1 Chaos and irregularity in karst percolation

- 2 Gregoire Mariethoz, ^{1,2} Andy Baker, ¹ Bellie Sivakumar, ^{2,3} Adam Hartland, ^{1,4}
- 3 and Peter Graham¹
- 4 Received 17 October 2012; accepted 7 November 2012; published XX Month 2012.
- 5 [1] This paper focuses on analyzing chaos in cave percola-6 tion water drip rates, which has implications for flow routing 7 in fractured media and on the use of speleothems for paleo-8 climate reconstructions. It has been shown that the physics of 9 dripping faucets involve a set of non-linear equations leading 10 to chaotic drip rate, meaning that, for a given drip rate, the 11 interval between individual drops can vary greatly. It can be 12 expected that drip waters supplying stalagmites show similar 13 properties, and consequently the dependency between water 14 flux and stalagmite growth rate or geochemistry could be more 15 complicated than usually assumed. We used high-frequency 16 monitoring of two contrasting drips in a cave in Australia, 17 and identified chaos in cave drip rate. Our findings also indi-18 cate that the occurrence of chaos can give insights into flow 19 routing in fractured media. Citation: Mariethoz, G., A. Baker, 20 B. Sivakumar, A. Hartland, and P. Graham (2012), Chaos and irreg-21 ularity in karst percolation, Geophys. Res. Lett., 39, LXXXXX, 22 doi:10.1029/2012GL054270.

23 1. Introduction

[2] Stalagmite dripping is influenced by a range of 25 hydrologic, climatic and physical processes [e.g., Jex et al., 26 2012; Mariethoz et al., 2012; Miorandi et al., 2010]. How-27 ever, very few studies have looked quantitatively at drip rate 28 variability on short timescales (<10 minutes). In contrast, 29 several studies have identified chaos in dripping faucets and 30 described the process both experimentally and theoretically 31 [Coullet et al., 2005; D'Innocenzo and Renna, 1996; Martien 32 et al., 1985; Pinto et al., 2001]. The mechanism is seen as 33 a continuous flow of water that builds up until the drop 34 becomes too heavy to be held by surface tension. When it 35 reaches this threshold point, the drop detaches and falls. This 36 induces a rebound mechanism and vibrations in the residual 37 water, affecting the time of formation of the next drop. Since 38 each falling drop influences the motion of the next forming 39 drop, the time intervals T_n between successive drop detach-40 ments become irregular or chaotic at certain discharges. It has

been shown that this process can be represented as an oscil- 41 lation according to a mass-spring model [Dreyer and Hickey, 42 1991; Sartorelli et al., 1994], with a threshold associated to 43 the drop detachment. These physics involve a set of non- 44 linear equations leading to chaotic drip intervals. Such a 45 chaotic behavior can be described as a system where a small 46 perturbation in the input parameters can result in dispropor- 47 tionate consequences on the outcome [see *Lorenz*, 1963]. It 48 has been observed that depending on the flow rate, the 49 interval between two drops can either be a constant value 50 (constant regular dripping), a superposition of discrete fre- 51 quencies (irregular but somewhat predictable dripping over a 52 reasonably long time horizon) or alternatively have a com- 53 pletely chaotic structure [Fuchikami et al., 1999]. Minute 54 differences in flow rate can trigger state transitions in this 55 continuum from regularity to chaos.

- [3] It is expected that similar processes can occur within 57 fracture-fed cave drip waters. The only experiments aimed at 58 identifying non-linear cave drip rates were measuring drip 59 rates averaged over periods of 10 minutes [Baker and 60] Brunsdon, 2003; Genty and Deflandre, 1998], and there- 61 fore were not able to clearly identify or quantify chaos. To 62 the best of our knowledge, the present study is the first to 63 draw a parallel between the fields of non-linear physics and 64 karst hydrogeology. The major difference between a leaky 65 faucet and a cave drip is that the cave drip is not a controlled 66 laboratory experiment and, therefore, several external factors 67 intervene. In particular, two main factors may influence the 68 transition between stages of regular/irregular/chaotic drip- 69 ping rates:
- [4] 1. The flow rate is controlled by hydrologic processes, 71 such as recharge, groundwater flow, evaporation, etc., which 72 necessarily induce variations in a karst system on different 73 timescales (e.g., daily air pressure changes, seasonal recharge 74 variability). In particular, rainfall has been shown to exhibit 75 chaotic behavior [e.g., Rodriguez-Iturbe et al., 1989] (see 76 Sivakumar [2004] for a review).

