# Application of Generative Models in Commodity Trading

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#### Abstract

Abstract. Sample references: Ref 1 [1], ref 2 Jebara [2].

#### 1 Introduction

Introduction, related works, business case

#### 2 Related works

Related works

## 3 Methodology

Methodology

#### 3.1 Stochastic differential equations

**Definition 3.1.** A stochastic process is a random variable of the form

$$X_t(\omega): T \times \Omega \to \mathbb{R}$$

**Definition 3.2.** A stochastic process  $W = W_0, W_1, \dots, W_T$  is called a Wiener process if the following properties hold:

- 1.  $W_0 = 0$  with probability 1
- 2.  $\mathbb{E}W_t = 0$
- 3.  $Var[W_b W_a] = b a$  for all  $0 \le a \le b \le T$
- 4.  $W_b W_a \sim \mathcal{N}(0, b a)$
- 5.  $W_b W_a$  and  $W_d W_c$  are independent for all  $a \le b \le c \le d$

**Definition 3.3.** A stochastic differential equation is an equation of the form

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

where  $\mu$  and  $\sigma$  are some constants or some functions which represent the interest rate of nonrisky activities and is the market volatility respectively.

**Definition 3.4.** An Ito process is a stochastic process such that

$$X_t = X_0 + \int_0^t g(s) ds + \int_0^t h(s) dW_s$$

where  $g(t,\omega)$  and  $h(t,\omega)$  are random functions such that

$$P\left(\int_0^T |g(w,t)| \mathrm{d}t < \infty\right) = 1 \text{ and } P\left(\int_0^T h(w,t)^2 \mathrm{d}t < \infty\right) = 1$$

**Definition 3.5.** A diffusion process is the solution of SDE of the form

$$dX_t = b(t, X_t)dt + \sigma(t, X_t)dW_t$$

where some deterministic functions b and  $\sigma$  are called the drift and the diffusion coefficient of SDE such that

$$P\left\{ \int_0^T \sup_{|x| \le \mathbb{R}} (|b(t,x)| + \sigma^2(t,x)) dt < \infty \right\} = 1$$

A Black-Scholes-Merton or geometric Brownian motion

**Definition 3.6.** The Black-Scholes-Merton is described by the stochastic differential equation of the form

$$dX_t = \theta_1 X_t dt + \theta_2 X_t dW_t, X_0 = x_0$$

with  $\theta_2 > 0$ . The parameter  $\theta_1$  represents the constant interest rate and  $\theta_2$  — the volatility of risky activities.

The explicit solution is

$$X_t = x_0 \exp\left\{ \left( \theta_1 - \frac{1}{2}\theta_2^2 \right) t + \theta_2 W_t \right\}$$

B Cox-Ingersoll-Ross

Cox-Ingersoll-Ross

#### 3.2 Generative and Discriminative models

Generative and Discriminative models

A Bayesian Inference

Bayesian Inference

B Support Vector Machines

Support Vector Machines

C Generative adversarial networks

Generative adversarial networks

# 4 Case study

Case study

#### 4.1 Model description

Model description

### 4.2 Model training

Model training

#### 4.3 Dataset description

Dataset description

#### 4.4 Data preprocessing

Data preprocessing

#### 4.5 Quality metrics and results of testing

Quality metrics and results of testing

#### 5 Conclusion

Conclusion

## References

- [1] Y. Chen, Y. Wang, D. Kirschen, and B. Zhang. Model-free renewable scenario generation using generative adversarial networks. *IEEE Transactions on Power Systems*, 33(3), 2018. doi: 10.1109/TPWRS.2018.2794541.
- [2] T. Jebara. MACHINE LEARNING: Discriminative and Generative. Kluwer Academic Publishers, 2004.