

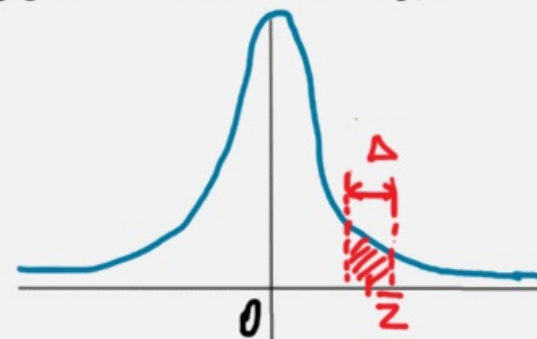
## Probability as "Measure"

Consider a normally distributed random variable

$$Z_t \sim N(0,1)$$

Probability density function:

$$f(z_t) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z_t^2}$$



The probability that  $Z_t$  falls *near* a specific value  $\bar{z}$  is:

$$\mathbb{P}\left(\bar{z} - \frac{1}{2}\Delta < Z_t < \bar{z} + \frac{1}{2}\Delta\right) = \int_{\bar{z} - \frac{1}{2}\Delta}^{\bar{z} + \frac{1}{2}\Delta} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z_t^2} dz_t$$


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If the region  $\Delta$  around  $\bar{z}$  is small  $\Rightarrow f(z_t) \cong f(\bar{z})$ :

$$\int_{\bar{z}-\frac{1}{2}\Delta}^{\bar{z}+\frac{1}{2}\Delta} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z_t^2} dz_t \cong \int_{\bar{z}-\frac{1}{2}\Delta}^{\bar{z}+\frac{1}{2}\Delta} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\bar{z}^2} = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\bar{z}^2} \cdot \Delta$$

This probability is a "mass" represented by a rectangle:



$$f(\bar{z}) = \mathbb{P}\left(\bar{z} - \frac{1}{2}\Delta < z_t < \bar{z} + \frac{1}{2}\Delta\right)$$

That is, probability corresponds to a "measure" associated with possible values of  $z_t$ .



For infinitesimal  $\Delta = dz_t$ , we denote these measures by the symbol  $dP(z_t)$ :

$$dP(z_t) = P\left(\bar{z} - \frac{1}{2}dz_t < z_t < \bar{z} + \frac{1}{2}dz_t\right)$$

$$\int_{-\infty}^{\infty} dP(z_t) = 1$$

The expected value of  $z_t$  is the "center of probability mass"

$$E[z_t] = \int_{-\infty}^{\infty} z_t dP(z_t)$$

The variance indicates how the probability mass spreads around the center:

$$E[z_t - E[z_t]]^2 = \int_{-\infty}^{\infty} [z_t - E[z_t]]^2 dP(z_t)$$



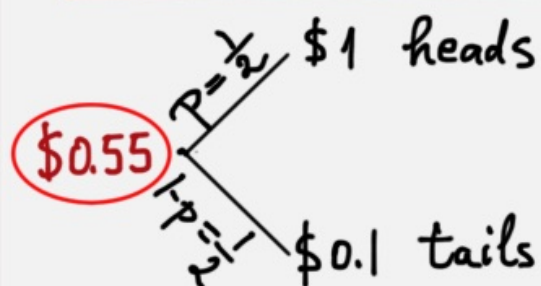
## We Change Probability Measures To Justify Prices

Game: You flip a coin, the values of the payoff are:

$$Z = \begin{cases} \$1, & \text{if you flip heads} \\ \$0.1, & \text{if you flip tails} \end{cases}$$

How much would you pay to play this game?

Real-World Measure



$$\frac{1}{2} \cdot \$1 + \frac{1}{2} \cdot \$0.1 = \$0.55$$

The Game Dealer Asks \$0.60



$$\$1 \cdot q + \$0.10(1-q) = \$0.60$$

$$0.9q = 0.5 \Rightarrow q = \frac{5}{9}$$

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## Static Theory

$$Z > 0, Q \text{ a.s.}, E[Z] = 1$$

$$\text{Define } \bar{Q}(A) \triangleq E[1_A \cdot Z]$$

$$\bar{Q}(\emptyset) = 0 : \bar{Q}(\emptyset) = E[1_{\emptyset} \cdot Z] = E[0 \cdot Z] = E[0] = 0$$

$$\bar{Q}(\Omega) = 1 : \bar{Q}(\Omega) = E[1_{\Omega} \cdot Z] = E[1 \cdot Z] = E[Z] = 1$$

$$A \in \mathcal{F} \Rightarrow \bar{Q}(A) \geq 0 : \text{Denote } N = \{\omega \in A \mid Z(\omega) < 0\}$$

$$\bar{Q}(A) = E[1_A \cdot Z] = \int_A Z dQ = \int_{A \setminus N} Z dQ + \int_N Z dQ = \int_{A \setminus N} Z dQ > \int_{A \setminus N} 0 dQ = 0$$

$$\forall A, B \in \mathcal{F}, A \cap B = \emptyset \Rightarrow \bar{Q}(A \cup B) = \bar{Q}(A) + \bar{Q}(B)$$

$$\begin{aligned} \bar{Q}(A \cup B) &= \int Z \cdot 1_{A \cup B} dQ = \int Z(1_A + 1_B) dQ = \int_{\Omega} Z \cdot 1_A dQ \\ &+ \int_{\Omega} Z \cdot 1_B dQ = \int_A Z dQ + \int_B Z dQ = \bar{Q}(A) + \bar{Q}(B) \end{aligned}$$