70

- [5] 2. A drip is fed by a series of interconnected fractures 78 and stores of various sizes which can be conceptualized as a 79 series of buckets that exchange water under unsaturated con- 80 ditions [Bradley et al., 2010]. Although these local processes 81 are difficult to investigate in detail, they are known to be non- 82 linear [Baker and Brunsdon, 2003; Genty and Deflandre, 83 1998] and could cause irregular flow rate feeding the drip.
- [6] To date, all experimental dripping faucet studies have 85 identified chaos for drip rates of at least 2 drops per second. 86 Cave drip rates are often much lower, of the order of one 87 drop per second to two drops per minute or less, especially 88 for stalagmite-forming drips. Such low drip rates have not 89 been studied previously, and the possibility of chaos to occur 90 in such environments remains to be investigated.

LXXXXX 1 of 6

¹Connected Waters Initiative Research Centre, National Centre for Groundwater Research and Training, University of New South Wales, Sydney, New South Wales, Australia.

²School of Civil and Environmental Engineering, University of New South Wales, Sydney, New South Wales, Australia.

³Department of Land, Air and Water Resources, University of California, Davis, California, USA.

⁴Department of Chemistry, University of Waikato, Hamilton, New Zealand.

Corresponding author: G. Mariethoz, School of Civil and Environmental Engineering, University of New South Wales, Sydney, NSW 2052, Australia. (gregoire.mariethoz@minds.ch)

^{©2012.} American Geophysical Union. All Rights Reserved. 0094-8276/12/2012GL054270

168

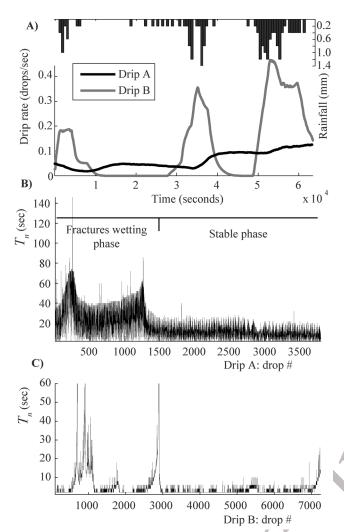


Figure 1. (a) Hydrograph of drips A and B, and rainfall averaged over a 10-minute interval. T_n for the entire recorded period at drips (b) A and (c) B.

[7] In this paper, we seek to establish whether chaos and 93 irregularity can occur in cave percolation water drip rates, 94 and, if it does, what insights it can provide into the related 95 hydrogeological processes. This is accomplished using high-96 frequency monitoring of two drips in a cave in Australia 97 during a recharge event. Not only have we identified the 98 occurrence of irregular and chaotic behavior, but we also 99 observed that the degree of disorder in the drip rate is related 100 to the degree of connectivity to the recharge and the influence 101 of a groundwater store. This dependence on the hydro-102 geological setting gives insights into flow routing in fractures. 103 Another immediate consequence is that the interval between 104 drops can vary greatly, depending on the flow rate feeding the 105 drip. The importance of such non-linearity is that the resultant 106 hydrologic variability may imprint on the signal of stalagmite 107 proxies which show a strong dependency on water supply. 108 For example, both δ^{13} C and δ^{18} O undergo within-cave frac-109 tionation, which can be dependent on the rate of water supply 110 vs. magnitude of within-cave processes, such as ventilation 111 [Hendy, 1971; Spötl et al., 2005], and speleothem growth rate 112 itself is in part dependent on water supply [Drevbrodt, 1999]. 113 Fractionation effects resulting from hydrologic variations can

presumably introduce a second order variability in proxies, 114 reducing the climatic signal retained. 115

