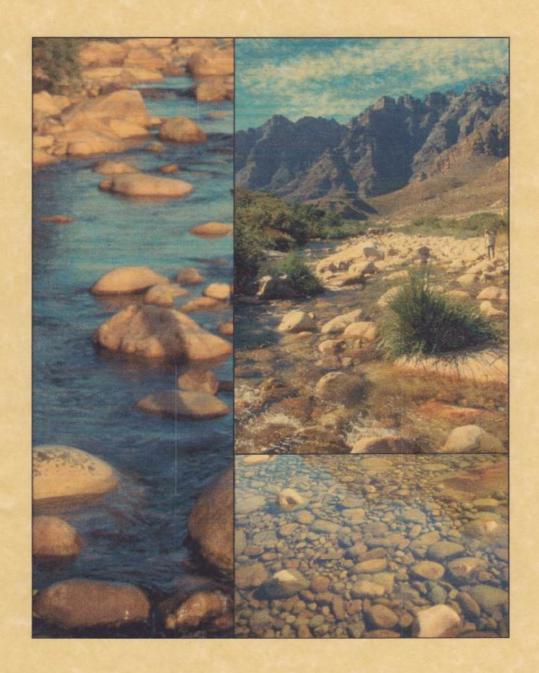
Microhabitat Suitability Curves derived for the Mayfly larvae (Ephemeroptera) of a south western Cape foothill river

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ABSTRACT: Benthic macroinvertebrate samples were taken from the Molenaars River of the south western Cape (33° 43' 23"; 19° 10' 32"), over three days during April and May 2000. Samples taken at depths ranging from 0.05 to 0.6 m and velocities ranging from 0 to 1.37 m.s⁻¹ yielded ten species of Ephemeroptera in five families. The abundance data were converted to densities and mean (±SD) density was estimated at 1669±1325 individuals.m⁻². Microhabitat suitability functions using single-factor and multiple polynomial regression for depth, velocity and Froude number were derived for the eight most common Ephemeropteran species using log-transformed (base 10) density data. Densities for all species exhibited significant correlation's (P<0.05) with velocity and Froude numbers where each variable was considered independently. Predicted optimal conditions for macroinvertebrates based on single-factor regression ranged from a velocity of 0.01 m.s-1 and Froude number 0.01 (Cloeodes sp. nov. 1), to 0.99 m.s-1 and Froude number 1.09 (Demoreptus capensis). The distribution by density of only one species (Lestagella sp.) was correlated with depth. The interaction terms of the multiple polynomial regressions for velocity and depth (vd) were significant (P<0.05) only for Baetis harrisoni. Weighted means for velocity and depth weighted by species density were used as additional descriptors of optimal conditions. Comparison of different descriptors of microhabitat suitability revealed that quadratic polynomial regressions fitted the data poorly and tended to over-estimate optimal conditions, while cubic polynomial regressions achieved a better fit and approximated the results obtained using weighted means, but tended to produce mathematical artifacts such as upturning

Introduction

General background

Changes to the ecological functioning of riverine ecosystems and the decline in biodiversity resulting from human modifications to natural flow regimes have focussed attention on the critical role played by local hydraulics in structuring the distribution, diversity and abundance of the biota. Of primary importance is current velocity, which affects the composition of the substrata, the delivery and cycling of nutrients and gases, the removal of metabolites and the distribution of food resources (Alan 1995; Davis and Barmuta 1989; Newson *et al.* 1998; Statzner *et al.* 1988; Statzner and Higler 1986). As a major physical force acting on the biota, creating conditions of stress and drag, current velocity also plays an important role in shaping the morphological and behavioural adaptations of aquatic organisms (Hynes 1984; Statzner *et*

al. 1988; Wiley and Kohler 1984). Alteration the flow regime and thereby local hydraulic conditions can therefore be expected to produce changes in species composition and standing crop of fish and aquatic macroinvertebrates.

A standard analytical technique for recommending environmental flows for rivers impacted by consumptive uses or development was described by Bovee (1986). The distribution of a species across a gradient of any physical habitat variable likely to change given an altered flow regime (i.e. substratum, velocity or depth) is represented on a frequency tally, bar graph or histogram. Based on the assumption that the most 'suitable' or 'optimal' conditions can be identified by the modal range or peak of the distribution, these 'habitat suitability criteria' (Bovee 1986) (in this paper referred to as 'microhabitat suitability curves') can then be used to predict how species composition or standing crop might change in response to an altered flow regime.

Microhabitat suitability curves are derived from a combination of 'utilisation' and 'availability' distributions. Utilisation describes the physical habitat conditions recorded where a particular species was observed or collected. To avoid bias introduced by environmental variability, utilisation is corrected using a distribution describing the availability of the physical habitat variable of interest. Availability is the full range of conditions available to the organism and is derived by randomly sampling the physical habitat conditions in the river at the same time as the biological data are collected (Bovee 1986). 'Suitability' or 'preference' curves for a particular species are then derived from the ratio between the utilisation and availability functions. This ratio is then as a value between 0 (least favourable) and 1 (optimal).

The replicability of the technique and the predictability of its output can be improved by fitting response functions to species distributions. A response function is a way of formally characterising the relationship between a dependent response (in this case the abundance or biomass of a species) and one, or several, explanatory habitat variable/s. The strength and direction of the relationship can thereby be statistically described and tested. Since the data often suggest a curvilinear relationship, polynomial regression (Kleinbaum and Kupper 1978;

Zar 1974) is most often used for this purpose. Using these methods, studies have revealed statistically significant relationships for the distribution by abundance or biomass of species over several physical habitat variables (Gore and Judy 1981; Jowett *et al.* 1991; Jowett and Richardson 1990; Orth and Maughan 1983).

Because microhabitat suitability curves provide a quantitative description of habitat, they are an effective means of translating the microhabitat requirements of species or communities into required flow conditions - and ultimately discharge - by contributing the biological input to hydraulic simulation techniques such as the Physical HABitat SIMulation model (PHABSIM) (Bovee and Milhous 1978). Flow conditions critical to the biota can then be identified, quantified and incorporated into a modified environmental flow regime.

Generally, substratum, depth and velocity have been used as standard explanatory variables. Statzner *et al.* (1988), however, have suggested that predictability and replicability are complicated by the complexity of flow behaviour in a natural river, and propose the use of hydraulic indices such Reynolds and Froude numbers. They suggest that these indices are more appropriate descriptors of flow in ecological studies than mean water column velocity. Furthermore, these indices can be calculated fairly simply from standard measurements. Froude number (Fr = v/\sqrt{gd}), the most commonly used index to describe the behaviour of flow, is a dimensionless number which describes the ratio of inertial to gravitational forces and gives an indication of the amount of turbulence (Davis and Barmuta 1989). By integrating the effects of velocity and depth, the Froude number may account for some of the variability evident in biological data which result from the interaction of velocity and depth.

While much of the earlier emphasis in developing microhabitat suitability curves was based on fish (Bovee & Cochnauer 1977; Bovee 1978;), benthic macroinvertebrates have increasingly become the focus of research (Gore 1978; Gore and Judy 1981; Jowett et al. 1991; Jowett and Richardson 1990; Orth and Maughan 1983) because of their widespread occurrence and fundamental importance to the functioning of riverine ecosystems (King and Tharme 1993). Being less mobile than fish, they are less capable of responding to altered flow regimes by avoiding unfavourable conditions and should, therefore, be expected to be more

sensitive than fish to changes in flow (Bovee, cited in King and Tharme 1993).

Overview of techniques for deriving habitat suitability curves for macroinvertebrates

A variety of techniques for deriving microhabitat suitability curves for macroinvertebrates has evolved from the basic principles described by Bovee (1986). Orth and Maughan (1983) fitted single factor polynomial regression models to log-transformed macroinvertebrate abundance's, biomass and community diversity indices. The response of ten macroinvertebrate species to depth, velocity, substratum and turbulence (Froude number) was investigated in this manner. A joint preference factor (JPF) derived from the combination of substratum, depth and velocity was found to be significantly correlated with biomass. Orth and Maughan (1983) found that the JPF is a reliable predictor of densities and could be used in place of multivariate preference functions to model the combined effect of the three variables.

In order to provide the input for the PHABSIM model, Gore and Judy (1981) further modified the above technique by relating cumulative mean abundance of the Trichopteran *Nectopsyche lahontonensis* to a predetermined increment of either velocity, depth or substrate. Fourth order polynomials were fitted to the data with the preference defined by the greatest incremental jump in the curve of the polynomial.

A major criticism of single factor habitat suitability functions so far described is that they assume that macroinvertebrates respond independently to velocity and depth (Mathur et~al. 1985). To overcome this assumption, Gore and Judy (1981) tested the interaction of velocity and depth by adding an interaction term (the product of velocity and depth, vd) to an exponential polynomial curve fitted to invertebrate densities. If the correlation coefficient for the interaction term (vd) was significantly different from zero, then the interdependence of velocity and depth could be demonstrated. All the tests were statistically significant, suggesting that the interaction term is an important element in the prediction of macroinvertebrate density.

The three approaches described above i.e. the single factor polynomial regression (Orth and Maughan 1983), the incremental approach (Gore and Judy 1981), and the multiple regression (Gore and Judy 1981), were tested by Morin *et al.* (1986), for precision and bias. Bias was

tested by comparing trends in the variability of the residuals. If residual variability differed significantly from zero, then preference curves were considered biased. Precision was tested by comparing the magnitudes of residual variability - high variability suggesting low precision. They found that the least biased and most precise method was the multiple regression approach proposed by Gore and Judy (1981) emphasizing the importance of factoring in the interaction of depth and velocity.

In several studies (Jowett and Richardson; Jowett *et al.* 1991 and Orth and Maughan 1983), weighted means have been used in conjunction with suitability functions to describe optimum physical habitat conditions. Gore (1978) calculated weighted means of velocity and depth weighted by the abundance data of 37 species. The weighted means for velocity and depth were plotted against each other and the optimal conditions for each species defined by a centroid. The optimal conditions for highest community diversity were also calculated and indicator species identified as those species whose centroids were located closest to the centroid for highest diversity. In future samples, the presence of the indicator species would suggest the maintenance of community structure.

Objectives of this study were to derive single factor and multivariate microhabitat suitability curves for the Ephemeroptera of the Molenaars River with respect to the following standard physical habitat variables: velocity, depth and Froude number - with the latter being used as a descriptor of turbulence. Values describing the optimal physical conditions for each species were calculated using weighted means weighted by the density of individual species at each increment of velocity, depth and Froude number in order to compare there performance against the predicted optimum conditions described by the suitability curves.

The Ephemeroptera was chosen as a target group because of their abundance, widespread occurrence, and taxonomic diversity in south western Cape foothill rivers.

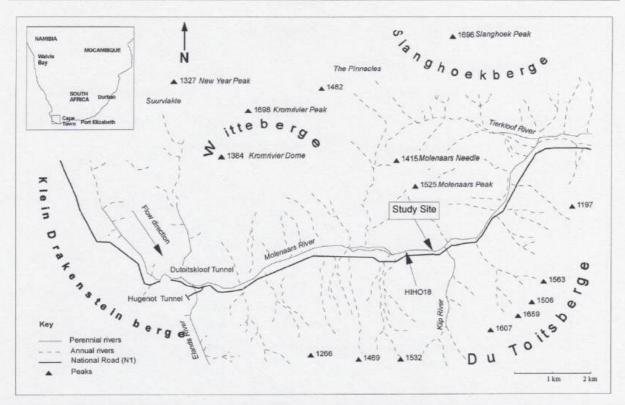


Fig. 1. The Molenaars River, south-western Cape, showing the study site (33° 43′ 23″; 19° 10′ 32″) located downstream of the gauging weir HIH018.

Study Area

The Molenaars River is a foothill cobble-bed river draining the Du Toits Kloof mountains, situated approximately 60 km east of Cape Town within the south-western Cape winter rainfall region of the fynbos biome. The Molenaars River, draining a catchment area of 113 km², is fairly typical of foothill rivers in the Cape Fold Belt of South Africa (Harrison 1965). The elevation of the catchment is approximately 400 m *amsl* at the study site, with surrounding mountain peaks rising to 1697 m (Slanghoek Peak) and 1838 m (Du Toit's Peak).

