

# Stochastic Modeling, 07.09.09

Note Title

7.9.2009

## Introduction

mathematical modeling:  
needed information

- natural laws
  - laws from economy
  - social laws
- scientific data

## 1. Weather forecast

1 day = 8am - 8am next day

day wet = if  $\geq 0.01$  inches  
precipitation

day dry = if  $< 0.01$  inches  
precipitation

1948 - 1983

January

$(X_{ij})_{i=1, j=1948}^{31, 1983}$

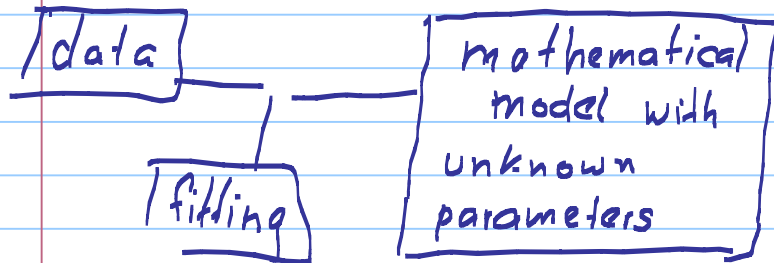
$X_{ij} = \begin{cases} 1 & \text{wet} \\ 0 & \text{dry} \end{cases}$

## 2. Share prices

NOKIA share, 1 year :

- price in Euro
  - price taken 12:00 each trading
- } rules

## 3 Modeling



model with  
parameters

describing  
the considered  
process



future  
events

## 2. Concepts of Stochastics

Probability space  $(\Omega, \mathcal{F}, \mathbb{P})$

$\Omega$  is a set

$$\Omega = \{\omega_1, \dots, \omega_n\}$$

$$\Omega = \{x, x \in [0, 1]\}$$

$$\Omega \neq \emptyset$$

$\mathcal{F}$  is a  $\sigma$ -algebra ( $\sigma$ -field)

Definition:  $\sigma$  algebra

Let  $\Omega \neq \emptyset$ . The collection  $\mathcal{F}$  of subsets  $A \subseteq \Omega$  is called a  $\sigma$ -algebra

$\sigma$ -algebra

$$(1) \emptyset, \Omega \in \mathcal{F}$$

$$(2) A \in \mathcal{F} \Rightarrow A^c = \Omega \setminus A \in \mathcal{F}$$

$$(3) A_1, A_2, \dots \in \mathcal{F} \Rightarrow \bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$$

Remark:  $A_1, A_2, A_3, \dots \in \mathcal{F}$

$$\Rightarrow \bigcap_{i=1}^{\infty} A_i \in \mathcal{F}.$$

Example  $\Omega$  is given

1. Largest  $\sigma$ -algebra:

$$\mathcal{F} = 2^{\Omega} = \text{all possible subsets } A \subseteq \Omega$$



2. The smallest  $\sigma$ -algebra:

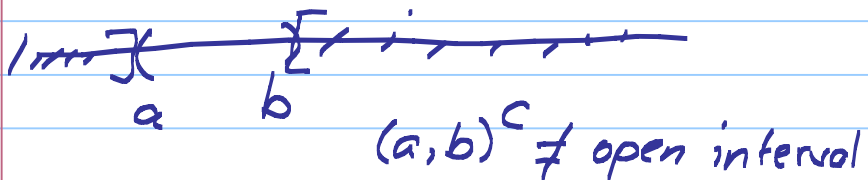
$$\mathcal{F} = \{\emptyset, \Omega\}$$

$$\emptyset^c = \Omega, \Omega^c = \emptyset$$

$$3. \mathcal{F} = \{ \emptyset, \Omega, A, A^c \}$$

$$\emptyset \neq A \neq \Omega$$

4.  $\mathcal{F} = \mathcal{B}(\mathbb{R}) =$  Borel sets of the real line  
= smallest  $\sigma$ -algebra,  
which contains all open  
intervals  $(a, b)$



$$(\Omega, \mathcal{F}, \mathbb{P})$$

↑   ↑   ↑

Definition: probability measure

Let  $\Omega \neq \emptyset$ ,  $\mathcal{F}$  is a  $\sigma$ -algebra on  $\Omega$ ,  
then the map

