

1st Variation of Brownian Motion

- ▶ We have already shown that sample paths of $W(t)$ are nowhere differentiable.
- ▶ The first variation does not make sense because of the above reason.
- ▶ Hence, 1st variation does not exist for Brownian motion.



Now we are going to discuss the first variation of Brownian motion. We have already shown that the sample path of W_t are nowhere differential. Therefore the first order variation does not make sense because of the above reason because the derivative it is nowhere differential therefore you cannot get the first order variation. Hence, the first order variation of the Brownian motion does not exist.

Quadratic Variation of Brownian Motion

- ▶ The quadratic variation for Brownian motion over the interval $[0, T]$, denoted by $[W(t), W(t)](T)$, is given by:

$$[W(t), W(t)](T) = V_{W(t)}^2(T) = \lim_{\|\pi\| \rightarrow 0} Q_\pi$$

where

$$Q_\pi = \sum_{i=0}^{n-1} (W(t_{i+1}) - W(t_i))^2$$

- ▶ Clearly, Q_π is a function of the sample points $w \in \Omega$.
- ▶ Hence, the quadratic variation calculated for a Brownian motion for each partition itself a random variable.
- ▶ Note that, the limit is taken over all partitions of $[0, T]$, with $\|\pi\| \rightarrow 0$ as $n \rightarrow \infty$.



Now we are moving into the quadratic variation of Brownian motion. The quadratic variation of Brownian motion over the interval 0 to t where t is a positive real number denoted by the notation W_t , W_t of t that is given by V suffix t is wrong notation it's a V suffix W_t superscript 2 of t that is nothing but limit Π tends to 0 of Q_{Π} where Q_{Π} is defined summation i is equal to 0 to n minus 1 the difference of W 's and the time pointer t_i to t_{i+1} the whole square; clearly because you are making a difference of W 's so the Q_{Π} is a function of a sample points of W belonging to Ω , and also hence the quadratic variation calculated for the Brownian motion for each part is in itself a random variable because this is a random variable, the difference is a random variable, the summation will be sum of random variables is a random variable therefore the Q_{Π} is a random variable and you are finding limit Π tends to norm of Π tends to 0 of Q_{Π} . That is nothing but note that this limit taken over all partition of 0 to Π with norm of Π tends to 0 as n tends to infinity. Norm of Π is defined as the maximum of i of the length of the interval $t_{i+1} - t_i$ therefore norm of Π tends to 0 means you are finding the limit is taken over all partitions of 0 to Π , 0 to t.

So we have to find out what is a limit norm Π tends to 0 of this random variable. For every n this will be a random variable so you have to find out the limit taken over all partitions of 0 to T with norm of Π tends to 0 as n tends to infinity.

Quadratic Variation of Brownian Motion...

Since, for each partition π , Q_{π} is a random variable, how to find the limiting distribution of Q_{π} for large n ? The question would be what is the proper mode of convergence in these random variables. We shall use convergence in mean square (convergence in L^2). The definition of convergence in L^2 is as follows:

Definition
Let $\{X_n, n \geq 1\}$ and X be random variables defined on a common probability space (Ω, \mathcal{F}, P) . We say that X_n converges in L^2 to X if

$$\lim_{n \rightarrow \infty} E(|X_n - X|^2) = 0.$$

Since for each Π for each partition Π the Q_{Π} is a random variable how to find the limiting distribution of Q_{Π} for a large n ? The question would be what is the proper mode of convergence in this random variables? We shall use the convergence in mean square that is a convergence in L^2 to find the limit of norm Π tends 0 Q of Π as n tends to infinity. So for that we are going define the convergence in L^2 . Let Π_n , let X of n , n is greater than or equal to 1 and X random variables defined on a common probability space Ω, \mathcal{F}, P . We say that X_n converges in L^2 to

the random variable X if limit n tends to infinity expectation of the absolute of X_n minus X whole square is equal to 0. So if this condition is satisfied this is a sequence of random variable and this is a random variable both are defined in the same probability space Ω, \mathcal{F}, P then we say the sequence X_n convergence to the random variable X in L^2 .

So this the same approach we are going to use to find out that limiting distribution of the random variables Q of P_i for a large n as n tends to infinity.