The study area (33° 43′ 23″; 19° 10′ 32″) indicated in Fig. 1, comprised a 100 m reach of the river, 22 km upstream of its confluence with the Breede River and was found to consist of a sequence of morphological units. Two riffles, upstream and downstream of a pool consisted of a mosaic of emergent boulders and boulder steps among cobbles. Flow types in these areas were predominantly Rippled Surface (RS) and Standing Waves (SW) interspersed by series of Cascades (CAS) over boulder steps and Chutes (CH) between emergent boulders. Flow-

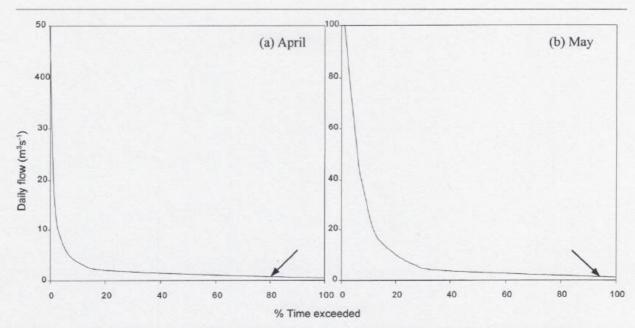


Fig. 2 Daily flow duration curves for gauging weir H1H018 for (a) April and (b) May calculated from flow data compiled from the years 1969 -2000. Arrows indicate the location on the flow duration curve of the discharge recorded for this study (0.6 m³ s⁻¹).

types in the pool were predominantly Smooth Boundary Turbulent (SBT) in the centre of the channel with Barely Perceptible Flow (BPF) along the channel margins (see Appendix A for a list of definitions for flowtypes).

Small and large cobble and boulders were the predominant substrata in both the pool and riffle areas. Depositions of sand and particulate organic matter were restricted to the banks and channel margins. The width of the river at the study site ranged from 10 to 20 m with an average gradient of 0.12 m.km⁻¹.

A Department of Water Affairs and Forestry gauging station, located approximately 300 m upstream of the study site provided the discharge data for this study. Low-flows occur during late summer - February and March - when monthly flows may vary between 1.7 - 2.7 10⁶ m³.s⁻¹, while high-flows occur during the winter months of June and July when monthly flows vary between 30 - 35 10⁶ m³.s⁻¹. Floods occur during the winter months of June through to September when maximum daily flow rates may exceed 80 m³ s⁻¹. The sampling, therefore, took place near the end of the low-flow season and beginning of the high-flow winter months. Discharge during the course of the study remained at 0.6 m³ s⁻¹ which is exceeded approximately 80% of the time in April and above 90% of the time in May (Fig. 2).

An open canopy of typical riparian species for the fynbos biome lined the left and right banks, which appeared to be fairly stable. Dominant species included: Wild almond (*Brabejum stellatifolium*) and Smalblaar (*Metrosideros angustifolium*). Palmiet (*Prionium serratum*), a rooted emergent reed, occurred along the banks and on channel islands. Black wattle (*Acacia mearnsii*), an exotic invasive that had infested the lower reaches of the river valley, had recently been cleared from this section of the river and was present but not dominant. The main human impacts on the study site were a trout farm and hotel located approximately 1 km upstream of the study site and the National Road (N1), which runs alongside the river approximately 50 m from the wetted area. Deposition of fine particulate organic matter (FPOM) from the outfall of the trout farm may result in the build up of detritus on the stream bed during low flows, although high flows during winter keep the bed well scoured (Simpson 1997).

Methods

Data collection

Sampling was carried out on three days between 28 April and 04 May 2000. Benthic macroinvertebrate samples were taken from 40 sampling points within the study site using a box sampler with a $0.34 \text{ m} \times 0.34 \text{ m}$ (0.1 m^2) frame size fitted with a collecting bottle of 250 μ m mesh diameter. This mesh size allowed the passage of finer organic material, while retaining all but the smallest larval instars of interest to this study. The sampler was positioned on the substratum and the larger particles removed, scrubbed of all fauna and set aside. The remaining particles were disturbed to a depth of 0.1 m. Samples were initially fixed in 4% formalin and taken back to the laboratory where they were transferred to 70% ethanol for preservation. The selected taxa were then separated from the detritus, identified to species level, where possible, and counted under a dissecting microscope.

Since logistical considerations limited the number of samples which could be taken to 40 (compared with 90 - 100 which are generally used to derive suitability curves; Jowett and Richardson 1990; Orth and Maughan 1983), sampling sites were chosen to ensure an even distribution of data points across all velocity and depth ranges. This was difficult in practice

and higher velocity ranges tended to be undersampled due to the higher proportion of slow moving water in the study reach.

Physical habitat variables (velocity and depth) were recorded at each sampling point. A Marsh-McBirney electromagnetic current meter on a metrically calibrated wading rod was used to measure mean current velocity. Since water velocity is assumed to vary logarithmically from zero at the stream bed to a maximum at the water surface, mean current velocity was measured at 0.6 of the water depth from the surface.

Abundance

The abundance (A) of each of the Ephemeropteran species (ten in number) collected at each of the 40 sampling points was determined and converted to densities by multiplication by 10 (D=10A) and reported in numbers.m⁻² (Appendix C, Tables 1 a - d and Table 2).

Development of Microhabitat Suitability Curves

The derivation of microhabitat suitability curves in this study differed in some respects from their original derivation as described by Bovee (1986) and summarized in Fig 3. The original derivation involved the computation of the suitability index or 'preference' as the ratio of utilisation to availability: ($P_i = U_i/A_i$) where, $P_i =$ an unnormalised index of preference, U_i the relative frequency of observations; A_i the relative frequency of x_i at the time of observation and x_i the microhabitat variable interval (Bovee 1986). Differences in the availability of habitat between streams can thereby be accounted for, allowing for the comparison of curves for the same species with data compiled from different systems.

Time and manpower constraints in this study did not allow for the collection of availability data. Since sampling sites were systematically chosen to reflect a wide range of environmental conditions in the river and no comparison between study sites was being attempted, it was therefore decided that the physical data from the biological samples would be an adequate reflection of conditions within the 100m reach. Thus the suitability curves in this study are more accurately described as 'utilisation curves'.

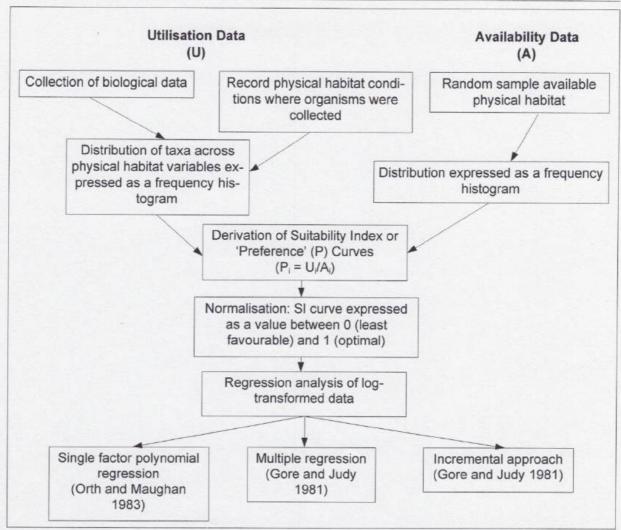


Fig. 3. Summary of the technique for arriving at microhabitat suitability curves as described by Bovee (1986), Gore and Judy (1981), and Orth and Maughan (1983).

A second difference is that the data were not expressed as a true 'index', i.e. they were not normalised. Suitability indices are arrived at by dividing the predicted abundance's by, A_{max} , the maximum predicted abundance: (SI = A/A_{max}, Bovee 1986). Thus, the suitability index is expressed as a ratio between 0 and 1. This procedure allows for a more pertinent method of comparing microhabitat suitability between species and sites, but since raw data cannot be displayed, it does not allow for examination of the degree of variability within the data, nor the comparison of relative abundance's between species nor for an evaluation of the goodness of fit.

Optimal habitat conditions in this study are therefore defined as those physical habitat conditions associated with the maximum predicted density (the peak of the curve) for each of the suitability functions, without normalisation.

Polynomial regression

(a) Polynomial regression (Kleinbaum and Kupper 1978; Zar 1974) was used to perform single factor analysis of the dependence of log-transformed (base 10) macroinvertebrate densities on the physical habitat variables of depth (d), velocity (v) and Froude number (Fr). Regressions were performed on the log-transformed data (Orth and Maughan 1983) to derive quadratic polynomial functions for each of the aforementioned variables using the following function, where D = density:

$$\log_{10} (D+1) = \alpha + \beta_1 x + \beta_2 x^2 \tag{2}$$

Habitat suitability was derived for those species where the regressions were significant (P<0.05).

(b) In order to test their performance against quadratic functions, cubic functions were then fitted to the data. Regression was performed on the log-transformed data (Orth and Maughan 1983) to derive cubic polynomial functions for each variable using the expression,

$$\log_{10} (D+1) = \alpha + \beta_1 x + \beta_2 x^2 + \beta_2 x^3$$
 (3)

(c) To account for the interaction of velocity and depth a multiplicative interaction term (vd) was added to the quadratic polynomial expression (Gore and Judy 1981) to obtain:

$$\log_{10} (D+1) = \alpha + \beta_1 v + \beta_2 d + \beta_3 v^2 + \beta_4 d^2 + \beta_5 v d$$
 (4)

If the interaction correlation coefficient β_s was not significantly different from zero (H_0 : β_s = 0), then the interaction of velocity and depth was assumed to be independent of macroinvertebrate densities.

All analyses were performed using the STATISTICA © computer software package.

Weighted means

Weighted means were used to describe the typical velocity and depth conditions for each species and a typical value for conditions of highest diversity (Gore 1978; Orth and Maughan 1983). Since the relative importance of each velocity and depth increment measured at each site differed according to the density of animals found there, each increment is assigned a proportionate degree of importance by multiplying it by the density found at that increment. By dividing the weighted physical habitat variables obtained by the sum of the weights, a weighted mean can be obtained. Weighted means for each physical habitat variable and diversity were therefore calculated as follows:

$$X_w = \sum_i W_i X_i / \sum W_i \tag{5}$$

where x_w = the weighted mean of a habitat variable x_i , the relative importance of which is determined by w_i , the weighting variable (i.e. species density or diversity) at site i.

Diversity

For each sample diversity was determined. The Shannon Weaver diversity index (H'), the most commonly used index of diversity, was used in this study. It accounts for both species richness and equitability, thus reflecting how evenly individuals are distributed over a range of species (Begon *et al.* 1990) and is calculated using the formula:

$$H' = -\sum_{i=1}^{s} P_i \log_e P_i \tag{1}$$

where P_i is the proportion of total individuals in the *i*th species and s is the total number of species at a site.

Hydraulic index

Froude number was calculated from:

$$Fr = U/\sqrt{gd}$$
 (6)

where U is the mean current velocity measured at 0.6 of the depth, g = acceleration due to gravity, and d = depth. Where the ratio is small (Fr<1), gravitational forces exceed inertial forces and the flow is described as subcritical or tranquil, where Fr = 1 the flow is critical or transitional and where Fr>1 inertial forces exceed gravitational forces and the flow becomes turbulent (supercritical) and is characterised by broken white water (Davis and Barmuta 1989).

See Appendix B for a glossary of all statistical and mathematical symbols used here.

Results

Physical habitat variables

A wide range of microhabitats was available to the benthos in each section of the riffle-pool sequence. Flow in pool areas was slower and deeper with a mean velocity in the pool (\pm SD) of 0.08 ± 0.04 m.s⁻¹. The average depth in the pool was 0.32 ± 0.09 m, although it exceeded 0.6 m in places. As expected, flow in riffles was shallower and faster with a mean velocity of 0.58 ± 0.29 m.s⁻¹ and depth of 0.16 ± 0.17 m. The maximum velocities recorded in the riffle areas was 1.37 m.s⁻¹. Froude numbers were between 0 to 0.05 (mean 0.03 ± 0.03) in pools and backwaters and 0.05 to 0.91 (mean 0.41 ± 0.25) in riffles. Froude numbers exceeding 1 were only found in shallow turbulent flows over boulder steps (maximum Fr was 1.71).

Invertebrate abundance

Ten species of Ephemeroptera from five families were found to occur at the study site. Rare

Table 1 Mean, standard deviation (SD) and coefficient of variation (CV) for counts of Ephemeroptera sampled from 40 points in the Molenaars river during the period 28 April - 04 May 2000.