2. Site and Setting

[8] To test the occurrence of chaotic behavior in cave drip 117 rates, we chose two specific drips near the main entrance of 118 the Cathedral Cave, part of the Wellington Caves complex in 119 New South Wales, Australia. Cathedral Cave, one of many 120 caves that form part of the larger Wellington Caves Reserve 121 (32°37′S; 148°56′E) is located west of the Blue Mountains, 122 part of the Great Dividing Range mountain belt that runs 123 approximately North-South along the Eastern sea-board of 124 Australia, approximately 7 km south of the town of Wel- 125 lington, New South Wales. The caves have developed in 126 folded Devonian limestone, at the boundary between two 127 distinct facies: a massive, marmorised limestone and a thinly 128 bedded limestone [Johnson, 1975]. Orogenetic development 129 during the mid-Devonian and early Carboniferous resulted 130 in widespread folding and the Lachlan fold belt in which the 131 caves are situated [Osborne, 2007]. Caves have developed 132 along faults and vertical joints, with the widespread presence 133 of paleokarst suggesting multiphase cave development 134 [Osborne, 2007].

[9] The two monitored drips are located in the massive 136 limestone. In dry periods, these drips are usually inactive. 137 Both drip locations are in the shallowest part of the cave, 138 approximately 2 to 5 meters below ground surface, therefore 139 they are well connected to the surface and responsive to 140 rainfall events. Drip A originates from a stalactite that is 141 located on a fracture in the cave ceiling, and which is actively 142 forming a stalagmite on the show-cave path. Since the drip is 143 located about 5 meters below ground level and has reached 144 super-saturation with respect to calcite, it can be assumed that 145 the drip is not directly fed by direct runoff from the surface 146 and that it is representative of karst drip water flow. Dripping 147 is continuous despite the intermittence of rainfall, confirming 148 that some storage is involved. Drip B is close to the surface 149 (about 2 meters), with no overlying soil cover, therefore it is 150 directly fed by surface runoff. It can be assumed to directly 151 reflect rainfall, with very little storage involved. The dripping 152 stops whenever rainfall stops, confirming a direct connection 153 with the surface.

[10] Both drips were recorded with Stalagmate acoustic 155 drip logger devices set to count the number of drops at every 156 2 second interval, meaning that individual drops can be captured if they are more than 2 seconds apart, or if drops occur 158 with a frequency lower than 0.5 drops/sec. Measurement time 159 was limited by the storage capacity of the loggers, which at 2 s 160 logging intervals allowed continuous measurement for less 161 than 24 hours. Therefore, we deployed the loggers during a 162 rainfall event, and the drips were recorded between the 22nd of 163 November 2011 at 2:30 PM and the 23rd of November 2011 at 164 8:10 AM, for a total 17 hours and 40 minutes.

[11] The hydrographs of drips A and B, computed using a 166 moving average of drop occurrence, are shown in Figure 1a. 167

3. Quantifying Irregularity and Chaos

[12] In most leaky faucet studies, a quantity of interest is 169 the interval between two successive drops, $T_n = t_{n+1} - t_n$, 170 with t_n representing the moment when drop n occurred. 171 Figures 1b and 1c show the record of T_n for drips A and B. 172

245

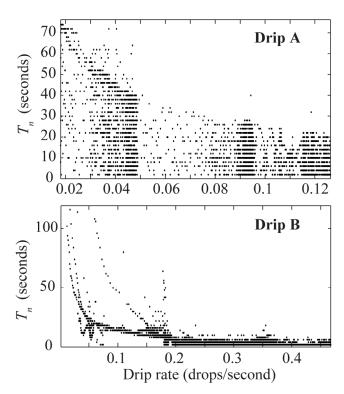


Figure 2. Dripping spectra of drops A and B (note the different axes).

173 When representing T_n , one no longer deals with a time 174 series, since the X axis is not an absolute measure of time, 175 but instead represents the number of drops since the begin-176 ning of the record. The Y axis is the time elapsed between 177 the previous and the current drop. The flow rate at both drips 178 was not constant during the measurement period, showing 179 an overall increasing trend, which translated in intervals 180 between drips being globally larger at the beginning of the 181 sequence than at the end. Studying variations of T_n allows 182 the determination of behaviors where, for example, there is a 183 systematic pattern of two drops coming after each other, then 184 a pause. Such an irregular pattern is clearly visible at the end 185 of the sequence in Figure 1b.