$$\mathbb{P}: \mathcal{F} \rightarrow [0, 1]$$

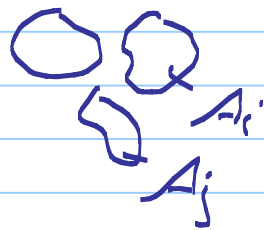
is called a probability measure if

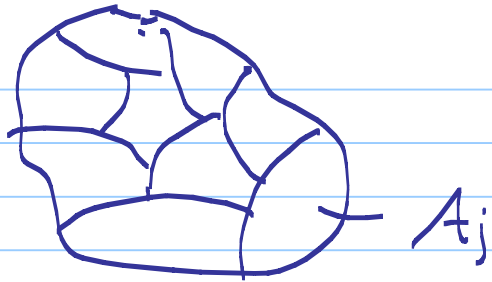
(1)  $\mathbb{P}(\Omega) = 1$ ,

(2)  $A_1, A_2, \dots \in \mathcal{F}$  such that

$$A_i \cap A_j = \emptyset, \quad i \neq j$$

$$\mathbb{P}\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mathbb{P}(A_i)$$





$$P(\Omega) = \sum_{i=1}^{\infty} P(A_i)$$

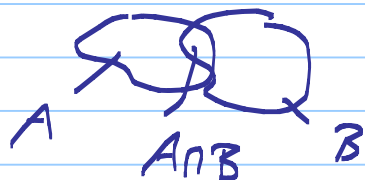
$$P(\Omega) = 1$$

## Properties of $\mathcal{P}$

1.  $\mathcal{P}(\emptyset) = 0,$

2.  $A, B \in \mathcal{F}$

$$\mathcal{P}(A \cup B) = \mathcal{P}(A) + \mathcal{P}(B) - \mathcal{P}(A \cap B)$$



3.  $A \in \mathcal{F}$

$$\mathcal{P}(A^c) = 1 - \mathcal{P}(A)$$

4. Continuity of  $\mathcal{P}$  from below

$$A_1, A_2, \dots \in \mathcal{F}$$

$$A_1 \subseteq A_2 \subseteq A_3 \subseteq \dots$$



$$\mathcal{P}\left(\bigcup_{i=1}^{\infty} A_i\right) = \lim_{i \rightarrow \infty} \mathcal{P}(A_i).$$

Example: Binomial distribution

$(\Omega, \mathcal{F}, \mathbb{P})$

$$\Omega = \{0, 1, \dots, n\}$$

$\mathcal{F}$  = set of all subsets of  $\Omega = 2^\Omega$

$$\mathbb{P}(\{k\}) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$0 < p < 1$$

$$k = 0, 1, \dots, n$$

$$\binom{n}{k} = \frac{n!}{k! (n-k)!}$$

$$\boxed{\mathbb{P}(A) = \sum_{k \in A} \mathbb{P}(\{k\})}$$

$A \in \mathcal{F}, A \subseteq \Omega \quad A = \{0, 2\}$

$$\mathbb{P}(\{0, 2\}) = \mathbb{P}(\{0\}) + \mathbb{P}(\{2\}).$$

### 3. Independence and conditional probability

Example: rolling a die

$$\begin{aligned} & \mathbb{P}(\text{1. time 3, 2. time 6}) \\ &= \mathbb{P}(\text{1 times 3})\mathbb{P}(\text{1 times 6}) \\ &= \frac{1}{6} \quad \frac{1}{6} \\ &= \frac{1}{36}. \end{aligned}$$

Definition (Independence  $A, B \in \mathcal{F}$ )

$A$  and  $B$  ( $\in \mathcal{F}$ ) are independent if  $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ .

Definition (Conditional Probability)

Let  $A \in \mathcal{F}$ ,  $P(A) > 0$ . Then

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

is called the conditional probability of B given A.

Notice if  $A$  and  $B$  are independent:

$$\begin{aligned} P(B|A) &= \frac{P(A \cap B)}{P(A)} = \frac{P(A)P(B)}{P(A)} \\ &= P(B). \end{aligned}$$

Definition: Independence of  $A_1, \dots, A_n \in \mathcal{F}$

$A_1, \dots, A_n$  are independent,

if for  $k=2,3,\dots,n$ ,  $1 \leq i_1 < i_2 < \dots < i_k \leq n$

$$P(A_{i_1} \cap \dots \cap A_{i_k}) = P(A_{i_1}) \dots P(A_{i_k})$$

Assume:  $(P(A_1 \cap A_2) \neq P(A_1)P(A_2))$

choose  $A_3 = \emptyset$

$$A_1 \cap A_2 \cap A_3 = \emptyset$$

$$P(A_1 \cap A_2 \cap A_3) = 0 = P(A_1) \times \\ \times P(A_2) \cdot 0 \\ \# \\ P(A_3)$$