	Numbers m ⁻²	±SD	CV
Baetis harrisoni	408	(±387)	- 1.0
Demoreptus capensis	384	(±441)	1.1
Afronurus sp	104	(±122)	1.1
Lestagella sp	183	(±171)	0.9
Aprionyx peterseni	153	(±241)	1.6
Cheleocloeon excisum	33	(±116)	3.4
Pseudopannota maculosum	10	(±22)	2.1
Castanophlebia sp.	2	(±5)	2.8
Cloeodes sp nov 1	417	(±1152)	2.7
Caenid sp	1	(±7)	6.3
Mean total taxa	1699	(±1325)	0.8

species, including *Castanophlebia* sp. and a species of Caenidae were found in less than five samples and were not included in the analysis. Caenidae were the rarest taxa and were found at only one sampling point. Mean (\pm SD) invertebrate density from site totals (Table 1) was found to be 1699 ± 1325 m⁻². *Cloeodes* sp nov 1 (undescribed) and *Baetis harrisoni* were the most abundant taxa. These accounted for 50 % of all Ephemeroptera, with mean (\pm SD) abundance's of 417 ± 1152 and 408 ± 387 , respectively. A high coefficient of variation (CV 2.7) for *Cloeodes* sp nov 1 suggested a highly clumped distribution, whereas *B. harrisoni* was more evenly distributed (CV 1.0).

Weighted means

Table 2 shows the weighted means obtained for depth, velocity and Froude number, weighted by the species density at each sample point. The results are plotted in Fig. 4. Organsims occurred at relatively shallow compared with the available range of depths sampled (0.65 m) and ranged from $x_d = 0.14$ m for *Caenid* sp. to $x_d = 0.31$ m for *Castanophlebia* sp. (since both species occurred in fewer than five samples, these estimates may reflect sampling bias and are therefore not included in Fig. 4). Most species had weighted means at depths between $x_d = 0.13$ and $x_d = 0.23$ m. Weighted means for velocity ranged from $x_v = 0.14$ m.s⁻¹ (*Cloeodes* sp nov 1) to $x_v = 0.82$ m.s⁻¹ (*Demoreptus capensis* - which had a correspondingly high Froude number: $x_f = 0.7$). The weighted mean for diversity was found at a depth of $x_d = 0.22$ m, a velocity of $x_v = 0.51$ m.s⁻¹ and Froude number of $x_f = 0.51$. The centroids for *B.harrisoni* and *Afronurus* sp. were located closest to the centroid for optimal diversity.

Table 2 Weighted means for physical habitat variables of depth (x_d) , velocity (x_v) and Froude number (x_f) weighted by species density.

	Weighted means (x _w)			
	Depth x_d (m)	Velocity x_v (m.s ⁻¹)	Froude x _f Number	
Baetis harrisoni	0.23	0.51	0.40	
Demoreptus capensis	0.18	0.82	0.70	
Afronurus sp.	0.22	0.40	0.31	
Lestagella sp.	0.20	0.45	0.36	
Aprionyx peterseni	0.21	0.69	0.54	
Cheleocloeon excisum	0.13	0.07	0.09	
Pseudopannota maculosum	0.16	0.61	0.53	
Castanophlebia sp.	0.31	0.47	0.34	
Cloeodes sp nov 1	0.23	0.14	0.13	
Caenidae sp.	0.14	0.00	0.05	

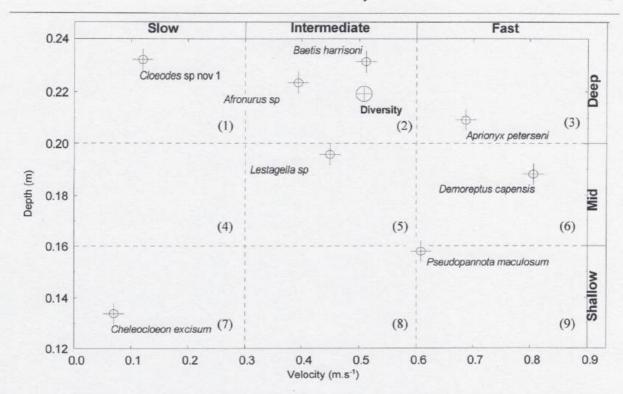


Fig. 4. Means of velocity and depth weighted by the density and diversity of each species at 40 sample points. Depth is categorised into 0.04 m increments: shallow 0.12 - 0.16 m; mid 0.16 - 0.20 m; deep 0.20 - 0.24 m. Velocity is categorised into 0.3 m.s⁻¹ increments: slow 0 - 0.30 m.s⁻¹; intermediate 0.3 - 0.6 m.s⁻¹; fast 0.6 - 0.9 m.s⁻¹. Permutations of depth and velocity categories are numbered (1) - (9).

The velocity and depth ranges for the species sampled were each arbitrarily divided into three categories: shallow-, mid-depth and deep; and slow-, intermediate- and fast-velocities. These categories are based on the minimum and maximum weighted means obtained for the invertebrates and not on the complete range of velocities and depths sampled. The centroids for each species are located in seven of the nine permutations of the velocity and depth categories: (1) slow-velocity, deep species (*Cloeodes* sp nov 1); (2) intermediate-velocity, deep species (*B harrisoni*, *Afronurus* sp. and *Lestagella* sp., including the centroid for diversity), slow-velocity, shallow-depth species (*C. excisum*); (3) fast-velocity, deep species (*Aprionyx peterseni*); (5) intermediate-velocity, mid-depth species (*Lestagella* sp.); (6) fast-velocity, mid-depth species (*Demoreptus capensis*) and (9) shallow-depth fast-velocity species (*P. maculosum*).

Microhabitat suitability curves

Microhabitat suitability curves derived for velocity and Froude number are presented in Fig.'s 5 and 6 (a) - (h) respectively and describe the distribution over these variables for eight of the

ten Ephemeropteran taxa collected. Correlation coefficients and maximum predicted values for each species are reported in tables 3 and 4 respectively.

Velocity

Polynomial regressions fitted to invertebrate abundances. Use of velocity (v) in quadratic and cubic response functions proved significant (p<0.05) for all species except P. maculosum (Table 3). The curves suggest that highest invertebrate densities are distributed across a range of velocities, increasing up to a maximum and decreasing thereafter. This range appears to be relatively broad compared with the available range of velocities (B. harrisoni, D. capensis, Afronurus sp and Lestagella sp Fig 5 a - d), except for those species occurring in the slower velocity ranges (C. excisum and Cloeodes sp nov 1, Fig 5 f and h) where most organisms were restricted to velocities less than 0.4 m.s^{-1} . Aprionyx peterseni (Fig. 1e) appears to be distributed across almost the entire recorded velocity range.

Correlation coefficients (R) were highest for B. harrisoni (0.63 for the quadratic and 0.71 for the cubic function) and D. capensis (0.79 for both the quadratic and cubic functions) and lowest for P. maculosum (0.41 and 0.42 for the quadratic and cubic functions respectively).

Suitability functions for most species are convex, either skewed towards the mid to lower velocity ranges (0.45 - 0.65 m.s⁻¹: *B. harrisoni, Afronurus* sp., *Lestagella* sp.), or in the single case of *D. capensis*, skewed toward the higher velocities (0.97 - 0.99 m.s⁻¹). *C. excisum* and *Cloeodes* sp. nov. 1 have curves which are concave, showing highest abundances in the lower velocity ranges (<0.01 m.s⁻¹).

Table 3 Correlation coefficients (R) for quadratic and cubic polynomical regression functions (Fig.s 5 and 6, a - h) derived from Ephemeropteran densities for velocity (v), depth (d) Froude number (Fr) and the interaction term for velocity and depth (vd). * P < 0.05; ** P < 0.01; *** P < 0.001.

	Correlation coefficients (R)					
	$\overline{v^2}$	v^3	d ²	Fr ²	Fr ³	vd
Baetis harrisoni	0.63***	0.71***	0.28	0.64***	0.68***	0.69***
Demoreptus capensis	0.79***	0.79***	0.36	0.80***	0.82***	0.83***
Afronurus sp.	0.46*	0.61***	0.23	0.19	0.35	0.48
Lestagella sp.	0.49**	0.60***	0.39*	0.30	0.52**	0.60**
Aprionyx peterseni	0.43*	0.51*	0.34	0.47**	0.49*	0.55*
Cheleocloeon exisum	0.44*	0.47*	0.33	0.38	0.44	0.54*
Psuedopannota maculosum	0.41*	0.42	0.26	0.40*	0.40*	0.45
Cloeodes sp. nov. 1	0.45*	0.45*	0.06	0.51**	0.51**	0.54*

Depth

None of the polynomial regression functions fitted to depth alone proved significant (p>0.05 in all analyses), suggesting the independence of depth and macroinvertebrate densities (Table 3). These functions were therefore not produced in this study.

Froude Number

The shape of the curves tended to be similar to the curves for velocity, although tolerance ranges for Froude number appear to be narrower for most species (see Fig 5a *B. harrisoni* and Fig 6c *Afronurus* sp.).

The highest densities of most species are limited to hydraulic conditions where Froude number is <1 (subcritical conditions). Most species occur within the range of Fr 0.50 - 0.60, with the exception of *D. capensis* which has a maximum predicted Fr of 0.86 - 1.09, suggesting a preference for turbulent conditions in shallow and fast water (supercritical conditions).

Interaction of Velocity and Depth

The correlation coefficients for multivariate functions were significant (p<0.05) for all species except *Afronurus* sp. and *P. maculosum*. However, *B. harrisoni* was the only species for which the interaction partial regression coefficient (vd) was significant (P<0.05). Response surfaces were derived for *B. harrisoni* and *D. capensis* (Fig. 7) as examples of contrasting microhabitat distributions and not for the remaining species.

Table 4 Values obtained by inspection for velocity (ν, m.s⁻¹) and Froude number (Fr) at which maxima of Ephemeropteran densities for quadratic and cubic polynomial regressions are predicted.

	Quadratic		Cubic	
	Velocity (m.s ⁻¹)	Froude No.	Velocity (m.s-1)	Froude No.
Baetis harrisoni	0.64	0.64	0.50	0.49
Demoreptus capensis	0.99	1.09	0.86	0.97
Afronurus sp.	0.59	0	0.31	0.42
Lestagella sp.	0.63	0.44	0.42	0.45
Aprionyx peterseni	0.90	0.86	0.65	0.49
Cheleocloeon exisum	< 0.01	< 0.01	< 0.01	< 0.01
Psuedopannota maculosum	0.70	0.64	0.31	0.49
Cloeodes sp nov 1	< 0.01	0.01	< 0.01	< 0.01

From Fig. 7 it can be seen that highest abundances of *B. harrisoni* occur at velocities between 0.42 and 0.71 m.s⁻¹ and depths of 0.21 and 0.44 m.s⁻¹. Highest densities of this species are predicted to occur at mid-depths and intermediate velocities (in relation to the maximum depth and velocities sampled). Highest densities of *D. capensis* (Fig 7b) were found between velocities of 0.59 and 1.6 m.s⁻¹ and depths < 0.11 m, suggesting that shallower depths and higher velocities are most suitable for this species

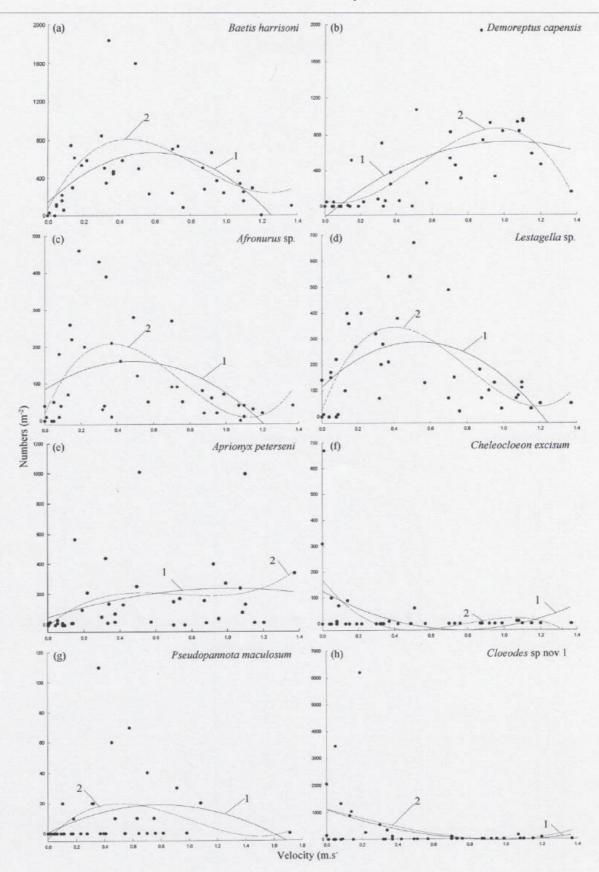


Fig. 5. Comparison of microhabitat suitability curves for Ephemeroptera from the Molenaars River fitted for velocity by (1) quadratic and (2) cubic polynomial regressions. Correlation coefficients and significance levels are reported in Table 3