[13] A useful measure of variability is to represent the 187 dripping spectrum, which is the occurrence of intervals T_n as 188 a function of the averaged drip rate over a ten minute period. 189 The dripping spectra of both drips are shown in Figure 2. Each 190 point in the graph represents the interval measured between 191 two successive drops, and under which average flow rate this 192 interval occurred. Dense areas of the spectrum correspond to 193 rates for which a large array of values of T_n are observed. 194 Empty areas denote intervals that are never observed for a given drip rate. Dripping spectra [D'Innocenzo and Renna, 196 1997; Martien et al., 1985] are a way to distinguish regions 197 of stability and chaos, certain rates resulting in regular drip-198 ping while others cause the emergence of several distinguish-199 able intervals between drops.

[14] Since the experimental setting does not allow mea-201 suring T_n with a greater precision than 2 seconds, points are 202 disposed along horizontal lines 2 seconds apart. However, it 203 is clear that certain flow rates result in more variability in T_n . 204 For drip A, it is noted that specific drip rates occurred more

frequently (0.035-0.049 drops/second, 0.089-0.096 drops/ 205 second, and above 0.114 drops/second), during which a 206 wide range of dripping intervals T_n is observed, indicating 207 irregularity. The dripping spectrum of drip B is very differ- 208 ent and indicates a limited number of specific intervals for 209 a wide range of drip rates between 0.010 and 0.20 drops/ 210 second. It displays clear bifurcations, with T_n values taking 211 two or more distinct modes, corresponding to a typical 212 pattern observed in leaky faucet experiments.

[15] The succession of intervals between drops is further 214 analyzed by considering the joint distribution $f(T_n, T_{n+1})$, 215 which is represented by plotting a point for each drop n, 216 whose X coordinate is the interval between T_{n-1} and T_n , and 217 the Y coordinate is the interval between T_n and T_{n+1} . Figure 3 218 (top) shows such joint distributions for both drips. Although 219 both drips show irregularity, one sees that drip B presents 220 certain recurrent patterns of intervals. Such features can be a 221 sign of non-linear, chaotic behavior.

[16] Since the data were collected during the initial stage of a 223 recharge event, temporal analysis of the dripping irregularity 224 can give insights into the fractured rock wetting processes. The 225 time-varying joint distribution of $f(T_n, T_{n+1})$ for both drips for 226 intervals of 500 drops are displayed in Figures S1 and S2 in 227 the auxiliary material. Drip A shows that the first 500 drops 228 exhibit the most variability, with no definite pattern in the 229 succession of intervals. Between the 501st and the 1000th 230 drop, a pattern seems to take place involving mostly suc- 231 cessions of 20 s after 20 s intervals, 5 s after 40 s intervals, 232 and 20 s after 40 s intervals. The next period, between drops 233 1001 and 1500, seems chaotic again, but less than the initial 234 one. The last three periods considered, consisting of drops 235 1501–2000, 2001–2500 and 2501–3000, seem to settle in a 236 pattern involving a constant irregular pattern of large inter- 237 vals followed by short intervals, also visible as oscillations at 238 the end of the series in Figure 1b. This suggests that from the 239 1500th drop onwards, the system has settled into a stable 240 state. Drip B shows a very different temporal behavior, with 241 no specific wetting pattern at the initial stage of the rain 242 event, and an irregular succession of stable stages (drops 1 243 500, 2001–2500, 3001–5000, 5501–6500) and unstable ones 244 (drops 501–2000, 2501–3000, 5001–6000, 6501–7000).

[17] Following up on our examination of the irregularity in 246 drip intervals, we investigated the presence of chaos using 247 the surrogate data method [*Theiler et al.*, 1992]. This method 248 consists of generating synthetic time series that possess 249 similar statistical characteristics as the data (e.g., distribution, 250 autocorrelation, frequency spectrum), but are nevertheless 251 the outcomes of a linear process. The identification of sig- 252 nificant differences between the original data and the surro- 253 gates is an indication of chaotic behavior. The method of 254 Schreiber and Schmitz [1996] was used to generate surro- 255 gates for both drips. Figure 3 (bottom) shows the joint dis- 256 tributions of the surrogates. While the surrogate for drip A 257 can reasonably well reproduce the irregularity of drip A, the 258 surrogate of drip B fails to present the specific recurrent 259 frequencies observed.

