables in the models for rainfall-driven loading. The variables are ordered by their importance to the model of the load.

Solids

Eagle:	nws_prec, tmax, nws_snow, antecedent_qbase, num_days, sin_julian
Joos:	num_days, tmax, nws_prec, antecedent_qbase, tmean, nws_snow, sin_julian
Otter:	tmean, tmin, nws_prec, antecedent_qbase, nws_snow, num_days, sin_julian
Brewery:	num_days, nws_prec, sin_julian, antecedent_qbase

Phosphorus

Eagle:	tmax, nws_prec, nws_snow, num_days, antecedent_qbase, sin_julian
Joos:	num_days, nws_prec, antecedent_qbase, sin_julian, nws_snow, tmax
Otter:	tmean, tmin, antecedent_qbase, nws_prec, nws_snow, num_days, sin_julian
Brewery:	num_days, nws_prec, sin_julian

Table 5: The most important variables in the models for snowmelt-driven loading. The variables are ordered by their importance to the model of the load.
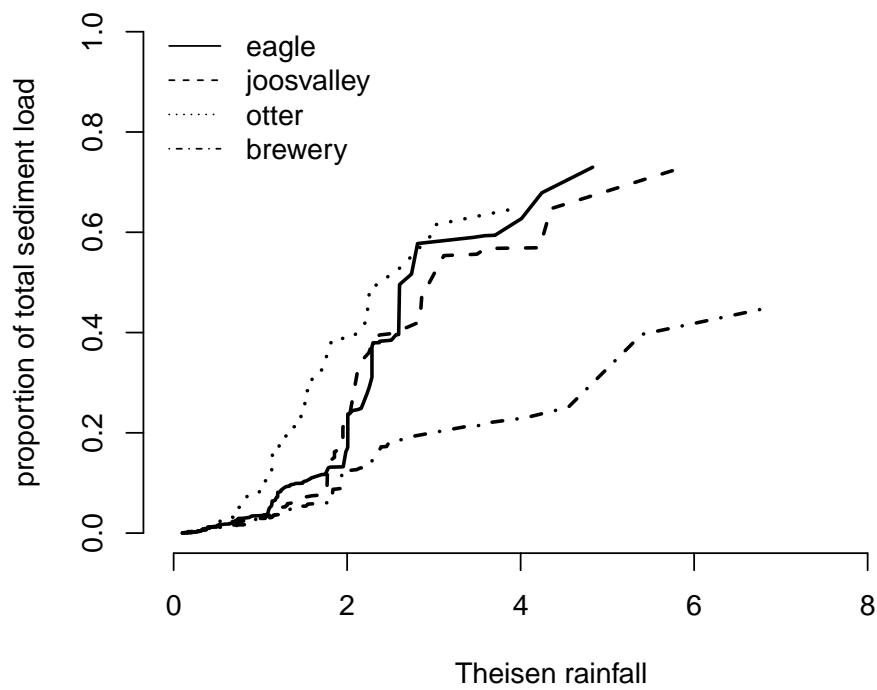
Figure 3: Proportion of the total sediment load contributed by rainfall events up to the size shown. Snowmelt-driven events are excluded.
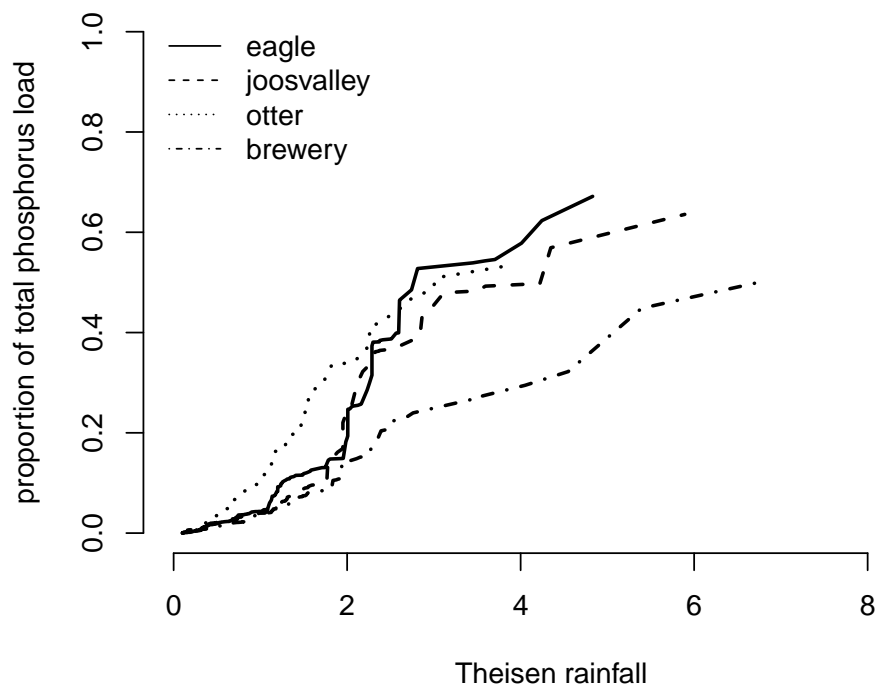
Figure 4: Proportion of the total phosphorus load contributed by rainfall events up to the size shown. Snowmelt-driven events are excluded.
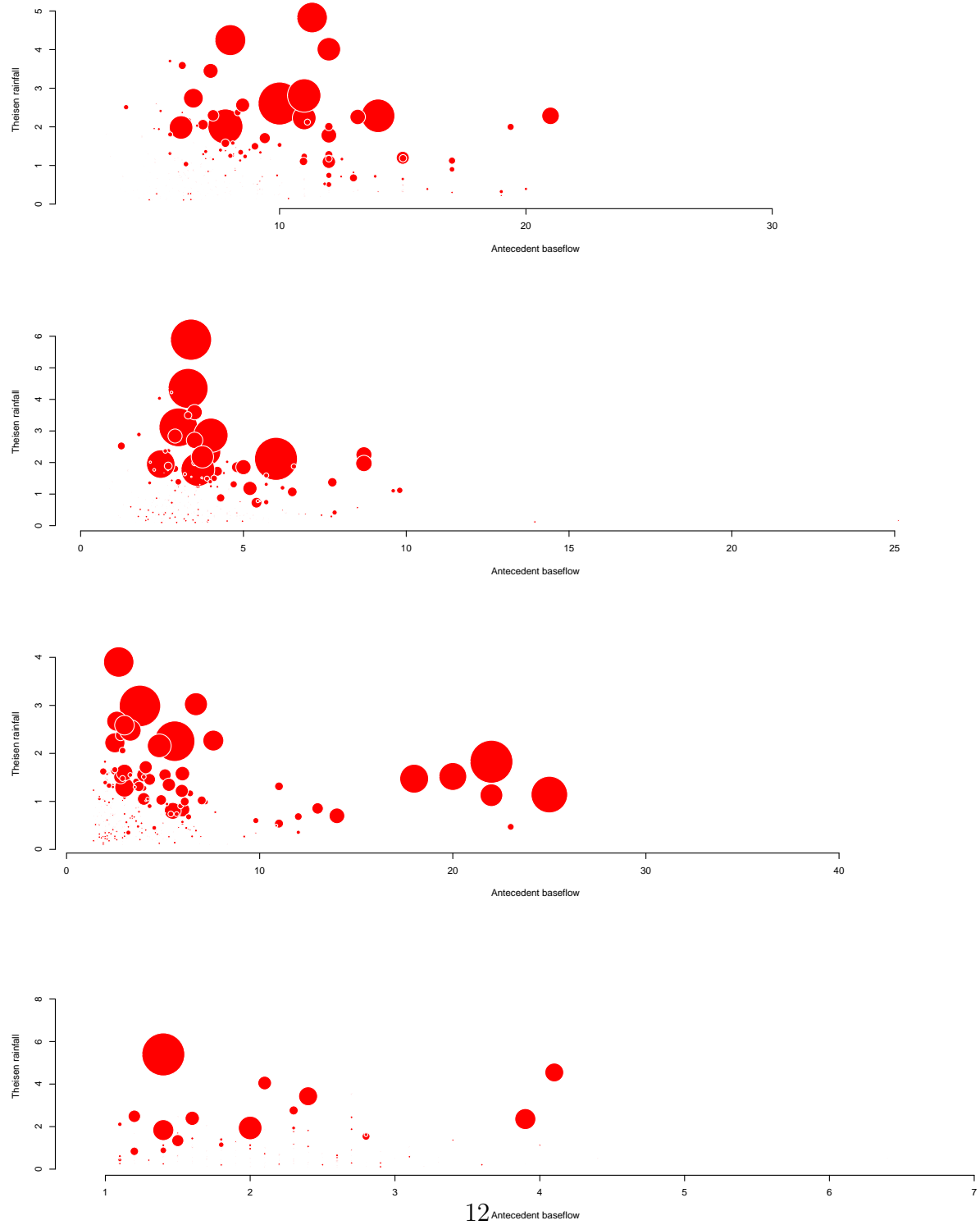
Figure 5: Antecedent base flow is the horizontal axis; theisen rainfall is the vertical axis. Each dot represents one event. The size of the dot shows the total sediment load contributed by that event.