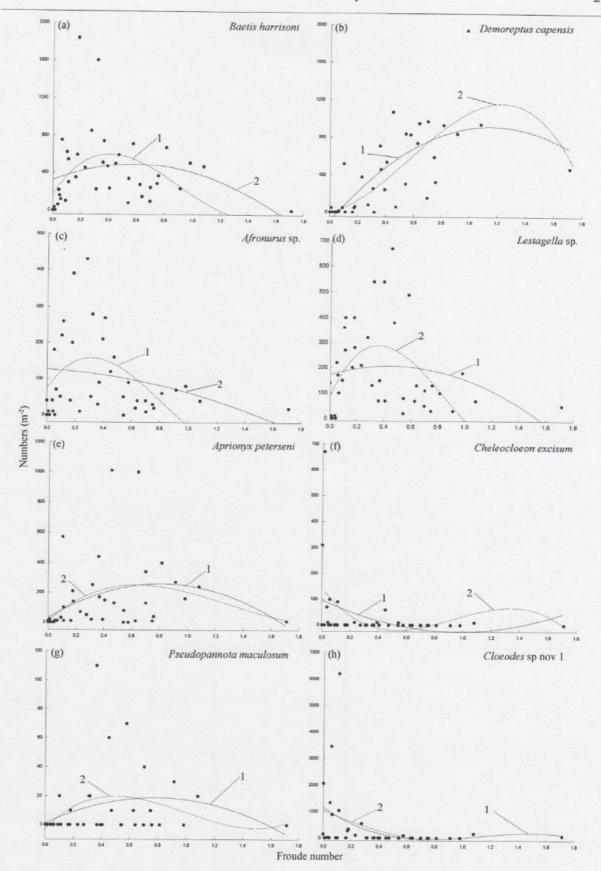


Fig. 6. Comparison of microhabitat suitability curves for Ephemeroptera from the Molenaars River fitted for Froude number by (1) quadratic and (2) cubic polynomial regressions. Correlation coefficients and significance levels are reported in Table 3

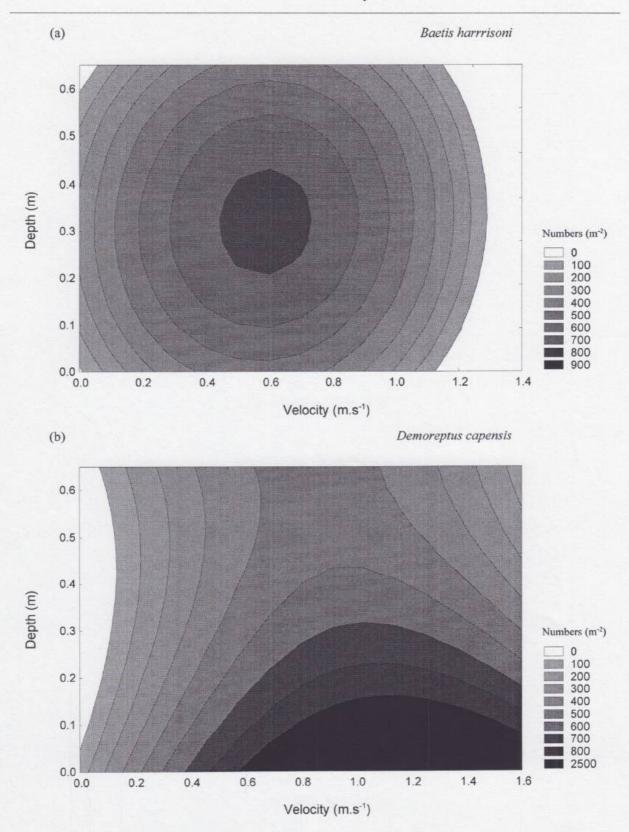


Fig. 7. Comparison of response surfaces derived from multivariate polynomial regressions including the interactive term (vd) for (a) B. harrisoni (1.633+4.244v-0.884d-4.091v²-0.812d²+4.163vd) and (b) D. capensis (1.379+4.896v-7.492d-2.920v²+7.330d²+3.189vd)

Discussion

The relationship between mean water column velocity and the microdistribution of aquatic invertebrates has been questioned by some authors (Statzner 1981; Macan 1974), primarily because stream organisms (with the exception of Simuliidae and Blephariceridae) are believed to occupy the viscous sublayer on the lower surface of stones (Statzner 1981). This suggests that the organisms are able to find sheltered refuges from the current, taking advantage of small-scale flows created by the substratum irrespective of mean water column velocity. Nowell and Jumars (cited in Allan 1995), however, have proposed that the viscous sublayer is an artifact of experimental conditions where flow is laminar. Under conditions of increasing turbulence (the predominating condition in natural streams), this viscous sublayer becomes thinned to the point where it may be of no consequence to the invertebrates at all, particularly when flows exceed 0.2 m.s⁻¹ (Allan 1995). However, while the flow microenvironment within the hyporheos does afford some refuge from the prevailing flow conditions, direct exposure of benthic invertebrates to mean current velocities can be assumed to take place for at least part of the time: Glozier and Culp (1989), for example, demonstrated significant diel movement by two mayfly species from the lower surfaces of the substratum during the day to the upper surfaces at night, and large numbers of at least one species of Ephemeroptera were observed grazing the periphyton on upper surfaces of rocks in the pool areas during the course of this study (pers.obs.).

One of the assumptions of the present study is, therefore, that while the measurement of mean water column velocity may not give a true reflection of the flow conditions experienced by the organism, it is in at least a limited sense, representative of these conditions. The statistically significant relationships between mean water column velocity and the distribution of the benthos obtained in this study appear to support this assumption. However, the discrepancies between measured water column velocity and flow near the bed may explain the relatively broad distributions (Fig., 5, a-h and 6, a-h) observed for invertebrate densities across physical habitat variables and the relatively low correlation coefficients in some cases. Actual tolerance ranges for individual species may therefore be far narrower and well below the velocities reported here, especially for samples taken from deeper water where mean

water column velocity is less representative of conditions near the bottom.

The weighted means for velocity and depth reported here for highest diversity are slightly lower than (0.51 m.s⁻¹ and 0.22 m respectively, Table 4), but compare favourably with, the findings of Gore (1978; 0.76 m.s⁻¹ velocity 0.28 m) and Orth and Maughan (1983; 0.6 m.s⁻¹ and 0.34 m). The discrepancies in these findings are probably due to the comparison of different species, but confirm the critical importance of the shallower depths (< 0.3 m) and mid-velocity ranges (0.5 - 0.7 m.s⁻¹) to most stream invertebrate species.

The performance of different descriptors of microhabitat suitability as reflected in maximum predicted density for a specific physical habitat variable are summarized in Fig. 8 a - f. Quadratic polynomial regressions tend to overestimate velocity and Froude number optima in relation to cubic polynomial regressions (Fig. 8 a and d) and weighted means (Fig. 8 b and e) and appear to fit the data poorly (Fig.s. 5 and 6 a - h). They also display very high y-intercepts for certain species predicting higher than expected densities for these species in the lower velocity ranges. This is particularly evident for the quadratic functions fitted to the distributions for Froude number of *B. harrisoni* (Fig. 6a), *Afronurus* sp. (Fig. 6c) and *Lestagella* sp. (Fig 6d) where the tolerance ranges are narrower. Slauson (1988), suggests correcting this by forcing the regression through the origin which can be achieved by omitting the intercept term (α). However, Slauson (1988) points out that this tends to give an inflated r-squared since the intercept was not estimated. Also, omission of the intercept term also does not represent skewed data effectively.

The optima for cubic polynomial regressions approximate those values obtained from the weighted means of velocity and Froude number (Fig. 8 c and f), but are in some instances slightly higher. While achieving a better fit, cubic polynomials tend to display mathematical artifacts such as upturning. However, unless there is reason to suspect a bimodal distribution of the species such as a differential distribution of size class, upturning (or 'curve tails') can be safely ignored. Decisions to omit curve tails should be based on a biological understanding of the species and an examination of the raw data (King and Tharme 1993).

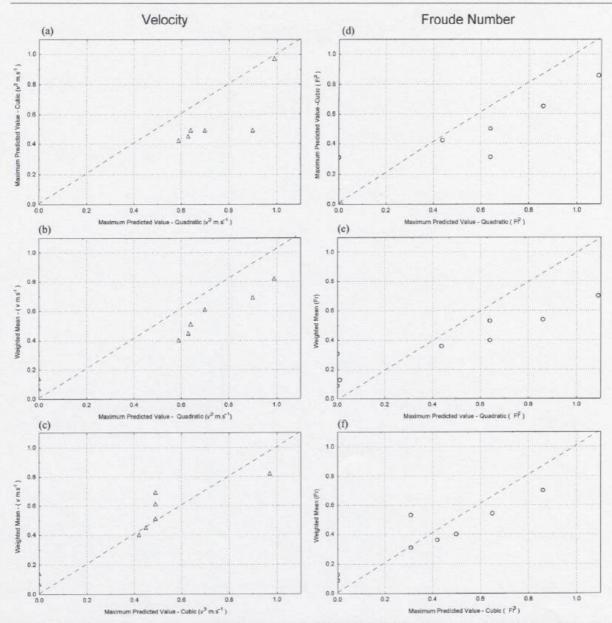


Fig. 8. Comparison of values for velocity (a - c) and Froude number (d - f) at which expected values of habitat suitability appear to be maximum for: weighted means, quadratic polynomial regression and cubic polynomial regression. The physical habitat value (velocity or Froude number) for which a particular model predicts maximum is plotted on the x-axis with the corresponding value for the model being compared on the y-axis. Where both predictions agree, the point falls on the broken line.

The disadvantages of using a single-figure descriptors (i.e. weighted means) of microhabitat suitability are clear if the weighted means of velocity for *A. peterseni* (Fig. 4) are compared with the suitability curve for this species (Fig 5e). Examination of optimal velocity conditions for *A. peterseni* described by the weighted mean (Fig. 4) can be found at 0.69 m.s⁻¹, whereas the suitability curve for this species (Fig. 5e) and correlation coefficient, as well as the distribution of the raw data, suggests that the relationship between velocity and the density of

A. peterseni while being significant (P<0.05) is relatively weak (R 0.51, Table 3), showing little or no trend toward any particular velocity value. Because they do not reflect the distribution of the species over the whole range of a particular variable, weighted means give no indication of strength of the relationship.

Interaction of velocity and depth

Microhabitat suitability curves have been criticised (Mathur et al. 1985) for not factoring in the interaction of depth and velocity. Single factor polynomial regressions were favoured above multivariate techniques because of early limitations in computer hardware and software (Hanson 1988). However, readily available software packages for analysing and visualising two dimensional data have simplified matters considerably.

The suitability of a certain depth for a particular species may depend on the velocity at that depth. This is due to the reduced thickness of the viscous sublayer with increasing turbulence (Allan 1995) and the subsequent exposure of organisms to higher velocities under conditions of higher turbulence. Since turbulence increases with increasing height of the roughness elements in relation to the speed and depth of the water (Davis and Barmuta 1989), the interaction of depth and velocity is an important factor in determining invertebrate distributions.

Bovee (1986) recommended restricting bivariate analyses to quadratic polynomial regressions since cubic polynomial regressions and higher produced bimodal response surfaces. However, in this study, the bivariate quadratic functions tended to demonstrate similar weaknesses to single factor quadratic functions, i.e., suggesting high densities at low physical habitat values (in this case depth, see *D. capensis* Fig. 7 b).