Quadratic Variation of Brownian Motion...

In the case of Brownian motion, we will show that

$$\lim_{\|\pi\| \rightarrow 0} E(|Q_\pi - T|^2) = 0 \quad (1)$$

When the above result holds good, we say that the quadratic variation accumulated by the Brownian motion over the interval $[0, T]$ is T in mean square and is denoted as $[W, W](T) = T$

Let us prove the above result in the following theorem.

Theorem

- 1 $E(Q_\pi) = T$
- 2 $Var(Q_\pi) \leq 2 \|\pi\| T$
- 3 $E(Q_\pi - T)^2 = Var(Q_\pi)$



In this case of Brownian motion we will show that the limit norm P_i tends to 0 expectation of absolute of $P_i Q$ sorry Q_{P_i} minus t whole square is equal to 0. That means the sequence of random variable Q of P_i as n tends to infinity converges to the random variable which is a constant to T in L^2 . If this condition is since this condition is satisfied. When the above results hold good we say that the quadratic variation accumulates accumulated by the Brownian motion over the interval $0, T$ is T in mean square and it is denoted W of W of the interval 0 to T that is T . So to prove the sequence of random variable Q_{P_i} converges to the random variable T as n tends to infinity in L^2 we prove it in a 3 stage. The first stage we find out we will prove that expectation of Q of P_i is equal to T . Then we will prove the variance of Q of P_i that is less than or equal to two times norm of P_i of T . Therefore you can prove the final result expectation of Q_{P_i} minus T whole square that is nothing but variance of Q_{P_i} because the expectation of Q_{P_i} is T therefore expectation of Q_{P_i} minus T whole square is variance of Q_{P_i} so as n tends to infinity like Q_{P_i} will converges to the random variable T in L^2 .

Quadratic Variation of Brownian Motion . . .

Proof:

$$E(Q_n) = \sum_{i=0}^{n-1} E[W(t_{i+1}) - W(t_i)]^2$$

Since, for fixed i , $W(t_{i+1}) - W(t_i)$ is normal distribution with mean 0 and variance $(t_{i+1} - t_i)$

$$= \sum_{i=0}^{n-1} (t_{i+1} - t_i) = T$$

$$\text{Var}(Q_n) = \sum_{i=0}^{n-1} \text{Var}[W(t_{i+1}) - W(t_i)]^2$$

$$\begin{aligned} \text{Var}[W(t_{i+1}) - W(t_i)]^2 &= E[W(t_{i+1}) - W(t_i)]^4 \\ &\quad - 2E[W(t_{i+1}) - W(t_i)]^2(t_{i+1} - t_i) \\ &\quad + (t_{i+1} - t_i)^2 \end{aligned}$$



The proof as follows first we will find out the expectation of Q_n that is nothing but the summation of i is equal to 0 to $n-1$ expectation of W of t_{i+1} minus W of t_i the whole square. Since for fixed i the difference of the W 's is normally distributed random variable with the mean 0 and the variance is nothing but the length of the interval. Therefore, the expectation of the difference of random variable whole square is nothing but the variance therefore our fixed i that is nothing but the T of t_{i+1} minus t_i . The summation is varies from i is equal to 0 to $n-1$ therefore you will get T . So the first part is proved that is expectation of Q_n is equal to T . Now we will find out the variance of Q_n that is less than or equal to we have to prove the second part variance of Q_n is less than or equal to two times norm of P_i multiplied by t . third part is quite easy.

Quadratic Variation of Brownian Motion...

In the case of Brownian motion, we will show that

$$\lim_{\|\pi\| \rightarrow 0} E(|Q_\pi - T|^2) = 0 \quad (1)$$

When the above result holds good, we say that the quadratic variation accumulated by the Brownian motion over the interval $[0, T]$ is T almost surely and is denoted as $[W, W](T) = T$. Let us prove the above result in the following theorem.

