Modeling beyond randomness

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College of William and Mary UMSA 2002 Let's go back to the 1-d drunkard's walk that we saw earlier this semester Assume:

- the drunkard tends to go left with probability 45% and right with 55%
- there are 60 different discrete points on the 1-D road
- at the two ends of the road the drunkard still chooses a direction
 - if the direction is within the road the drunkard goes for it
 - if it is outside the road the drunkard stays at the same point

Question: After infinitely many steps what is the probability of the drunkard being at a specific point (state)?

A classical Markov chain problem

Here is a simple (simplistic?) way of simulating the walk in matlab:

```
function drunk1d_rand(sample,states,time)
middle = ceil(states/2);
drunkards = middle*ones(sample,1);
State_plot = zeros(states,1);
State_plot(middle) = sample;
for t = 1:time
   for i = 1:sample
        coin = floor(rand(1,1) + 0.55);
        State_plot(drunkards(i)) = State_plot(drunkards(i)) - 1;
        if (coin == 0)
          if (drunkards(i) > 1)
            drunkards(i) = drunkards(i) - 1;
          end
        else
          if (drunkards(i) < states)</pre>
            drunkards(i) = drunkards(i) + 1;
          end
        end
        State_plot(drunkards(i)) = State_plot(drunkards(i)) + 1;
   end
end;
plot( State_plot );
```

- Easy to implement
- Relatively easy to enhance with additional complications of the model
- BUT Converges slowly !!

Question: How can we model this problem so we obtain analytic solutions?

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Assume a set of *N* states s_1, \ldots, s_N .

Let the drunkard be at state s_j

Then the drunkard can choose to go to s_i with probability $p_{i,j}$, with $\sum_{i=1,N} p_{i,j} = 1$

Let's accumulate these column probabilities for each state in a matrix *A*. *A* is called Probability matrix and note that:

 $[11\cdots 1]*A = [11\cdots 1]$

For our drunkard example our matrix looks like this:

Notice now that we look for the steady state probability, i.e., the probability that does not change from step to step (at infinity):

 $Prob(s_i) = Prob(s_{i-1}) * .55 + Prob(s_{i+1}) * .45$

or equivalently, we want the eigenvector *p* with eigenvalue 1:

Ap = p

This is a 60×60 problem and can be solved in a jiffy numerically

However, note this is only for the steady state

Consider the classical 1-D decceleration problem:

At time zero a rocket is fired from the ground upwards and after T seconds the rocket hits back the ground. What is the height of the rocket at any time in the interval [0, T]?

(For simplicity assume constant gravity acceleration = -10 m/sec^2 and T = 1)

This physical process can be described with a PDE involving the height s(t):

$$\frac{d^2s(t)}{dt^2} = -10$$

with the boundary conditions s(0) = s(T) = 0.

Analytic solution is trivial, but not for even slightly more complicated PDEs

Assume we want the solution only at a set of time points t_i , i = 0, ..., N + 1

Assume equally spaced $t_i = i * h$ with h = 1/(N+1) and $t_0 = 0, t_N = 1$

More accuracy can be obtained by increasing N

Since s(t) continuous and doubly differentiable use Taylor series approximation

$$\frac{ds}{dt}(t) = \frac{s(t+h) - s(t)}{h} + \frac{h}{2}\frac{d^2s(t)}{dt^2} + O(h^2)$$

Write the same based on the previous point approximation t - h and substra

$$\frac{ds}{dt}(t) = -\frac{s(t-h) - s(t)}{h} - \frac{h}{2}\frac{d^2s(t)}{dt^2} + O(h^2)$$

Substract the two equations:

$$\frac{d^2s(t)}{dt^2} = \frac{s(t+h) - 2s(t) + s(t-h)}{h^2} + O(h^2)$$

We look for $s(t_i) \equiv s_i, i = 1, ..., N$. Then the formula becomes

$$\frac{d^2 s(t)}{dt^2} \approx \frac{1}{h^2} (s_{i-1} - 2s_i + s_{i+1})$$

and our equations become:

$$s_{i-1} - 2s_i + s_{i+1} = -h^2 10, \quad i = 1, \dots, N$$

This is just a set of linear equations that can be represented in matrix form