[18] As in the work by Martien et al. [1985], we used 261 entropy to further characterize the joint distributions. A dis- 262 tribution with high entropy means that a wide variety of drop 263 intervals is observed, and is a sign of disorder and lack of 264

¹Auxiliary materials are available in the HTML. doi:10.1029/ 2012GL054270.

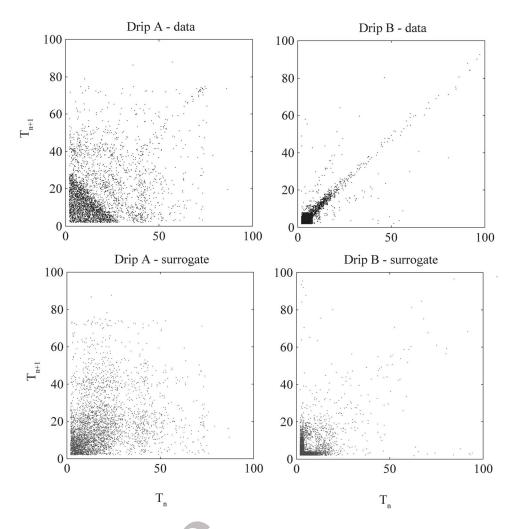


Figure 3. Joint distributions for drips (left) A and (right) B, for (top) data and (bottom) surrogates.

265 definite dripping patterns. Entropy was computed for each 266 sequence of 100 drops for both drips A and B using the 267 formulation of Shannon [1948]. Figure 4 shows these 268 entropy values plotted against the mean drip rate for each 269 100 drop sequence. Triangles correspond to drip A and 270 squares to drip B. Blue symbols represent the data and red 271 symbols the surrogates. This figure shows a consistent trend 272 of higher entropy with lower drip rates. Although both data 273 and surrogates seem to indicate a similar relationship for drip 274 A, the entropy of drip B notably deviates from the surrogate. 275 This indicates a non-linear behavior in drip B that cannot be 276 replicated by the linear process at the origin of the surrogate. [19] As a final test for chaotic behavior, we used the Lya-278 punov spectrum method of Sano and Sawada [1985] to 279 determine the maximum Lyapunov exponents of both drips 280 (data and surrogates). A positive maximal Lyaponuv expo-281 nent is generally a sign of an unstable chaotic system. For 282 drip A, the maximum exponent was found to be -0.0097 for 283 the original data and 0.0136 for the surrogate data. Values 284 close to 0 indicate that the series is quasi-periodic and that the 285 surrogate is able to reproduce such characteristics. For drip B, 286 the original data yielded a maximal Lyaponuv exponent of 287 0.9488, indicating chaos, whereas the surrogate data had an

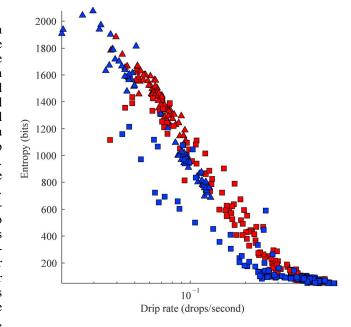


Figure 4. Entropy graph. Triangles: drip A; Squares: drip B. Blue: data; red: surrogate.

369

370

371

374

375

376

377

379

380

381

383

384

385

386

387

388

389

390

391

392

393

394

395

396

397

398

399

400

401

402

403

404

407

408

409

410

412

413

414

415

416

417

418

288 exponent of only 0.1565, indicating once again that a linear 289 process is unable to reproduce the characteristics.

