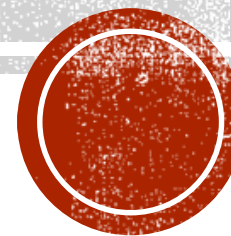


# 3.7 REFLECTION PRINCIPLE

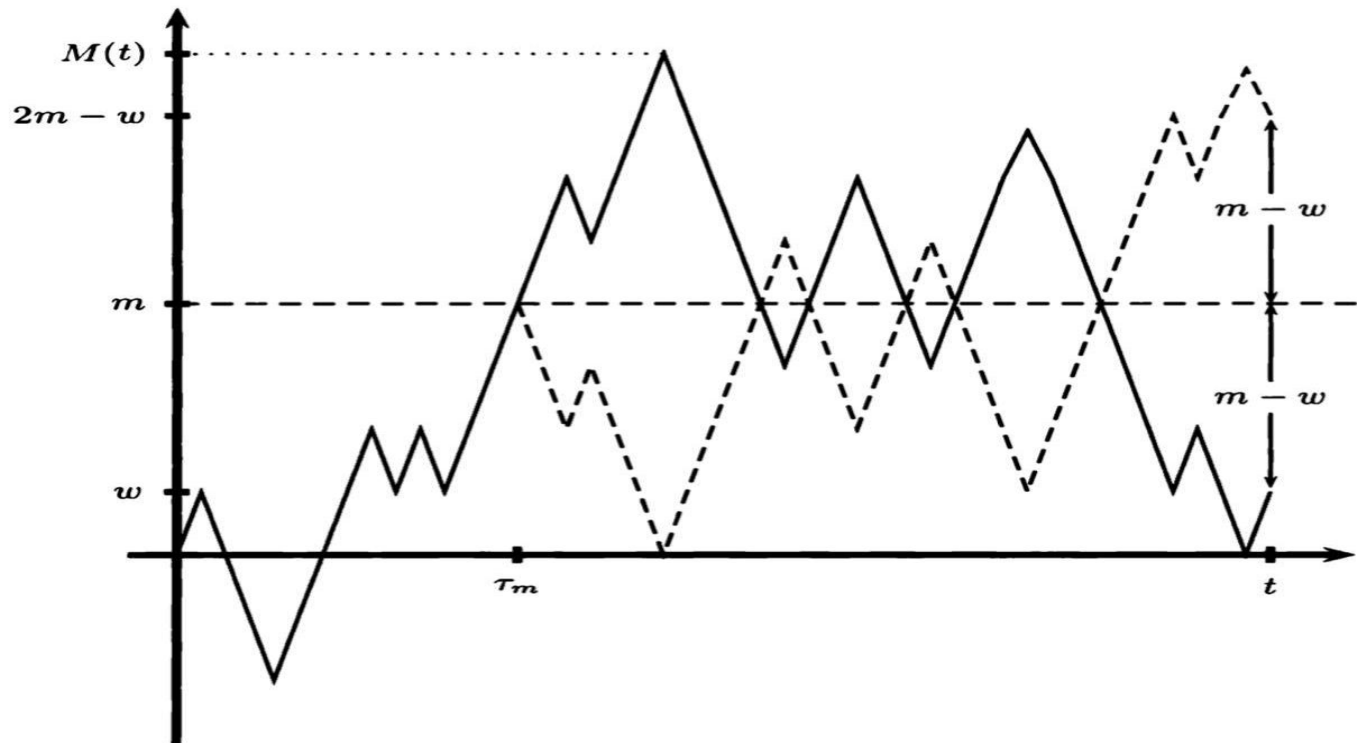
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- **3.7.1 Reflection Equality**
- **3.7.2 First Passage Time Distribution**
- **3.7.3 Distribution of Brownian Motion and Its Maximum**



# 3.7.1 Reflection Equality



$$p\{\tau_m \leq t, W(t) \leq w\} = p\{W(t) \geq 2m - w\}, \quad w \leq m, m > 0 \quad (3.7.1)$$



## 3.7.2 First Passage Time Distribution

- Thm 3.7.1

For all  $m \neq 0$ , the random variable  $\tau_m$  has cumulative distribution function

$$p\{\tau_m \leq t\} = \frac{2}{\sqrt{2\pi}} \int_{\frac{|m|}{\sqrt{t}}}^{\infty} e^{-\frac{y^2}{2}} dy, \quad t \geq 0$$

(3.7.2)

And density

$$f_{\tau_m}(t) = \frac{d}{dt} p\{\tau_m \leq t\} = \frac{|m|}{t\sqrt{2\pi t}} e^{-\frac{m^2}{2t}}, \quad t \geq 0$$

(3.7.3)



## ■ Proof of Thm 3.7.1

- We first consider the case  $m > 0$ . We substitute  $w = m$  into the reflection formula (3.7.1) to obtain

$$p\{\tau_m \leq t, W(t) \leq w\} = p\{\tau_m \leq t, W(t) \leq m\} = p\{W(t) \geq m\}$$