The only significant interaction term (vd) for multivariate quadratic polynomial regressions was obtained for *B. harrisoni* (P < 0.05). An examination of Fig. 7 (a) reveals that the optimum velocity range described by single factor polynomial regression and weighted means (0.64 and 0.51 m.s⁻¹ respectively) are only valid for depths between 0.25 and 0.45 m. Although the interaction term for *D. capensis* was not statistically significant (P > 0.05), it

Since the dataset was small in this particular study an overreliance on statistical significance was avoided (Day and Quinn 1989).

was decided to include the results of the multiple regression analysis (Fig. 7 b and Appendix B) for reasons of biological significance¹. Fig 7 (b) suggests that highest abundances of D. capensis can be found between 0.6 and >1.6 m.s⁻¹ only where depths are < 0.15 m. This is confirmed by the location of highest abundances of D. capensis in the higher Froude number ranges (>1 Table 4). The usefulness of including the interaction term in regressions and the danger in interpreting microhabitat suitability on the basis single-factor regression is therefore evident. However, the absence of significance for most interactive terms suggest that the interaction of velocity and depth and its effect on invertebrate distributions may be better integrated by hydraulic indices such as Froude number for which all correlation's were significant (Table 3).

Limitations and recommendations

Although weighted means supported the results obtained using cubic polynomial regression, the use of weighted means to describe habitat suitability is not recommended since they merely describe 'average' conditions and give no reflection of the variability within the data or the strength and trend of the relationship. Similarly, the use of an indicator species is questionable since an indicator species merely represents 'average' or 'typical' conditions for diversity. The results from this study suggest that there is high variability in the selection of microhabitat conditions even within families and the presence of a single 'indicator' species may not adequately reflect the conditions required for the maintenance of community structure. Furthermore, the effect of losing species from a community cannot be predicted unless the species which is lost and its function within that community is understood (Harris 1999). The analysis of a representative group of species whose microhabitat requirements are known may be more effective than the use of a single indicator species.

A major criticism of microhabitat suitability curves (Mathur *et al.*1985) are the weak correlation's between observed and predicted abundances. These weak correlation's may be partly related to the inherent imprecision and bias of the models themselves (Morin *et al.* 1986), but may also be related to the patchiness of lotic communities, a feature evident from the high coefficients of variation for density of several species in this study (Table 1). Patchy distributions are a common feature of freshwater communities (Hildrew and Giller 1994) due

to variability in a multitude of environmental factors. It is therefore difficult to investigate any single factor in isolation. The influence of other factors not considered here such as food preference (Macan 1974), competition (McAuliffe 1984; Hemphill 1988), predation (Peckarsky, Horn and Statzner 1990), recent disturbance events and colonisation rates, should therefore not be ruled out when considering invertebrate distributions.

A further complication in the development of habitat suitability curves and one which may lead to a lack of significance and low correlation's is the coarseness of standard sampling procedures for benthic macroinvertebrates (Davis and Barmuta 1989). The use of a box sampler mixes spatial scales and makes the characterisation of substratum difficult. Individual rocks may be more appropriate than box samplers as sampling units for the derivation of microhabitat suitability curves.

Ideally availability data should be determined independently of use data using the random sampling techniques described by Bovee (1986). Microhabitat suitability curves need to be developed for each river and site to account for differences in species composition, temperature and water quality regimes. The development of microhabitat suitability curves should be done in conjunction with habitat mapping (Gore and Judy 1981), or comparison with availability data, to be most effective in determining environmental flows. However, more studies need to be directed at investigating the relationship between 'use' and 'availability'. This relationship remains unclear (Mathur *et al.* 1985) and may be critical to the prediction of the effects of altered flow regimes on aquatic communities.

Patterns of microhabitat use and relative abundance of macroinvertebrates can also be expected to vary from year to year and between seasons as well as rivers (Orth 1987). While it is impossible to control for all variables, suitability curves should be compiled from data which are accumulated over several seasons in order to take this into account. Orth (1987) has also suggested that critical flows, (i.e. flows when microhabitat is most limited) need to be identified to avoid underestimating the amount of habitat available during such times.

Suitability curves should not be relied upon to account for all aspects of instream environ-

mental flows. For example, suitability curves do not account for flows which may be sufficient to maintain high levels of diversity, but may not be sufficient to provide flushing flows to clear accumulated sediment from the hyporheos, thereby maintaining important sites for colonisation and refugia during spates (Simpson 1997).

Conclusions

The results presented in this study, indicate that there may be a set of flow conditions which are most likely to maintain the present composition of the Ephemeropteran community in the river during the months in which the river was sampled i.e. the maintenance of relatively shallow riffle areas (<0.3 m) with velocities exceeding 0.3 m.s⁻¹. Should flow conditions become reduced below these values it is likely that species such as *D. capensis* with higher velocity tolerances will become reduced in numbers and community composition will shift in favour of the slow water – shallow species such as *C. excisum* and *Cloeodes* sp. nov. 1. However, microhabitat suitability curves for the same species need to be derived for other rivers in the south- western Cape before the validity of the present functions can be confirmed.

Despite the assumed sensitivity of the curves to the site and time of sampling, as well as to the type of model or descriptor used, they represent a useful way of providing biologically meaningful data for environmental flow assessments. In order to confirm their usefulness, studies will need to be conducted which investigate the response of the biota after the implementation of specified flow regimes. Also, microhabitat suitability curves need to be linked to models which predict how available habitat varies with discharge. Current methodologies which are able to do this, such as the Instream Flow Incremental Methodology (IFIM, Bovee 1986), link microhabitat suitability curves to simulated changes in the availability of physical habitat in cells in a cross-sectional profile of the channel at incremental changes in flow. However, these models were developed in the United States and are inappropriate for developing countries where time and manpower constraints are limiting. Until new, more appropriate models have been developed and tested, a conservative estimation of habitat requirements using the techniques prescribed here will need to suffice for informing management decisions and conserving species diversity in local rivers.

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APPENDIX B
GLOSSARY OF SYMBOLS

Table 1	Glossary of mathematical and statistical study are defined here.	symbols. Each of the symbols used in this
A _{max}	maximum predicted abundance/ density	
A_i	the relative frequency of sites with the value \boldsymbol{x}_i	
α	y intercept in regression equation	
β	partial regression coefficient of regression equation	
d	depth	
Fr	Froude number	Fr < 1 subcritical flow
		Fr = 1 critical flow
		Fr >1 supercritical flow
g	acceleration due to gravity (9.8 m·s ⁻²)	
P_i	unnormalised index of preference	
Ui	relative frequency of x _i	
v	velocity	
Ú	mean water column velocity at 0.6 depth from the surface	
W_i	weighting variable at site i	
X_d	weighted mean depth	
X_f	weighted mean Froude number	
X_{ν}	weighted mean velocity	
X _w	weighted mean of a habitat variable (x)	

APPENDIX A DESCRIPTIONS OF FLOWTYPES

Table 1 Categories of flow types identified in the reach (Rowntree 1996)

Flow Type	Definition
Rippled Surface (RS)	The water surface has regular smooth distur- bances which form low transverse ripples across the direction of flow
Standing Waves (SW)	Standing waves which may or may not be broken at the crest
Cascades (CAS)	Water tumbling down a stepped series of boulders, large cobble or bedrock
Chutes (CH)	Water forced between two rocks, usually large cobble or boulders; flowing fast with the fall too low to be considered free falling.
Smooth Boundary Turbulent (SBT)	The water surface remains smooth, medium to slow streaming flow takes place through- out the water profile, turbulence can be seen as the upward movement of fine suspended particles.
Barely Perceptible Flow (BPF)	Smooth surface flow, only perceptible through the movement of floating objects

APPENDIX C
Ephemeropteran taxa densities and physical habitat variables

Table 1 (a) Ephemeropteran taxa densities (numbers.m⁻²), velocity (m.s⁻¹), depth (m) and Froude number for sample points 1 - 10.

Family	Sample No.	BRBSMo1	BRBSM ₀ 2	BRBSM ₀ 3	BRBSM04	BRBSM ₀ 5	BRBSMo6	BRBSMo7	BRBSM08	BRBSM09	BRBSMo10
	Velocity (v)	1.08	0.32	0.99	0.19	0.04	0.13	0	1.07	1.37	0.37
	Depth (d)	0.4	90.0	0.12	0.3	0.5	0.41	0.14	0.1	0.39	60.0
	Fr (v/√gd)	0.55	0.36	0.91	0.11	0.02	90.0	0.00	1.08	0.70	0.39
Baetidae	Baetis harrisoni	340	510	240	940	0	750	0	470	100	470
	Cloeodes sp nov 1	0	0	0	6200	0	880	140	150	0	0
	Pseudopannota maculosum	0	110	30	0	0	0	0	20	0	0
	Demoreptus capensis	840	710	840	0	0	10	0	940	160	250
	Cheleocloeon exisum	10	0	0	0	0	0	310	10	0	10
Heptageniidae	Afronurus sp.	0	30	70	460	0	70	0	40	40	210
Ephemerellidae	Lestagella sp.	80	70	30	270	0	100	0	70	20	940
Leptophlebiidae	Aprionyx peterseni	80	440	270	100	0	10	0	240	340	20
	Castanophlebia sp.	0	0	0	30	0	0	0	0	10	0
Caenidae	Caenidae sp	0	0	0	0	0	0	90	0	0	0
	Total	1350	1870	1480	7600	0	1820	200	1940	700	1500
	Diversity	1.01	1.42	1.23	0.721	0	1.06	0.62	1.43	1.38	1.4

Table 1 (b) Ephemeropteran taxa densities (numbers.m⁻²), velocity (m.s⁻¹), depth (m) and Froude number for sample points 11 - 20.

Family	Sample No.	BRBSMo11	BRBSMo12	BRBSMo13	BRBSMo14	BRBSMo15	BRBSMo16	BRBSMo17	BRBSMo18	BRBSMo19	BRBSMo20
	Velocity (v)	0.33	0.34	0.08	0.42	0.3	1.15	0	0.05	0.14	0.09
	Depth (d)	0.41	0.36	0.28	90.0	0.12	0.24	0.24	0.03	0.16	9.0
	Fr (v/\gd)	0.16	0.18	0.05	0.47	0.28	0.75	00.00	60.0	0.11	0.04
Baetidae	Baetis harrisoni	350	1840	160	290	850	290	0	100	300	09
	Cloeodes sp nov 1	0	340	1340	0	240	0	2060	0	1040	20
	Pseudopannota maculosum	0	10	0	10	0	10	0	0	0	0
	Demoreptus capensis	0	09	0	09	80	280	0	20	0	0
	Cheleocloeon exisum	0	0	0	0	0	0	0	0	06	70
Heptageniidae	Afronurus sp.	40	390	180	160	430	30	40	20	260	40
Ephemerellidae	Lestagella sp.	200	280	220	380	320	30	140	150	400	10
Leptophlebiidae	Aprionyx peterseni	10	140	10	130	20	10	10	30	10	0
	Castanophlebia sp	0	0	10	0	0	0	0	0	0	10
Caenidae	Caenidae sp	0	0	0	0	0	0	0	0	0	0
	Total	009	3060	1920	1330	2270	096	2250	380	2100	210
	Diversity	0.929	1.27	0.983	1.38	1.5	0.972	0.349	1.45	1.36	1.55

Table 1 (c) Ephemeropteran taxa densities (numbers.m⁻²), velocity (m.s⁻¹), depth (m) and Froude number for sample points 21 - 30.