290 4. Implications for Karst Hydrology 291 and Stalagmite Growth

[20] Dripping spectra, entropy of joint distributions and 293 Lyapunov exponents have been used to characterize drip 294 intervals from two percolation drips in Cathedral Cave, 295 Australia. All methods consistently indicate that the pattern 296 of drip intervals involves irregularity for both drips. While 297 drip A only shows irregularity, drip B also displays a chaotic 298 behavior, which is for the first time rigorously quantified in a 299 stalactite drip.

300 [21] In drip A, which is located slightly deeper and not 301 directly connected to the surface, irregular behavior is more 302 pronounced at the beginning of the recharge event. It can 303 be conceptualized that rainfall flows through multiple, 304 solutionally-widened fractures, which combine to supply the 305 drip. At the start of the recharge event, water flows through 306 the system from multiple sources at different rates, causing 307 high irregularity. Eventually, the fractures fill and the water 308 source becomes more homogenous, causing entropy to 309 diminish while remaining irregular. This point is reached 310 after about 1500 drops. Such varying irregularity during the 311 fractures' wetting phase is remarkable given the presence of 312 a groundwater store above the drip. However, despite caus-313 ing irregularity, the fracture wetting phenomenon does not 314 seem to lead to chaos, possibly because the resulting drip 315 rates are too low to incur the non-linear drop rebound mech-316 anism observed in leaky faucets and because of the effect of a 317 groundwater store above the drip.

[22] Drip B, which is known to be directly connected to 319 the surface water source, has a completely different behavior, 320 without initial pattern related to wetting. While it displays drip 321 rates less irregular than drip A, with a set of defined recurring 322 intervals, it is shown to be more chaotic than drip A. It is 323 hypothesized that the regularity of the drip rate allows for the 324 oscillatory effect observed in dripping faucet experiments to 325 take place. Another compounding effect may be related to the 326 dominant influence of rainfall, which has been identified to 327 exhibit chaotic behavior [Puente and Obregón, 1996].

[23] This contrasting behavior of drips located in different 329 hydrogeological settings, but subject to the same hydrologic 330 forcings, gives insights into percolation processes. One impli-331 cation of this research is that chaos can be used as a marker of 332 surface-subsurface connectivity. Observation of chaos is likely 333 to be an indication of a direct routing between surface recharge 334 and cave drip waters. We argue that high, irregular drip inter-335 vals in the absence of chaos are a sign of progressive wetting of 336 fractures, with reduced irregularity as the wetting progresses.

[24] Our work also has implications on the validity of 338 paleoclimate reconstructions based on cave stalagmites. One 339 of the controlling factors of many stalagmite proxies is the drip 340 rate feeding a stalagmite. For example, the optimal growth rate 341 occurs when the drip rate is between ~ 1 and ~ 5 minutes per 342 drip [Dreybrodt, 1999]. At very high drip rates, degassing is 343 not complete, and at slow drip rates, degassing occurs to 344 equilibrium and water supply is limiting. Kinetically-enhanced 345 degassing of CO₂ has been demonstrated to affect stalagmite 346 δ^{13} C with a dependency on drip rate and cave ventilation 347 [Baker et al., 2011], with increasing fractionation observed at 348 slower drip rates. Modeling studies of the relationship between

stalagmite δ^{13} C and δ^{18} O and drip rate provide further insights 349 into the discharge dependency [Fohlmeister et al., 2011; 350 Scholz et al., 2009]. Our results suggest that the relationship 351 between water flux and stalagmite climate proxies could be 352 more complicated than usually assumed, with strong non-lin- 353 earity occurring. One implication of this research is that drip 354 points that are well connected with surface recharge are likely 355 to exhibit chaotic behavior over an undefined but specific 356 range of flow rates. Such intermittent discharge is not captured 357 by classic time-averaged monitoring of discharge behavior. 358 Because of this variability, it follows that climate proxies 359 which show a strong growth rate effect (e.g., δ^{13} C, extension 360 rate) will have a degree of variance imposed by chaotic discharge behavior. Because natural variations occur at all scales, 362 it is probable that chaotic discharge is a second order effect on 363 proxy variance, but will nevertheless impact on the information encoded.

[25] Our work shows that drip rate averaging fails to record 366 important information about hydrogeological processes. It is, therefore, recommended for future studies to use long-term, 368 high-frequency recording of individual drip counts.