If  $W(t) \geq m$ , then we guaranteed that  $\tau_m \leq t$ , so

$$p\{\tau_m \leq t, W(t) \geq w\} = p\{\tau_m \leq t, W(t) \geq m\} = p\{W(t) \geq m\}$$

We obtain

$$\begin{aligned} p\{\tau_m \leq t\} &= p\{\tau_m \leq t, W(t) \leq m\} + p\{\tau_m \leq t, W(t) \geq m\} \\ &= 2p\{W(t) \geq m\} = \frac{2}{\sqrt{2\pi t}} \int_m^{\infty} e^{-\frac{x^2}{2t}} dx \end{aligned}$$



$$p\{\tau_m \leq t\} = p\{\tau_m \leq t, W(t) \leq m\} + p\{\tau_m \leq t, W(t) \geq m\}$$

$$= 2p\{W(t) \geq m\}$$

$$= \frac{2}{\sqrt{2\pi t}} \int_m^\infty e^{-\frac{x^2}{2t}} dx$$

Let  $y = \frac{x}{\sqrt{t}}$ ,  $dy = \frac{dx}{\sqrt{t}}$

$$= \frac{2}{\sqrt{2\pi}} \int_{\frac{m}{\sqrt{t}}}^\infty \frac{1}{\sqrt{t}} e^{-\frac{y^2}{2}} dy$$

$$= \frac{2}{\sqrt{2\pi}} \int_{\frac{|m|}{\sqrt{t}}}^\infty e^{-\frac{y^2}{2}} dy, \quad t \geq 0$$

leads to (3.7.2)



- If  $m < 0$ ,

$\tau_m$  and  $\tau_{|m|}$  have the same distribution, and (3.7.2) provides the c.d.f of the latter.

$$p\{\tau_m \leq t\} = p\{\tau_{|m|} \leq t\} = \frac{2}{\sqrt{2\pi}} \int_{\frac{|m|}{\sqrt{t}}}^{\infty} e^{-\frac{y^2}{2}} dy, \quad t \geq 0 \quad (3.7.2)$$

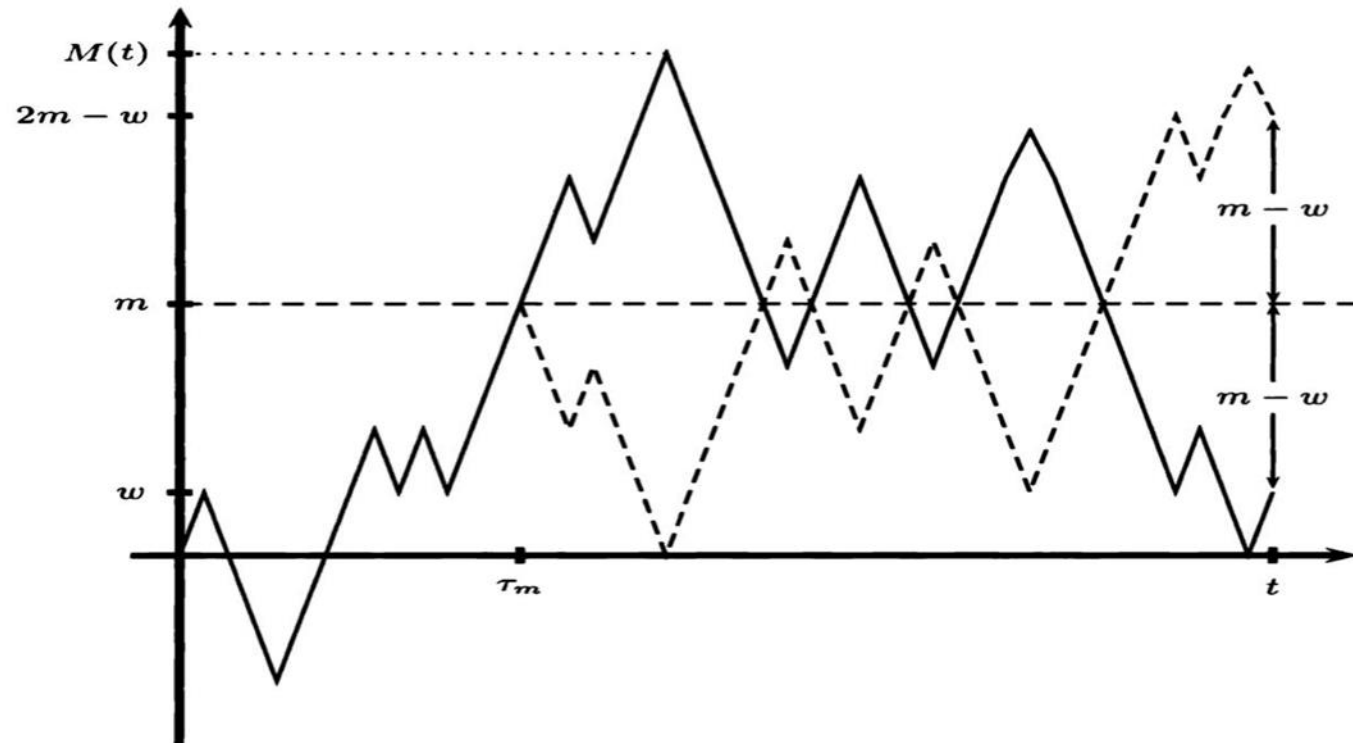
- Finally, (3.7.3) is obtained by differentiating (3.7.2) with respect to  $t$ .