(d) (d) harrisoni harrisoni des sp nov 1 topannota maculosum reptus capensis ocloeon exisum rrus sp. yella sp. nyx peterseni dae sp	BRBSMo21 BRBSMo22 BRE	BRBSMo23 BRBSMo24	24 BRBSMo25	BRBSMo26	BRBSMo27	BRBSMo28	BRBSMo29	BRBSMo30
Depth (d) Fr (vl·\gd) Baetis harrisoni Cloeodes sp nov 1 Pseudopannota maculosum Demoreptus capensis Cheleocloeon exisum niidae Afronurus sp. etilidae Lestagella sp. castanophlebia sp Caenidae sp	0.08	0.49 0.76	0.95	0.57	0.7	0.22	1.2	0.87
Fr (v/·lgd) Baetis harrisoni Cloeodes sp nov 1 Pseudopannota maculosum Demoreptus capensis Cheleocloeon exisum niidae Afronurus sp. ellildae Lestagella sp. castanophlebia sp Castanophlebia sp	0.36	0.24 0.2	0.16	0.34	0.3	0.16	90.0	0.08
Baetis harrisoni Cloeodes sp nov 1 Pseudopannota maculosum Demoreptus capensis Cheleocloeon exisum niidae Afronurus sp. etilidae Lestagella sp. castanophlebia sp Caenidae sp	0.04	0.32 0.54	92.0	0.31	0.41	0.18	1.71	0.98
Cloeodes sp nov 1 Pseudopannota maculosum Demoreptus capensis Cheleocloeon exisum Afronurus sp. Lestagella sp. Aprionyx peterseni Castanophlebia sp	220	1600 80	370	230	240	280	0	210
Pseudopannota maculosum Demoreptus capensis Cheleocloeon exisum Afronurus sp. Lestagella sp. Aprionyx peterseni Castanophlebia sp	0	0 0	0	0	0	260	70	0
Demoreptus capensis Cheleocloeon exisum Afronurus sp. Lestagella sp. Aprionyx peterseni Castanophlebia sp	0	20 0	0	20	0	0	0	0
Cheleocloeon exisum Afronurus sp. Lestagella sp. Aprionyx peterseni Castanophlebia sp	0	0 310	330	260	540	90	470	1940
Afronurus sp. Lestagella sp. Aprionyx peterseni Castanophlebia sp	10	0 0	0	0	0	0	0	0
Lestagella sp. Aprionyx peterseni Castanophlebia sp	10	280 50	20	90	270	200	20	80
Aprionyx peterseni Castanophlebia sp Caenidae sp	0	540 20	130	130	70	400	90	180
Castanophlebia sp Caenidae sp	0	250 0	40	20	150	210	10	160
Caenidae sp	0	0	0	0	0	0	0	0
	0	0 0	0	0	0	0	0	0
Total 1230	240	2690 460	890	710	1270	1710	620	2870
Diveristy 1.48	0.345	1.12 0.948	1.24	1.43	1.42	1.61	0.837	1.01

Table 1 (d) Ephemeropteran taxa densities (numbers.m⁻²), velocity (m.s⁻¹), depth (m) and Froude number for sample points 31 - 40.

Family	Sample No.	BRBSMo31	BRBSMo32	BRBSMo33	BRBSMo34	BRBSMo35	BRBSMo36	BRBSMo37	BRBSMo38	BRBSMo39	BRBSM040
	Velocity (v)	0.92	0.7	1.1	1.1	0.88	0.73	0.51	0.01	0.05	0.15
	Depth (d)	0.13	0.15	0.3	0.25	0.2	0.4	0.13	0.08	0.08	0.22
	Fr (v/√gd)	0.82	0.58	0.64	0.70	0.63	0.37	0.45	0.01	90.0	0.10
Baetidae	Baetis harrisoni	029	710	150	250	280	740	200	30	120	620
	Cloeodes sp nov 1	0	06	0	0	0	10	0	0	3450	0
	Pseudopannota maculosum	0	02	0	40	10	0	09	0	0	20
	Demoreptus capensis	930	830	950	970	740	460	1070	90	0	520
	Cheleocloeon exisum	0	0	0	0	0	0	90	029	100	0
Heptageniidae	Afronurus sp.	09	06	40	10	20	06	120	10	0	220
Ephemerellidae	Lestagella sp.	100	490	130	110	70	150	029	10	170	360
Leptophlebiidae	Aprionyx peterseni	400	0	1000	130	10	170	1010	20	10	920
	Castanophiebia sp	10	10	0	0	0	0	0	0	0	0
Caenidae	Caenidae sp	0	0	0	0	0	0	0	0	0	0
	Total	2170	2290	2270	1510	1130	1620	3490	790	3850	2310
	Diversity	1.3	1.45	1.14	1.11	0.95	1.36	1.57	0.642	0.454	1.59

Table 2 Total, mean and standard deviation (SD) of Ephemeropteran densi ties (numbers.m⁻²)

Family	Species	Total	Mean	(#SD)
Baetidae	Baetis harrisoni	16320	408	(±387)
	Cloeodes sp nov 1.	16690	417	(±1152)
	Psuedopannota maculosum	430	10	(±22)
	Demoreptus capensis	15390	384	(±441)
	Cheleocloeon exisum	1350	33	(±116)
Heptageniidae	Afronurus sp.	4190	104	(±122)
Ephemerellidae	Lestagella sp.	7330	183	(±171)
Leptophlebiidae	Aprionyx peterseni	6130	153	(±241)
	Castanophlebia sp.	80	2	(45)
Caenidae	Caenidae sp.	20	-	(±7)

APPENDIX D

Regression summaries of Microhabitat Suitability Functions

Table 1 Regression summary for *Beatis harrisoni* densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are

	7)=12.141 p<.0 BETA	St. Err.	В	St. Err.	t(37)	p-level
9			1.40452	0.22411	6.26698	2.73E-07
,	2.37472	0.48193	4.82705	0.9796	4.92758	1.76E-05
2	-2.2873	0.48193	-3.7588	0.79196	-4.7462	3.08E-05
	0775199 R²= .£					
F(3,36	6)=12.044 p<.0			50 SW 11 W.		- 4-11-1
	BETA	St. Err.	В	St. Err.	t(36)	p-level
а			1.03852	0.24579	4.22524	0.000155
/	4.6819	0.94987	9.51682	1.93079	4.92899	1.87E-05
V ²	-8.23027	2.20785	-13.525	3.62823	-3.7277	0.000662
V ³	3.78214	1.37636	5.1588	1.87734	2.74793	0.009312
	8347195 R ² = .0					
F(2,3	7)=1.6165 p<.2 BETA	St. Err.	B	St. Err.	t(37)	p-level
	DETA	St. Ell.	1.70453	0.42871	3.97591	0.000313
a	0.94569	0.57103	5.88775	3.55513	1.65613	0.106154
d d	-1.01932	0.57103	-11.188	6.26737	-1.7851	0.082452
			ujusteu it	0290297		
F(5,3	4)=6.2632 p<.0		djusted R²= .4 rror of estimat			
F(5,3					t(34)	p-level
	4)=6.2632 p<.0	00032 Std.E	rror of estimat	e: .65335	3.61432	0.000963
a	4)=6.2632 p<.0	00032 Std.E	rror of estimat	e: .65335 St. Err.	E CONTRACTOR OF THE PARTY OF TH	0.000963 0.00017
a v	4)=6.2632 p<.0 BETA	St. Err.	rror of estimat B 1.63385	st. Err. 0.45205	3.61432	0.000963
a v d	4)=6.2632 p<.0 BETA 2.08826	St. Err. 0.4945	B 1.63385 4.24476	St. Err. 0.45205 1.00515	3.61432 4.22299	0.000963 0.00017
a v d vd	4)=6.2632 p<.0 BETA 2.08826 -0.14209	00032 Std.Err. St. Err. 0.4945 0.56587	B 1.63385 4.24476 -0.8847	e: .65335 St. Err. 0.45205 1.00515 3.523	3.61432 4.22299 -0.2511	0.000963 0.00017 0.803242
a v d vd v ²	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328	0.4945 0.56587 0.29622	B 1.63385 4.24476 -0.8847 4.16351	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436	3.61432 4.22299 -0.2511 2.03658	0.000963 0.00017 0.803242 0.04954
a v d vd vd v ² d ²	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403	0.4945 0.56587 0.29622 0.49975 0.51224	B 1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .:	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218	3.61432 4.22299 -0.2511 2.03658 -4.9814	0.000963 0.00017 0.803242 0.04954 1.82E-05
a v d vd vd v ² d ²	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R ² = . 7)=12.741 p<.0	00032 Std.Er. St. Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = 3	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947
a v d vd v ² d ² R= .6	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403	0.4945 0.56587 0.29622 0.49975 0.51224	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .:	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err.	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947
a v d vd v ² d ² R= .6 F(2,3	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R ² = . 7)=12.741 p<.0	0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R²= 3	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10
a v d vd v ² d ² R= .6 F(2,3	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R ² = . 7)=12.741 p<.0 BETA	0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err.	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = 3. rror of estima B 1.69618 3.28388	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05
a v d vd v² d² R= .6 F(2,3	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R ² = . 7)=12.741 p<.0	0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R²= 3	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10
a v v d d vvd vv² c² c² R= .6 Fr Fr² R= .6	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R²=. 7)=12.741 p<.0 BETA 1.44312 -1.56106	0.0032 Std.Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err. 0.30937 0.30937	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .: rror of estima B 1.69618 3.28388 -2.5471	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399 0.50478	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05
a v v d d vvd vv² c² c² R= .6 Fr Fr² R= .6	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R²=. 7)=12.741 p<.0 BETA 1.44312 -1.56106 37556072 R²=. 6)=10.074 p<.0	0.0032 Std.Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err. 0.30937 0.30937	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R²=	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399 0.50478 41108081 te: .64886	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465 -5.0459	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05 1.22E-05
a v v d d vvd vvd vv² c² c² Fr Fr² R= .6 F(3,3)	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R²=. 7)=12.741 p<.0 BETA 1.44312 -1.56106	0.0032 Std.Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err. 0.30937 0.30937	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .: rror of estima B 1.69618 3.28388 -2.5471	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399 0.50478 41108081 te: .64886 St. Err.	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465 -5.0459 t(36)	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05 1.22E-05
a v v d v d v v d v v d v v d v v d v v d v v d v v d v v d v v d v v d v d v v d v v d v v d v v d v v d v v d v v d v v d v v d v v d v d v v d v d v v d v d v v v d	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R²=. 7)=12.741 p<.0 BETA 1.44312 -1.56106 37556072 R²=. 6)=10.074 p<.0 BETA	00032 Std.Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err. 0.30937 0.30937 0.45638229 A 00006 Std.E St. Err.	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .: rror of estima B 1.69618 3.28388 -2.5471 djusted R ² = .:	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399 0.50478 41108081 te: .64886 St. Err. 0.22469	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465 -5.0459 t(36) 6.5422	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05 1.22E-05
a v v d v d v d v d v d v d v d v d v d	4)=6.2632 p<.0 BETA 2.08826 -0.14209 0.60328 -2.48947 -0.07403 3861861 R²=. 7)=12.741 p<.0 BETA 1.44312 -1.56106 37556072 R²=. 6)=10.074 p<.0	0.0032 Std.Err. 0.4945 0.56587 0.29622 0.49975 0.51224 40783373 A 00006 Std.E St. Err. 0.30937 0.30937	1.63385 4.24476 -0.8847 4.16351 -4.091 -0.8125 djusted R ² = .: rror of estima B 1.69618 3.28388 -2.5471	e: .65335 St. Err. 0.45205 1.00515 3.523 2.04436 0.82126 5.62218 37582475 te: .66800 St. Err. 0.19141 0.70399 0.50478 41108081 te: .64886 St. Err.	3.61432 4.22299 -0.2511 2.03658 -4.9814 -0.1445 t(37) 8.86172 4.66465 -5.0459 t(36)	0.000963 0.00017 0.803242 0.04954 1.82E-05 0.885947 p-level 1.11E-10 3.96E-05 1.22E-05

2.1416

1.15792

Fr3

2.07621

1.19439

1.79305

0.08137

Table 2 Regression summary for *Demoreptus capensis* densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

R= .78701205 R2=	.61938797	Adjusted R2= .598	881435
E/2 37\=30 106 pc	nonno Sta	Error of estimate	77392

	BETA	St. Err.	В	St. Err.	t(37)	p-level	_
а			0.21103	0.25714	0.82067	0.41708931	
V	1.82851	0.38263	5.37116	1.12397	4.77874	2.78873E-05	*
V ²	-1.14305	0.38263	-2.7145	0.90868	-2.9873	0.00497299	*

R= .78865866 R2= .62198249 Adjusted R2= .59048103

r(3,3	0)-19.740 p0	JUDOU Sta.LI	TOI OI COLITIC	10		
	BETA	St. Err.	В	St. Err.	t(36)	p-level
а			0.2943	0.30912	0.95204	0.34742525
v	1.46529	0.82667	4.30423	2.4283	1.77252	0.08477047
v ²	-0.20745	1.92149	-0.4926	4.56314	-0.108	0.91462499
V ³	-0.59542	1.19784	-1.1736	2.36109	-0.4971	0.6221571

R= .36526848 R²= .13342106 Adjusted R²= .08657896

F(2,3	7)=2.8483 p<.0	3/0/1 Sta.Er	ror or estima	te: 1.10/0			_
	BETA	St. Err.	В	St. Err.	t(37)	p-level	
а			2.06045	0.6014	3.42609	0.00151404	-
d	0.15892	0.55431	1.42978	4.98714	0.28669	0.7759468	
of	-0.51536	0.55431	-8.1741	8.79186	-0.9297	0.35853186	

R= .83943233 R2= .70464663 Adjusted	R ² = .66121232
F(5,34)=16.223 p<.00000 Std.Error of es	stimate: .71120