## Dynamic Measure Change Consistency Condition

For  $0 \leq t \leq T$ ,  $\mathcal{F}_t \subset \mathcal{F}_T : A \in \mathcal{F}_t \Rightarrow A \in \mathcal{F}_T$   
 $\tilde{Q}_t(A) = \tilde{Q}_T(A), \quad \forall A \in \mathcal{F}_t$

$A \in \mathcal{F}_t \Rightarrow A, 1_A - \mathcal{F}_t$ -measurable

$Z_t$  is a martingale  $\Rightarrow Z_t = \mathbb{E}[Z_T | \mathcal{F}_t]$

$$\begin{aligned} \tilde{Q}_t(A) &= \mathbb{E}[1_A Z_t] = \mathbb{E}[1_A Z_t | \mathcal{F}_0] = \mathbb{E}[1_A \mathbb{E}[Z_T | \mathcal{F}_t] | \mathcal{F}_0] \\ &= \mathbb{E}[\mathbb{E}[1_A Z_T | \mathcal{F}_t] | \mathcal{F}_0] = \mathbb{E}[1_A Z_T | \mathcal{F}_0] = \mathbb{E}[1_A Z_T] = \tilde{Q}_T(A) \end{aligned}$$

Tower Rule



## Bayes' Rule

If  $0 \leq t \leq T$ , and  $X$  is an  $\mathcal{F}_T$ -measurable r.v.  
 so that  $\tilde{\mathbb{E}}[|X|] < +\infty$ , then  $\tilde{\mathbb{E}}[X|\mathcal{F}_t] = \frac{1}{Z_t} \mathbb{E}[Z_T X | \mathcal{F}_t]$

By definition,  $Y = \tilde{\mathbb{E}}[X|\mathcal{F}_t]$  is the conditional expectation  
 of  $X$  if  $\tilde{\mathbb{E}}[1_A \cdot Y] = \tilde{\mathbb{E}}[1_A \cdot X]$ .

Set  $Y = Z_t^{-1} \mathbb{E}[Z_T X | \mathcal{F}_t]$  and fix  $A \in \mathcal{F}_t \Rightarrow$   
 $Y$  and  $1_A$  are  $\mathcal{F}_t$ -measurable

$Z$  is a martingale  $\Rightarrow Z_t = \mathbb{E}[Z_T | \mathcal{F}_t]$

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$$\begin{aligned}
 \widetilde{E}[1_A Y] &= E[Z_T 1_A Y] = E[Z_T 1_A Y / \mathcal{F}_0] \stackrel{\text{Tower Rule}}{=} E[E[Z_T 1_A Y / \mathcal{F}_t] / \mathcal{F}_0] \\
 &= E[1_A Y \underbrace{E[Z_T / \mathcal{F}_t]}_{Z_t} / \mathcal{F}_0] = E[1_A Y Z_t / \mathcal{F}_0] \\
 &= E[1_A \cancel{Z_t}^{-1} E[Z_T X / \mathcal{F}_t] \cancel{Z_t} / \mathcal{F}_0] \\
 &= E[\underbrace{1_A}_{\text{Tower Rule}} E[Z_T X / \mathcal{F}_t]] = E[E[1_A Z_T X / \mathcal{F}_t]] \\
 &= E[1_A Z_T X] = \widetilde{E}[1_A X] \Rightarrow Y \text{ is conditional expectation of } X \\
 &\quad \frac{1}{Z_t} E[Z_T X / \mathcal{F}_t] = \widetilde{E}[X / \mathcal{F}_t]
 \end{aligned}$$

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## Forward Measure Pricing Formula

$X$  is an  $\mathcal{F}_T$ -measurable payoff. Then

$$\Pi_t(X) = P(t, T) E^T[X | \mathcal{F}_t]$$

$$B_t = \frac{P(t, T)}{Z(t, T)} ; B_T^{-1} = Z(T, T)$$

$$\Pi_t(X) = B_t E[B_T^{-1} X | \mathcal{F}_t] = \frac{P(t, T)}{Z(t, T)} E[Z(T, T) X | \mathcal{F}_t]$$

$$= P(t, T) \frac{Z(0, T)}{Z(t, T)} E\left[\frac{Z(T, T)}{Z(0, T)} X | \mathcal{F}_t\right]$$

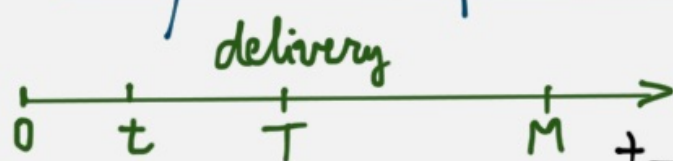
$$= P(t, T) \frac{1}{Z_t} E[Z_T X | \mathcal{F}_t] \rightarrow \text{Bayes' Rule}$$

$$= P(t, T) E^T[X | \mathcal{F}_t]$$

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## Properties of Forward Measure