$$\begin{aligned} f_{\tau_m}(t) &= \frac{d}{dt} p\{\tau_m \leq t\} = \frac{d}{dt} \left\{ \frac{2}{\sqrt{2\pi}} \int_{\frac{|m|}{\sqrt{t}}}^{\infty} e^{-\frac{y^2}{2}} dy \right\} \\ &= \frac{d}{dy} \cdot \frac{dy}{dt} \left\{ \frac{2}{\sqrt{2\pi}} \int_{\frac{|m|}{\sqrt{t}}}^{\infty} e^{-\frac{y^2}{2}} dy \right\} \\ &= \frac{dy}{dt} \left\{ -\frac{2}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} \right\}, \text{代入 } y = \frac{|m|}{\sqrt{t}} \\ &= -\frac{2}{\sqrt{2\pi}} e^{-\frac{m^2}{2t}} \cdot -\frac{|m|}{2t\sqrt{t}} \\ &= \frac{|m|}{t\sqrt{2\pi t}} e^{-\frac{m^2}{2t}}, \quad t \geq 0 \end{aligned}$$



# 3.7.3 Distribution of Brownian Motion and Its Maximum

■ We define



$$(3.7.5)$$

For positive

$p\{M$

$> 0.$

$$(3.7.6)$$





# The joint distribution of $W(t)$ and $M(t)$

- Thm 3.7.3

For  $t > 0$ , the joint density of  $(M(t), W(t))$  is

$$f_{M(t), W(t)}(m, \omega) = \frac{2(2m - \omega)}{t\sqrt{2\pi t}} e^{-\frac{(2m - \omega)^2}{2t}}, \omega \leq m, m > 0. \quad (3.7.7)$$



▪ Proof

Because

$$p\{M(t) \geq m, W(t) \leq w\} = \int_m^\infty \int_{-\infty}^w f_{M(t),W(t)}(x, y) dy dx$$

and

$$p\{W(t) \geq 2m - w\} = \frac{1}{\sqrt{2\pi t}} \int_{2m-w}^\infty e^{-\frac{z^2}{2t}} dz$$

Since  $p\{\tau_m \leq t, W(t) \leq w\} = p\{W(t) \geq 2m - w\}$ ,  $w \leq m, m > 0$  (3.7.1)

We have

$$\int_m^\infty \int_{-\infty}^w f_{M(t),W(t)}(x, y) dy dx = \frac{1}{\sqrt{2\pi t}} \int_{2m-w}^\infty e^{-\frac{z^2}{2t}} dz$$



$$\int_m^\infty \int_{-\infty}^w f_{M(t),W(t)}(x, y) dy dx = \frac{1}{\sqrt{2\pi t}} \int_{2m-w}^\infty e^{-\frac{z^2}{2t}} dz$$

We differentiate first with respect to **m** to obtain

$$-\int_{-\infty}^w f_{M(t),W(t)}(x, y) dy = -\frac{2}{\sqrt{2\pi t}} e^{-\frac{(2m-w)^2}{2t}}$$

We next differentiate with respect to **w** to see that

$$-f_{M(t),W(t)}(x, y) dy = -\frac{2(2m-w)}{\sqrt{2\pi t}} e^{-\frac{(2m-w)^2}{2t}}$$

$$f_{M(t),W(t)}(x, y) dy = \frac{2(2m-w)}{\sqrt{2\pi t}} e^{-\frac{(2m-w)^2}{2t}}$$



### Corollary 3.7.4

- The conditional distribution of  $M(t)$  given is  $W(t) = w$

$$f_{M(t)|W(t)}(m|w) = \frac{2(2m - w)}{t} e^{-\frac{2m(m-w)}{t}}, \quad w \leq m, m > 0.$$

### Proof

$$\begin{aligned} f_{M(t)|W(t)}(m|w) &= \frac{f_{M(t),W(t)}(m,w)}{f_{W(t)}(w)} \\ &= \left( \frac{2(2m-w)}{t\sqrt{2\pi t}} e^{-\frac{(2m-w)^2}{2t}} \right) / \left( \frac{1}{\sqrt{2\pi t}} e^{-\frac{w^2}{2t}} \right) \\ &= \frac{2(2m-w)}{t\sqrt{2\pi t}} \cdot \sqrt{2\pi t} \cdot e^{-\frac{(2m-w)^2}{2t} + \frac{w^2}{2t}} \\ &= \frac{2(2m - w)}{t} e^{-\frac{2m(m-w)}{t}} \end{aligned}$$

$$\begin{aligned} f_{M(t),W(t)}(x,y) &= \frac{2(2m-w)}{\sqrt{2\pi t}} e^{-\frac{(2m-w)^2}{2t}} \\ f_{W(t)}(w) &\sim N(0,1) \end{aligned} \tag{3.7.7}$$



Thank you for listening~

