

## Record hot years in autocorrelated series

We want to know the probability that among 129 annual temperature readings, at least 15 of the hottest years appear in the last 19 places (1990-2008), much like they did in the reality

(\* if there were cooling, we could be equally concerned, so we will flip the sign of a series to guarantee that there was a net warming in those 129 years.

If you're interested in the condition requiring strict "hottest" years in the recent decades, discarding the cases of "coldest" years, you should approximately divide the resulting probability by two. \*)

## Run Evaluation / Evaluate notebook

```
planets = 5000; (* how many
random sequences we analyze *)

(* Generating a random sequence
of 129 temperature readings *)
(* First, the universal constants *)

{t, recordYearsLookedFor,
 whereAllowed, damping} =
{127, 13, 17, 0.99}
(* the total number of years;
how many record breakers we looked
for; and in how many recent years
they are required to appear *)

(* Von Storch et al. have
{127,13,17} but two years later,
you can change it to {129,15,19} *)
```

```
alarmistPlanets = 0;
(* it counts the planets where
the condition is satisfied *)

For [j = 1, j ≤ planets, j++,

a = Table[0, {i, t}]; b = a; c = a;
(* a[[1]] is 1880, assumed to have
temperature anomaly of zero *)

For [i = 2, i ≤ t, i++,
a[[i]] = a[[i - 1]] +
(RandomReal[] - 0.5) * 1;
(* This adds a Brownian motion-
like noise to the previous year,
to derive the new temperature.
The overall normalization,
determined by "times one" at
the end, doesn't influence
```

the statistics of records but  
it is useful to adjust the  
random series to resemble the  
temperature record in deg C \*)

```
a[[i]] = a[[i]] * damping;
```

```
(* But the temperature anomaly is  
shrunk by a factor to avoid  
sqrt(time)-like Brownian  
runaway behavior. Physically,  
this line gives the climate  
the effects that stabilize the  
temperature if it deviates too  
much. Any factor below 1 keeps  
the temperature oscillating  
in a finite range around zero.  
Actually, 0.99 is completely  
tolerable and doesn't increase  
the standard deviation of  
temperature over very long
```

time scales too much...

In fact, the long-term standard deviation of temperature seems to scale like  $1/\sqrt{1-\text{damping}}$ , as I determined by games below the memo.

It follows that for 129 years, damping can be so close to one that we may pretty much say that the temperature dynamics is essentially Brownian motion \*)

```
];
(* Dynamic[
  ListLinePlot[a,Filling→Axis]] *)

If[a[[t]] < 0, a = -a, a = a];
recordYearIDs = Ordering[
```

```
-a, recordYearsLookedFor];  
yesNoRecent = If[# > t - whereAllowed,  
  1, 0] & /@ recordYearIDs;  
howManyRecent = Total[yesNoRecent];  
alarmistPlanets =  
  alarmistPlanets + If[howManyRecent ==  
    recordYearsLookedFor, 1, 0];  
];  
  
{alarmistPlanets,  
  N[alarmistPlanets / planets]}  
(* probability that we get the  
  desired recent record breakers *)
```

```
{127, 13, 17, 0.99}
```

```
{329, 0.0658}
```

(\* For 129 years and 15 record breakers required to fall into the most recent 19 years, the probability is around 6%.

For 127 years and 13 record breakers required to fall to the most recent 15 years, the probability is also close to 6%.

This is for damping=0.99 which is a realistic value (the standard deviation is about 10 times larger than the maximum random increments). For damping at 0.999 or higher (which is actually claimed to be relevant in the newer, normally distributed notebook), both figures get close to 10% or slightly above. Further increases of damping towards 1 have almost no effect.

Adding non-Brownian, random noise (without inertia), like the noise observed at local stations or (partly) regional scales, would clearly make the years more independent, and it would reduce the predicted probability that the 13 or so record breakers appear recently. \*)

Between 5% and 10% of the planets observe 13 record-breaking years (either hottest or coldest) in the most recent 17 years out of 127 on their record.

And that's the memo.

```
(* What follows are various tests to learn
how the parameters influence the series *)

{t, damping} = {127000, 0.99}
a = Table[0, {i, t}]; b = a; c = a; (* a[[1]] is 1880,
assumed to have temperature anomaly of zero *)

For [i = 2, i ≤ t, i++,
  a[[i]] = a[[i - 1]] + (RandomReal[] - 0.5) * 1;
  (* This adds a Brownian motion-
  like noise to the previous year,
  to derive the new temperature. The overall normalization,
  determined by "times one" at the end,
  doesn't influence the statistics of records
  but it is useful to adjust the random series
  to resemble the temperature record in deg C *)

  a[[i]] = a[[i]] * damping;
  (* But the temperature anomaly is shrunk by a factor to avoid
  sqrt(time)-like Brownian runaway behavior. Physically,
  this line gives the climate the effects that
  stabilize the temperature if it deviates too
  much. Any factor below 1 keeps the temperature
  oscillating in a finite range around zero.
  Actually, 0.999 is tolerable and doesn't
  increase the standard deviation of temperature
  over very long time scales too much... *)

];
(* ListPlot[a] *)
{ "Standard deviation", Sqrt[ Total[a^2] / t ] ,
  "Theoretical formula for the SD", 0.2 / Sqrt[1 - damping] }

{127000, 0.99}
```

```
{Standard deviation, 2.0533, Theoretical formula for the SD, 2.}
```

```
(* Yes, I tried to change damping from 0.999 to 0.99999,  
i.e. the deviation from 1 decreases 100 times,  
and the standard deviation for a long-term dynamics,  
determined by "t" of order tens of thousands of years,  
only grew about 10 times. Try it. *)
```