Set up a forward contract to deliver at T the M-maturity bond

At T: Payoff  $= X = P(T, M) - F(t, T, M)$

At t:  $\Pi_t(X) = 0$

$$F(t, T, M) = \frac{P(t, M)}{P(t, T)}$$

$$0 = \Pi_t(X) = P(t, T) \mathbb{E}^T[X | \mathcal{F}_t]$$

$$P(t, T) \mathbb{E}^T[P(T, M) - F(t, T, M) | \mathcal{F}_t] = 0$$

$$\mathbb{E}^T[P(T, M)] = \mathbb{E}^T[F(t, T, M)] = F(t, T, M)$$

→  $\mathcal{F}_t$ -measurable

$$F(t, T, M) = \mathbb{E}^T[P(T, M) | \mathcal{F}_t]$$

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The process  $\{F(t, T, M)\}_{t=0}^T$  is a martingale under  $Q^T$

$$F(t, T, M) = \frac{P(t, M)}{P(t, T)}$$

$$t=T: F(T, T, M) = \frac{P(T, M)}{P(T, T)} = P(T, M)$$

$$F(t, T, M) = E^T[P(T, M) | \mathcal{F}_t]$$

$$F(t, T, M) = E^T[F(T, T, M) | \mathcal{F}_t]$$

$\Rightarrow F(t, T, M)$  is a  $Q^T$ -martingale

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## Forward Measure in Gaussian HJM

$Z(t, T) = \frac{P(t, T)}{B_t}$  - is a martingale under  $\mathbb{Q}$

$$dZ(t, T) = \Sigma(t, T) Z(t, T) dW_t \quad / : Z(t, T)$$

$$\frac{dZ(t, T)}{Z(t, T)} = \Sigma(t, T) dW_t \Rightarrow \text{geometric process}$$

Solution of SDE: Doleans exponential of Brownian motion

$$Z(t, T) = Z(0, T) \exp \left\{ \int_0^t \Sigma(s, T) dW_s - \frac{1}{2} \int_0^t \Sigma(s, T)^2 ds \right\}$$

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$$\frac{P(t, T)}{B_t} = \frac{P(0, T)}{B_0} \exp \left\{ \int_0^t \Sigma(s, T) dW_s - \frac{1}{2} \int_0^t [\Sigma(s, T)]^2 ds \right\}$$

$$\frac{dQ^T}{dQ} = \frac{P(t, T)/B_t}{P(0, T)/B_0} = \exp \left\{ \int_0^t \Sigma(s, T) dW_s - \frac{1}{2} \int_0^t [\Sigma(s, T)]^2 ds \right\}$$

$$\frac{dQ^T}{dQ} = \frac{P(t, T)/B_t}{P(0, T)/B_0}$$

Radon-Nikodym derivative of  $Q^T$  w.r.t.  $Q$

Compare with  $\xi_t$  in Girsanov's Theorem  $\Rightarrow$

$$\Sigma(s, T) = -\theta_s$$

Girsanov:  $dW_t^T = dW_t + \theta_t dt \Rightarrow$

$$dW_t^T = dW_t - \Sigma(s, T) dt \text{ is Brownian motion under } Q^T$$

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## Forward Bond Prices under Gaussian HJM

$$dF(t, T, M) = F(t, T, M) [\Sigma(t, M) - \Sigma(t, T)] dW_t^T$$

$$\frac{dF(t, T, M)}{F(t, T, M)} = [\Sigma(t, M) - \Sigma(t, T)] dW_t^T$$

geometric process

Solution of SDE: Doléans exponential

$$F(t, T, M) = F(0, T, M) \exp \left\{ \int_0^t [\Sigma(s, M) - \Sigma(s, T)] dW_s^T - \frac{1}{2} \int_0^t [\Sigma(s, M) - \Sigma(s, T)]^2 ds \right\}$$

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$$\log\left(\frac{F(t,T,M)}{F(0,T,M)}\right) = \int_0^t [\Sigma(s,M) - \Sigma(s,T)] dW_s^T - \frac{1}{2} \int_0^t [\Sigma(s,M) - \Sigma(s,T)]^2 ds$$

$$\begin{aligned} \mathbb{E}^T\left[\log\left(\frac{F(t,T,M)}{F(0,T,M)}\right)\right] &= -\frac{1}{2} \int_0^t [\Sigma(s,M) - \Sigma(s,T)]^2 ds \\ &= -\frac{1}{2} \text{Var}\left[\log\left(\frac{F(t,T,M)}{F(0,T,M)}\right)\right] \end{aligned}$$

$$\text{Var}\left[\log\left(\frac{F(t,T,M)}{F(0,T,M)}\right)\right] = \int_0^t [\Sigma(s,M) - \Sigma(s,T)]^2 ds$$

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