Acknowledgments. This work was supported by the Australian Research Council, the National Water Commission, and the Groundwater Education Investment Fund. We thank Mike Augee and the Wellington Caves Management Committee for kindly letting us access the Wellington site, and Matthew McCabe for installation and maintenance of the weather station. We also thank two anonymous reviewers for comments leading to improvements of the paper.

[27] The Editor thanks Carlo Camporeale for his assistance in evaluating this paper.

References

Baker, A., and C. Brunsdon (2003), Non-linearities in drip water hydrology: An example from Stump Cross Caverns, Yorkshire, J. Hydrol., 277(3-4), 151–163, doi:10.1016/\$0022-1694(03)00063-5.

Baker, A., R. Wilson, I. J. Fairchild, J. Franke, C. Spötl, D. Mattey, V. Trouet, and L. Fuller (2011), High resolution δ^{18} O and δ^{13} C records from an annually laminated Scottish stalagmite and relationship with last millennium climate, Global Planet. Change, 79(3-4), 303-311, doi:10.1016/j.gloplacha.2010.12.007.

Bradley, C., A. Baker, C. N. Jex, and M. J. Leng (2010), Hydrological uncertainties in the modelling of cave drip-water δ^{18} O and the implications for stalagmite palaeoclimate reconstructions, Quat. Sci. Rev., 29(17-18), 2201-2214, doi:10.1016/j.quascirev.2010.05.017.

Coullet, P., L. Mahadevan, and C. Riera (2005), Hydrodynamical models for the chaotic dripping faucet, J. Fluid Mech., 526, 1-17, doi:10.1017/ S0022112004002307

D'Innocenzo, A., and L. Renna (1996), Analytical solution of the dripping faucet dynamics, Phys. Lett., Sect. A, 220(1-3), 75-80.

D'Innocenzo, A., and L. Renna (1997), Modeling leaky faucet dynamics, *Phys. Rev. E*, 55(6), Suppl. A, 6776–6790, doi:10.1103/PhysRevE.55.6776. Dreybrodt, W. (1999), Chemical kinetics, speleothem growth and climate, Boreas, 28(3), 347–356, doi:10.1080/030094899422073.

Dreyer, K., and F. Hickey (1991), The route to chaos in a dripping water faucet, Am. J. Phys., 59(7), 619-627, doi:10.1119/1.16783

Fohlmeister, J., D. Scholz, B. Kromer, and A. Mangini (2011), Modelling carbon isotopes of carbonates in cave drip water, Geochim. Cosmochim. Acta, 75(18), 5219-5228, doi:10.1016/j.gca.2011.06.023.

Fuchikami, N., S. Ishioka, and K. Kiyono (1999), Simulation of a dripping faucet, J. Phys. Soc. Jpn., 68(4), 1185-1196, doi:10.1143/JPSJ.68.1185. Genty, D., and G. Deflandre (1998), Drip flow variations under a stalactite of the Père Noël cave (Belgium). Evidence of seasonal variations and air pressure constraints, J. Hydrol., 211(1-4), 208-232, doi:10.1016/S0022-

Hendy, C. H. (1971), The isotopic geochemistry of speleothems—I. The calculation of the effects of different modes of formation on the isotopic composition of speleothems and their applicability as palaeoclimatic indicators, Geochim. Cosmochim. Acta, 35(8), 801-824, doi:10.1016/ 0016-7037(71)90127-X.

Jex, C., G. Mariethoz, A. Baker, P. Graham, M. Andersen, I. Acworth, N. Edwards, and C. Azcurra (2012), Spatially dense drip hydrological monitoring at the Wellington Caves, south east Australia, Int. J. Speleol., 41(2), 283–296, doi:10.5038/1827-806X.41.2.14.