Theorem

- 1 $E(Q_\pi) = T$
- 2 $\text{Var}(Q_\pi) \leq 2 \|\pi\| T$
- 3 $E(Q_\pi - T)^2 = \text{Var}(Q_\pi)$



So the variance of Q_π is nothing but summation i is equal to 0 to n minus 1 variance of the difference of random variable whole square but variance of difference of the random variable whole square that is nothing but the expectation of difference of random variable whole power 4 minus 2 times expectation of the difference of the random variable whole square multiplied by T_i plus 1 minus T_a plus a T_i plus 1 minus T_a whole square for fixed i . Using the fourth order moment of normally distributed random variable with the mean 0 and the variance T_i plus 1 minus t minus 1, the first term in the right hand side expectation of difference of the random variable power 4 that is the fourth order moment about the fourth order momenta that is nothing but at 3 times T_i plus 1 minus T_i whole square. Therefore, the right hand side variance of the difference of the random variable whole square that is nothing but three times the difference the time difference whole square minus 2 times the time difference whole square plus time difference whole square. So therefore this is nothing but 2 times time difference whole square.

Quadratic Variation of Brownian Motion...

Then, we get

$$[W, W](T) = \lim_{\|\pi\| \rightarrow 0} Q_\pi = T$$

since

$$\lim_{\|\pi\| \rightarrow 0} E((Q_\pi - T)^2) = 0$$

Hence $[W, W](T) = T$. Also, for $0 < T_1 < T_2$, $[W, W](T_2) - [W, W](T_1) = T_2 - T_1$, i.e., the Brownian motion accumulates $T_2 - T_1$ units of quadratic variation over the interval T_1 to T_2 .

Informally we can write as $dW(t) dW(t) = dt$ and this dt is in fact $1 \cdot dt$.



The 2 times a time difference whole square is nothing but that is less than or equal to 2 times norm of Π multiplied by the time difference. Therefore, the variance of Q_{Π} is less than or equal to two times norm of Π times T . Therefore since you know that the expectation of Q_{Π} is equal to T therefore expectation of norm of Q_{Π} minus T whole square that is nothing but variance of Q_{Π} . Therefore the second order variation of the Brownian motion W_t between the powers interval 0 to T is nothing but limiter norm of Π tends to 0, Q_{Π} that is same as T . Since limit norm Π tends to 0 the expectation of Q_{Π} minus T whole square is equal to 0. So the conclusion is the second order or the quadratic variation of a Brownian motion is T between the interval 0 to capital T . This means it accumulates a unit quadratic variation per unit. Also for 0 less than T_1 less than T_2 the quadratic variation till T_2 the quadratic variation, sorry, the quadratic variation till T_2 minus quadratic variation till T_1 that is same as T_2 minus T_1 . That is the Brownian motion accumulates T_2 minus T_1 units of quadratic variation over the interval T_1 to T_2 . Since this is true for every interval we refer that the Brownian motion accumulates quadratic variation at the rate 1 per unit. This last statement we write informally as $dW_t dW_t$ is equal to dt and this dt is in fact 1 times dt . In other words the above phenomena can be represented in a differential form as a differential of W_t multiplied with a differential of W_t . This is a quadratic relation that is nothing but the differential of t , that is a meaning of the Brownian motion accumulates unit quadratic variation per unit time.

Quadratic Variation of Brownian Motion . . .

By applying Theorem 1, we have

$$V_W^p(T) = \begin{cases} \infty, & p = 1 \\ T, & p = 2 \\ 0, & p > 2 \end{cases}$$

This concludes that, the Brownian motion $\{W(t), t \geq 0\}$ is of unbounded variation and finite quadratic variation for every t .



Now we are applying the same theorem to discuss for the variations of the real valued function we have the p th order variance of Brownian motion between the interval 0 to T that value will be does not exist for the first order variation therefore P is equal to 1 it is infinity. For the quadratic variation it is a T . It's a bounded variation of that a quadratic variation whereas the first order is unbounded variation. For P greater than 2 it is 0. So the example we have taken is G of t is equal to t Square for that the first order variance is finite and further variations are 0 whereas for the Brownian motion the first order variation is a infinity that is unbounded variation and the second order variation is a finite value that is T and the further variations are 0. This concludes that the Brownian motion is of unbounded variation because P is equal to 1 the variation is infinity and the finite quadratic variation because P is equal to 2 value will be T for every t .