	BETA	St. Err.	В	St. Err.	t(34)	p-level	
а			1.37957	0.49207	2.80361	0.00828991	*
V	1.66678	0.37248	4.89607	1.09415	4.47479	8.1498E-05	*
d	-0.83281	0.42624	-7.4928	3.83491	-1.9538	0.05898641	
vd	0.31978	0.22313	3.18926	2.22536	1.43314	0.16094862	
V ²	-1.22988	0.37644	-2.9207	0.89397	-3.2671	0.00248626	*
ď	0.4622	0.38585	7.33092	6.11994	1.19787	0.23925044	

R= .80662377 R²= .65064190 Adjusted R²= .63175768 F(2,37)=34.454 p<.00000 Std.Error of estimate: .74147

-	BETA	St. Err.	В	St. Err.	t(37)	p-level	
а			0.36998	0.21246	1.74143	0.08991732	
Fr	1.64024	0.23763	5.39379	0.78142	6.90257	3.8237E-08	*
Fr ²	-1.04876	0.23763	-2.4729	0.5603	-4.4135	8.4878E-05	*

R= .81930016 R²= .67125276 Adjusted R²= .64385715 F(3.36)=24.502 p< 00000 Std Error of estimate: .72918

	BETA	St. Err.	В	St. Err.	t(36)	p-level	
а			0.15697	0.2525	0.62165	0.53808933	
Fr	2.42843	0.57433	7.98568	1.88865	4.22825	0.00015406	*
Fr ²	-3.07401	1.36817	-7.2482	3.22598	-2.2468	0.03087821	*
Fr ³	1.35279	0.90046	2.0165	1.34224	1.50234	0.14173126	

Table 3 Regression summary for *Afronurus* sp. densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

(2,01)=5.0015 p<.0 BETA	St. Err.	В	St. Err.	t(37)	p-level
1	DEIA	OL LII.	1.23006	0.21713	5.66506	1.7848E-06
	1.65292	0.55028	2.85082	0.94907	3.0038	0.00476186
2		0.55028	-2.4237	0.76728	-3.1588	0.00315103
	-1.7382	0.55026	-2.4231	0.70720	-3.1300	0.00010100
	1343148 R ² = . 6)=7.2400 p<.0					
(0,00	BETA	St. Err.	В	St. Err.	t(36)	p-level
			0.84197	0.23314	3.61148	0.00092063
,	4.53612	1.06186	7.82353	1.8314	4.27188	0.00013535
2	-9.16491	2.46815	-12.779	3.44147	-3.7133	0.00068967
3	4.7264	1.53862	5.47004	1.7807	3.07184	0.00403694
	3742976 R ² = . 7)=1.1052 p<.					
(2,5)	BETA	St. Err.	В	St. Err.	t(37)	p-level
9	DETA	Ot. Em	1.3995	0.36847	3.7981	0.00052574
1	0.61691	0.57843	3.25886	3.05559	1.06653	0.29309627
f	-0.75835	0.57843	-7.0623	5.38672	-1.3111	0.1979191
	8879315 R ² = 4)=2.1347 p<.					
	8879315 R²= 4)=2.1347 p<. BETA		ror of estima B	te: .67032 St. Err.	t(34)	p-level
F(5,34	4)=2.1347 p<. BETA	08495 Std.Er St. Err.	ror of estima B 0.99165	St. Err. 0.46379	2.13815	0.03977613
F(5,34	4)=2.1347 p<. BETA 1.54388	08495 Std.Er St. Err. 0.59793	nor of estima B 0.99165 2.66276	St. Err. 0.46379 1.03126	2.13815 2.58205	0.03977613 0.01430283
=(5,34 a v	4)=2.1347 p<. BETA 1.54388 0.60144	08495 Std.Er St. Err. 0.59793 0.68423	nor of estima B 0.99165 2.66276 3.17714	st. Err. 0.46379 1.03126 3.61449	2.13815 2.58205 0.879	0.03977613 0.01430283 0.38557151
F(5,34 a v d	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853	08495 Std.Er St. Err. 0.59793 0.68423 0.35818	0.99165 2.66276 3.17714 -0.9869	st. Err. 0.46379 1.03126 3.61449 2.09746	2.13815 2.58205 0.879 -0.4705	0.03977613 0.01430283 0.38557151 0.64100105
F(5,34 a v d d vd vd v ²	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818	08495 Std.Err. St. Err. 0.59793 0.68423 0.35818 0.60429	0.99165 2.66276 3.17714 -0.9869 -2.1308	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259	2.13815 2.58205 0.879 -0.4705 -2.5289	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919
F(5,34 a v d d vvd vvd	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853	08495 Std.Er St. Err. 0.59793 0.68423 0.35818	0.99165 2.66276 3.17714 -0.9869	st. Err. 0.46379 1.03126 3.61449 2.09746	2.13815 2.58205 0.879 -0.4705	0.03977613 0.01430283 0.38557151 0.64100105
F(5,34 a v d d vvd v ² d ² R= .1	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819	2.13815 2.58205 0.879 -0.4705 -2.5289	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919
F(5,34 a v d d vvd v ² d ² R= .1	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819	2.13815 2.58205 0.879 -0.4705 -2.5289	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919
F(5,34 a v d vd v ² d ² R= .1 F(2,3	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p<	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 03727958 A	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = -	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819	2.13815 2.58205 0.879 -0.4705 -2.5289 -1	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08
F(5,34 a v v d d vvd v ² d ² R= .11 F(2,3	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p<	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 03727958 A	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² =-rror of estima	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err.	2.13815 2.58205 0.879 -0.4705 -2.5289 -1	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496
F(5,34 a a v v d d vvd v ² d ² R= .1 F(2,3 a Fr	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p< BETA	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 .03727958 A 49517 Std.Err.	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = -	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08
F(5,34) A A A A A A A A A A A A A A A A A A A	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p<. BETA 0.44604 -0.47038	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 .03727958 A 49517 Std.Err. 0.39447 0.39447	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = - rror of estimal B 1.49435 0.86121 -0.6512	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708 0.76163 0.54611	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644 1.13075	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08 0.26543701
F(5,34 1) 1) 1) 1) 1) 1) 1) 1) 1) 1)	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R²= 7)=.71638 p<. BETA 0.44604 -0.47038	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 .03727958 A 49517 Std.Err. 0.39447 0.39447	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = - rror of estimal B 1.49435 0.86121 -0.6512	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708 0.76163 0.54611	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644 1.13075	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08 0.26543701
F(5,34) a y d d d v d f F(2,3) F(2,3) F(3,3)	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p<. BETA 0.44604 -0.47038	0.59793 0.68423 0.35818 0.60429 0.61939 0.03727958 A 49517 Std.E St. Err. 0.39447 0.39447	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = - rror of estima B 1.49435 0.86121 -0.6512	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708 0.76163 0.54611 05217016 ate: .69845	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644 1.13075 -1.1924	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08 0.26543701 0.24067844
F(5,34 a v v d d vvd vv² c² R= .1 F(2,3 a Fr Fr² R= .3	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R ² = 7)=.71638 p<. BETA 0.44604 -0.47038 0.5366672 R ² = 6)=1.7155 p< BETA	08495 Std.Err. 0.59793 0.68423 0.35818 0.60429 0.61939 .03727958 A 49517 Std.E St. Err. 0.39447 0.39447 .12508015 A .18113 Std.E St. Err.	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = - rror of estima B 1.49435 0.86121 -0.6512	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708 0.76163 0.54611 05217016 ate: .69845 St. Err.	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644 1.13075 -1.1924	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08 0.26543701 0.24067844 p-level
F(5,34 a v v d d vvd vv² cd² R= .1 F(2,3) a Fr Fr² R= .3	4)=2.1347 p<. BETA 1.54388 0.60144 -0.16853 -1.52818 -0.61938 9307922 R²= 7)=.71638 p<. BETA 0.44604 -0.47038	0.59793 0.68423 0.35818 0.60429 0.61939 0.03727958 A 49517 Std.E St. Err. 0.39447 0.39447	0.99165 2.66276 3.17714 -0.9869 -2.1308 -5.7681 djusted R ² = - rror of estima B 1.49435 0.86121 -0.6512 djusted R ² = -	te: .67032 St. Err. 0.46379 1.03126 3.61449 2.09746 0.84259 5.76819 te: .72269 St. Err. 0.20708 0.76163 0.54611 05217016 ate: .69845 St. Err. 0.24186	2.13815 2.58205 0.879 -0.4705 -2.5289 -1 t(37) 7.21644 1.13075 -1.1924 t(36) 5.11128	0.03977613 0.01430283 0.38557151 0.64100105 0.01624919 0.32438496 p-level 1.462E-08 0.26543701 0.24067844 p-level 1.0689E-05

Table 4 Regression summary for Lestagella sp. densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

)=5.7669 p<.0 BETA	St. Err.	В	St. Err.	t(37)	p-level
9			1.41434	0.21236	6.6601	8.0758E-08
,	1.83554	0.54153	3.14624	0.92822	3.38955	0.00167605
2	-1.80027	0.54153	-2.4947	0.75042	-3.3244	0.00200666
				4057474		
)528827 R ² = .3 6)=6.9386 p<.0					
	BETA	St. Err.	В	St. Err.	t(36)	p-level
9			1.07209	0.23353	4.59071	5.2095E-05
/	4.39398	1.07027	7.5316	1.83452	4.10549	0.00022133
v ²	-8.39046	2.4877	-11.627	3.44733	-3.3728	0.00179129
V ³	4.19403	1.55081	4.82395	1.78374	2.70441	0.01038349
	9331631 R²= .° 7)=3.3857 p<.0					
(2,0)	BETA	St. Err.	В	St. Err.	t(37)	p-level
а			1.8192	0.3466	5.24876	6.5255E-06
d	0.63426	0.54746	3.32987	2.87416	1.15855	0.25406247
d ²	-0.96179	0.54746	-8.9016	5.06688	-1.7568	0.08722186
	0407424 R ² = .(4)=3.9071 p<.(BETA				t(34)	p-level
а			1.8769	0.42105	4.45767	8.5702E-05
v	1.39961	0.5462	2.39903	0.93623	2.56244	0.01499427
d	-0.19882	0.62503	-1.0438	3.28142	-0.3181	0.75235271
vd	0.51713	0.3272	3.00955	1.90418	1.5805	0.12325235
v ²	-1.80892	0.55201	-2.5067	0.76495	-3.277	0.00242139
ď	-0.27106	0.5658	-2.5087	5.23666	-0.4791	0.63496172
	0762530 R ² = . 7)=1.9337 p<.				Harris	
F(Z,3		St. Err.	B	St. Err.	t(37)	p-level
_	BETA	St. EII.	1.66668	0.19957	8.35119	4.8875E-10
a	0.74000	0.38254	1.43689	0.73404	1.95751	0.0578622
Fr F-2	0.74882		-0.9808	0.73404	-1.8635	0.07034629
Fr ²	-0.71285	0.38254	-0.3600	0.02003	1,3000	
	2041479 R ² = 6)=4.4571 p<.					
-	BETA	St. Err.	В	St. Err.	t(36)	p-level
			1.30326	0.21944	5.93915	8.3802E-07
а			1.00020			
a Fr	3.05336	0.85536	5.85899	1.64132	3.56968	0.00103569
a Fr Fr ²	3.05336 -6.63433	0.85536 2.03761			3.56968 -3.2559	0.00103569 0.00246497

Table 5 Regression summary for Aprionyx peterseni densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