451

452

458

- 421 Johnson, B. D. (1975), The Garra Formation (Early Devonian) at Welling-422 ton, N. S. W., *J. Proc. R. Soc. N. S. W.*, 108, 111–118.
- 423 Lorenz, E. (1963), Deterministic nonperiodic flow, *J. Atmos. Sci.*, 20, 424 130–141, doi:10.1175/1520-0469(1963)020<0130:DNF>2.0.CO;2.
- Mariethoz, G., B. F. J. Kelly, and A. Baker (2012), Quantifying the value of
 laminated stalagmites for paleoclimate reconstructions, *Geophys. Res.* Lett., 39, L05407, doi:10.1029/2012GL050986.
- 428 Martien, P., S. C. Pope, P. L. Scott, and R. S. Shaw (1985), The chaotic 429 behavior of the leaky faucet, *Phys. Lett. A*, *110*(7–8), 399–404, 430 doi:10.1016/0375-9601(85)90065-9.
- 431 Miorandi, R., A. Borsato, S. Frisia, I. J. Fairchild, and D. K. Richter (2010), 432 Epikarst hydrology and implications for stalagmite capture of climate 433 changes at Grotta di Ernesto (NE Italy): Results from long-term monitor-434 ing Hydrol Processes 24(21) 3101–3114 doi:10.1002/hyp.7744
- 434 ing, Hydrol. Processes, 24(21), 3101–3114, doi:10.1002/hyp.7744.
 435 Osborne, R. A. L. (2007), Cathedral Cave, Wellington Caves, New South
 436 Wales, Australia. A multiphase, non-fluvial cave, Earth Surf. Processes
 437 Landforms, 32(14), 2075–2103, doi:10.1002/esp.1507.
- 438 Pinto, R. D., J. C. Sartorelli, and W. M. Gonçalves (2001), Homoclinic tan-439 gencies and routes to chaos in a dripping faucet experiment, *Physica A*, 440 291(1–4), 244–254, doi:10.1016/S0378-4371(00)00513-6.
- 441 Puente, C. E., and N. Obregón (1996), A deterministic geometric representation of temporal rainfall: Results for a storm in Boston, *Water Resour.* 443 *Res.*, 32(9), 2825–2839, doi:10.1029/96WR01466.
- 444 Rodriguez-Iturbe, I., B. Febres De Power, M. B. Sharifi, and K. P. Georga 445 kakos (1989), Chaos in rainfall, *Water Resour. Res.*, 25(7), 1667–1675,
 446 doi:10.1029/WR025i007p01667.

- Sano, M., and Y. Sawada (1985), Measurement of the Lyapunov spectrum from a chaotic time series, *Phys. Rev. Lett.*, 55(10), 1082–1085, 448 doi:10.1103/PhysRevLett.55.1082.
- Sartorelli, J. C., W. M. Gonçalves, and R. D. Pinto (1994), Crisis and intermittence in a leaky-faucet experiment, *Phys. Rev. E*, 49(5), 3963–3975, doi:10.1103/PhysRevE.49.3963.
- Scholz, D., C. Mühlinghaus, and A. Mangini (2009), Modelling δ^{13} C and δ^{18} O in the solution layer on stalagmite surfaces, *Geochim. Cosmochim.* 454 *Acta*, 73(9), 2592–2602, doi:10.1016/j.gca.2009.02.015.
- 456 Schreiber, T., and A. Schmitz. (1996), Improved surrogate data for nonlinearity 456 tests, *Phys. Rev. Lett.*, 77(4), 635–638, doi:10.1103/PhysRevLett.77.635.
- Shannon, C. (1948), A mathematical theory of communication, *Bell Syst. Tech. J.*, 27, 379–423.
- Sivakumar, B. (2004), Chaos theory in geophysics: Past, present and future, 460 Chaos Solitons Fractals, 19(2), 441–462, doi:10.1016/S0960-0779(03) 461 00055-9. 462
- Spötl, C., I. J. Fairchild, and A. F. Tooth (2005), Cave air control on dripwater geochemistry, Obir Caves (Austria): Implications for speleothem deposition in dynamically ventilated caves, *Geochim. Cosmochim. Acta*, 465 (90), 2451–2468, doi:10.1016/j.gca.2004.12.009.
- Theiler, J., S. Eubank, A. Longtin, B. Galdrikian, and J. Doyne Farmer (1992), Testing for nonlinearity in time series: The method of surrogate data, *Physica D*, 58(1–4), 77–94, doi:10.1016/0167-2789(92)90102-S.
 469