# STAT4404: Advanced Stochastic Processes II, Semester 1, 2013. Quiz 3 (with solution)

Let  $\{B_1(t)\}\$  and  $\{B_2(t)\}\$  be two independent standard Brownian motion processes defined on the same probability space.

Denote,

$$A(n) = \frac{B_1(n) - B_2(n)}{n^{\alpha}}, \quad n = 1, 2, \dots, \quad \alpha \ge 0$$

## Problem 1:

Denote,

$$A_0(n) = A(n) - nA(100), \quad n = 0, 1, 2, \dots, 100.$$

Find  $\mathbb{E} [A_0(n)]$  and  $\operatorname{Var}(A_0(n))$ .

### Problem 2:

Describe the convergence of  $\{A(n), n = 0, 1, 2, ...\}$ , prove your findings. That is find when,

$$A(n) \to Y,$$

where the convergence is in one of several ways (almost sure, probability, r'th mean, or distribution). Your answer may depend on  $\alpha$ . The element Y may either a constant, a non-degenerate random variable or infinity.

You can assume any theorem stated or proved in class as known.

## Solution (for both problems):

We have,

$$\begin{aligned} A(n) &= n^{-\alpha} \left( B_1(n) - B_2(n) \right) \\ &= n^{-\alpha} \left( \sum_{i=1}^n \left( B_1(i) - B_1(i-1) \right) - \sum_{i=1}^n \left( B_2(i) - B_2(i-1) \right) \right) \\ &= n^{-\alpha} \sum_{i=1}^n \Delta_1(i) - \Delta_2(i) \\ &= n^{-\alpha} \sum_{i=1}^n \tilde{\Delta}(i), \end{aligned}$$

where for k = 1, 2 and i = 1, ..., n,

$$\Delta_k(i) := B_k(i) - B_k(i-1) \qquad \sim N(0,1) \quad \text{i.i.d.},$$

and thus,

$$\tilde{\Delta}(i) := \Delta_1(i) - \Delta_2(i) \qquad \sim N(0,2) \quad \text{i.i.d.}.$$

So we have,

$$A(n) \sim N(0, (n^{-\alpha})^2 n^2) = N(0, 2n^{1-2\alpha}).$$

Thus observe that for  $\alpha < \frac{1}{2}$ , Var(A(n)) is increasing while for  $\alpha > \frac{1}{2}$  it is decreasing. For  $\alpha = \frac{1}{2}$ , Var(A(n)) = 2 for all n.

### Problem 1:

First, Showing  $E[A_0(n)] = 0$  is trivial, even though, A(n) and A(100) are not independent, we just have,

$$E[A_0(n)] = E[A(n)] - nE[A(100)] = 0 - n0 = 0.$$

To compute the variance, break up to independent increments (in the standard way):

$$\begin{aligned} A_0(n) &= A(n) - nA(100) \\ &= n^{-\alpha} \sum_{i=1}^n \tilde{\Delta}(i) - \frac{n}{100^{\alpha}} \sum_{i=1}^n \tilde{\Delta}(i) \\ &= n^{-\alpha} \sum_{i=1}^n \tilde{\Delta}(i) - \frac{n}{100^{\alpha}} \Big( \sum_{i=1}^n \tilde{\Delta}(i) + \sum_{i=n+1}^{100} \tilde{\Delta}(i) \Big) \\ &= \Big( n^{-\alpha} - \frac{n}{100^{\alpha}} \Big) A(n) - \frac{n}{100^{\alpha}} \Big( A(100) - A(n) \Big). \end{aligned}$$

Now A(n) and the increment A(100) - A(n) are independent, so take variance and rearrange to obtain:

$$Var(A_0(n)) = 2n^{1-2\alpha} - \frac{4}{100^{\alpha}}n^{2-\alpha} + \frac{2}{100^{2\alpha-1}}n^2.$$

## Problem 2:

It was only required to give an answer of the form below (or more detailed). A complete characterization of when and how A(n) converges (based on  $\alpha$ ), is possible, yet was not required in the given time frame.

- First observe that for  $\alpha < \frac{1}{2}$ , Var(A(n)) is increasing and thus it does not converge in any of the modes to any RV.
- Now for  $\alpha = 1/2$  the distribution of A(n) is fixed as N(0,2) thus it converges in distribution to N(0,2).
- For  $\alpha > 1/2$  we have by Chebyshev's inequality that A(n) converges in probability to 0. Similarly we have convergence in the 2'nd mean to 0.
- For  $\alpha = 1$  we have by the SLLN that A(n) converges to 0.