1-1-1	r)=4.1363 p<.0 BETA	St. Err.	В	St. Err.	t(37)	p-level
2	52.77		0.91005	0.27658	3.2904	0.00220332
,	1.11931	0.56069	2.41334	1.20892	1.99628	0.05330151
2	-0.77153	0.56069	-1.3449	0.97735	-1.376	0.17708513
	1130335 R ² = .2 6)=4.2476 p<.0					
	BETA	St. Err.	В	St. Err.	t(36)	p-level
9			0.57343	0.31715	1.80805	0.07895968
,	3.1198	1.15551	6.72661	2.49139	2.69994	0.01049965
2	-5.92451	2.68582	-10.327	4.6817	-2.2058	0.03386382
v ³	3.27938	1.67432	4.74465	2.42243	1.95863	0.05793791
	4778497 R²= . 7)=2.5456 p<.0			e: .86330	The state of the s	
	BETA	St. Err.	В	St. Err.	t(37)	p-level
a			1.33138	0.44459	2.99459	0.00487867
d	0.66941	0.55828	4.42073	3.68683	1.19906	0.23812744
d ^e	-0.93801	0.55828	-10.92	6.49953	-1.6802	0.10135012
	5436868 R ² = . 4)=3.0170 p<.0	02329 Std.Er	ror of estimat	te: .79943		a taval
	BETA	St. Err.	В	St. Err.	t(34)	p-level
a			1.53612	0.55312	2.77718	0.00885707
V	0.68848	0.57043	1.48443	1.2299	1.20696	0.23577958
d	-0.31739	0.65275	-2.096	4.31071	-0.4862	0.62992316
vd	0.62147	0.34171	4.54945	2.50146	1.81871	0.0777731
V ²	-0.86059	0.57649	-1.5001	1.00489	-1.4928	0.14470975
of	-0.14535	0.5909	-1.6921	6.87925	-0.246	0.80717808
	7412170 R²= . 7)=5.3645 p<.					
-	BETA	St. Err.	В	St. Err.	t(37)	p-level
a			0.94692	0.2323	4.07633	0.00023275
Fr	1.12275	0.35397	2.71001	0.85439	3.17187	0.00304174
Fr ²	-0.90626	0.35397	-1.5685	0.61262	-2.5602	0.01467598
	49055477 R ² = 36)=3.8029 p<.					
	STREET, STREET	to the management	В	St. Err.	t(36)	p-level
	BETA	St. Err.				
F(3,3	BETA	St. Ell.	0.8098	0.28168	2.87488	0.00674589
F(3,3			0.8098 4.37849	0.28168 2.1069	2.87488 2.07817	0.00674589 0.04488285
F(3,3	1.814 -2.68241	0.87288 2.07937				200000000000000000000000000000000000000

Table 6 Regression summary for *Cheleocloeon excisum* densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

R= .44077190 R2= .19427987	Adjusted R2= .15072743
F(2,37)=4.4608 p<.01838 Std.	Error of estimate: .75074

	BETA	St.	Err.	В		St.	Err.	t(37)	p-level	
а					1.0909		0.24944	4.3733	4 9.5807E-0	5 *
v	-1.25098		0.55672		-2.45		1.09031	-2.247	1 0.0306870	6 *
V ²	0.91585		0.55672		1.45008		0.88146	1.6450	8 0.1084233	8

R= .47336889 R²= .22407811 Adjusted R²= .15941795 F(3,36)=3.4655 p<.02607 Std.Error of estimate: .74689

	BETA	St.	Err.	В		St.	Err.	t(36)	p-level	
а			Children		1.27904		0.29528	4.33167	0.000113	3 *
V	-2.48192		1.18437	7	-4.8607		2.31953	-2.0956	0.0432160	8 *
V ²	4.08655		2.7529	9	6.47033		4.35874	1.48445	0.1463931	9
V ³	-2.01785		1.71614	1	-2.6518		2.25532	-1.1758	0.2473879	2

R= .33388871 R²= .11148167 Adjusted R²= .06345366 F(2,37)=2.3212 p<.11229 Std.Error of estimate: .78838

BETA	St.	Err.	В		St. Err.	t(37)	p-level	
				1.26283	0.40601	3.11036	0.00358852	*
-1.2066		0.56128		-7.2378	3.36685	-2.1497	0.03818309	*
1.13724		0.56128		12.026	5.93544	2.02614	0.05000556	*
	-1.2066	-1.2066	-1.2066 0.56128	-1.2066 0.56128	1.26283 -1.2066 0.56128 -7.2378	1.26283 0.40601 -1.2066 0.56128 -7.2378 3.36685	1.26283 0.40601 3.11036 -1.2066 0.56128 -7.2378 3.36685 -2.1497	1.26283 0.40601 3.11036 0.00358852 -1.2066 0.56128 -7.2378 3.36685 -2.1497 0.03818309

R= .54243834 R²= .29423936 Adjusted R²= .19045103 F(5,34)=2.8350 p<.03033 Std.Error of estimate: .73298

	BETA	St. Err.	В	St. Err.	t(34)	p-level
a			2.06193	0.50714	4.06578	0.00026806
v	-1.28413	0.57579	-2.5149	1.12766	-2.2302	0.03244061
d	-1.3495	0.65889	-8.095	3.95236	-2.0481	0.04833128
vd	0.44622	0.34492	2.96712	2.29352	1.2937	0.20448989
v2	0.6323		1.00113	0.92135	1.08659	0.28486344
ď	1.03527	0.59645	10.9478	6.30738	1.73572	0.09166975

R= .38005944 R²= .14444518 Adjusted R²= .09819897 F(2,37)=3.1234 p<.05579 Std.Error of estimate: .77361

- CONTRACTOR - CON									
	BETA	St. Err.	В		St.	Err.	t(37)	p-level	
a				0.90987		0.22167	4.10469	0.000214	A
Fr	-0.79353	0.37186		-1.7398		0.81529	-2.1339	0.03954278	*
Fr ²	0.52628	0.37186		0.82734		0.58459	1.41526	0.16535509	

R= .43753772 R²= .19143926 Adjusted R²= .12405920 F(3,36)=2.8412 p<.05140 Std.Error of estimate: .76244

	BETA	St.	Err.	В		St. Err.		t(36)	p-level
а				170	1,12432	0.26	6402	4.25848	0.00014085
Fr	-1.98369		0.90072		-4.3491	1.9	7479	-2.2023	0.03413131
Fr ²	3.58438		2.14568		5.63483	3.3	7311	1.67051	0.10349108
Fr ³	-2.0427		1.41217		-2.0301	1.4	0346	-1.4465	0.15669017

Table 7 Regression summary for *Pseudopannota maculosum* densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are

R= .41353786 R²= .17101356 Adjusted R²= .12620348 F(2,37)=3.8164 p<.03113 Std.Error of estimate: .64618

	BETA	St.	Err.	В		St.	Err.	t(37)	p-level
а					-0.0576		0.2147	-0.2681	0.79007947
v	1.48544		0.5647		2,46858		0.93845	2.63049	0.01235368
V2	-1.30588		0.5647		-1.7545		0.75869	-2.3125	0.02641471

R= .42086420 R²= .17712667 Adjusted R²= .10855389 F(3,36)=2.5830 p<.06838 Std.Error of estimate: .65267

	BETA	St.	Err.	В		St.	Err.	t(36)	p-level
а					-0.1299		0.25803	-0.5034	0.61777836
v	2.04297		1.21967		3.39511		2.02692	1.67501	0.1025981
V ²	-2.74201		2.83497		-3.684		3.80888	-0.9672	0.33989331
V3	0.91396		1.7673		1.0192		1.97081	0.51715	0.60821497

R= .26470639 R2= .07006947 Adjusted R2= .01980296

F(2,37)=1.3940 p<.26082 Std.Error of estimate: .68439

	BETA	St. Err.	В		St. Err.	t(37)	p-level
а				0.4522	0.35246	1.28299	0.2074758
d	0.29165	0.5742	1	1.48451	2.92277	0.50791	0.61453003
ď	-0.53248	0.5742	1	-4.7781	5.15257	-0.9273	0.3597734

R= .45156186 R²= .20390811 Adjusted R²= .08683577

F(5,34)=1.7417 p<.15170 Std.Error of estimate: .66057

	BETA	St. Err.	В	St. Err.	t(34)	p-level
а			-0.003	4 0.45705	5 -0.0074	0.99412549
v	1.40951	0.61153	2.342	4 1.01627	7 2.3049	0.02740546
d	0.14493	0.69979	0.7377	2 3.56195	0.20711	0.8371582
vd	-0.1525	0.36633	-0.860	5 2.06697	7 -0.4163	0.67981166
v ²	-1.12912	0.61803	-1.51	7 0.83034	4 -1.827	0.07649498
ď	-0.25797	0.63347	-2.314	8 5.68434	4 -0.4072	0.68639648

R= .40563582 R2= .16454042 Adjusted R2= .11938044

F(2,37)=3.6435 p<.03594 Std.Error of estimate: .64870

	BETA	St.	Err.	В		St.	Err.	t(37)	p-level	
a					0.04848		0.18587	0.2608	5 0.79565692	2
Fr	0.97822		0.36747		1.8199		0.68365	2.6620	4 0.01142519	*
Fr ²	-0.82541		0.36747		-1.1011		0.49019	-2.246	2 0.03074885	5 *

R= .41728323 R²= .17412529 Adjusted R²= .10530240

F(3,36)=2.5300 p<.07253 Std.Error of estimate: .65386

	BETA	of BETA	В		of	В	t(36)	p-level	
а				-0.0337		0.226419	-0.14882		0.882527
Fr	1.51572	0.910313		2.819864		1.693558	1.665054		0.104583
Fr2	-2.2065	2.168528		-2.9434		2.892747	-1.01751		0.315701
Fr ³	0.922521	1.427213		0.777976		1.203591	0.646379		0.522134

Table 8 Regression summary for Cloeodes sp. nov. 1. densities by velocity, depth and Froude number for quadratic and cubic functions as well as interaction terms. Significance level (P) and Standard Error (St.Err.) for the individual regression coefficients (B) are shown.

R= .45162034 R²= .20396094 Adjusted R²= .16093180 F(2.37)=4,7401 p<.01470 Std.Error of estimate: 1.2167

	BETA	St.	Err.	В		St.	Err.	t(37)	p-level
а					1.98649		0.40425	4.91405	1.8392E-05
v	-1.01643		0.55336		-3.2456		1.76695	-1.8368	0.07427905
v ²	0.61762		0.55336		1.59438		1.4285	1.11612	0.27156678

R= .45559131 R²= .20756345 Adjusted R²= .14152707

F(3,36)=3.1432 p<.03693 Std.Error of estimate: 1.2306

	BETA	St.	Err.	В		St.	Err.	t(36)	p-level
а		0 9 8 8			1.87983		0.48652	3.86382	0.00044772
V	-0.58843		1,1969)	-1.8789		3.82185	-0.4916	0.62596577
V ²	-0.48484		2.78205	5	-1.2516		7.18182	-0.1743	0.86262554
V ³	0.70161		1.73431		1.50333		3.71605	0.40455	0.68820238

R= .06223417 R2= .00387309 Adjusted R2= ----

F(2,37)=.07193 p<.93073 Std.Error of estimate: 1.3610

8 11	BETA	St.	Err.	В		St. Err.	t(37)	p-level
а					0.81479	0.70091	1.16249	0.25248134
d	0.20215		0.5943		1.97706	5.8123	0.34015	0.73566622
ď	-0.22183		0.5943		-3.8246	10.2466	-0.3733	0.71108526

R= .54605964 R²= .29818113 Adjusted R²= .19497247

F(5,34)=2.8891 p<.02803 Std.Error of estimate: 1.1917

	BETA	St. Err.	В		St. Err.	t(34)	p-level
а				0.84301	0.82454	1.0224	0.31381249
V	-1.06783	0.57418		-3.4097	1.83341	-1.8598	0.07158981
d	1.24117			12.1388	6.42599	1.88901	0.06744468
vď	-0.55398	0.34395		-6.0059	3.72895	-1.6106	0.11651009
v ²	1.0559			2.72579	1.49799	1.81963	0.07763037
de	-1.11649	1000000000		-19.25	10.2549	-1.8771	0.06910235

R= .51556452 R2= .26580677 Adjusted R2= .22612065

F(2,37)=6.6977 p<.00329 Std.Error of estimate: 1.1684

11000	BETA	St	Err.	В		St. Err.	t(37)	p-level	
а					1.99709	0.3348	5.965	7.0012E-07	-
Fr	-1.23962		0.34448		-4.4312	1.23139	-3.59	0.00093195	5 *
Fr ²	1.03715		0.34448		2.65834	0.88294	3.010	0.00467514	1 '

R= .51959888 R2= .26998299 Adjusted R2= .20914824

F(3,36)=4,4380 p<.00939 Std.Error of estimate: 1.1812

	BETA	St. Err.	В		St. Err.	t(36)	p-level
а		227 (1174))		2.10132	0.40902	5.13744	9.8652E-06
Fr	-1.59441	0.85585	;	-5.6994	3.05936	-1.8629	0.07064468
Fr ²	1.94879	2.0388	3	4.99496	5.22567	0.95585	0.34552205
Fr ³	-0.60894	100000000000000000000000000000000000000	3	-0.9867	2.17425	-0.4538	0.65268785