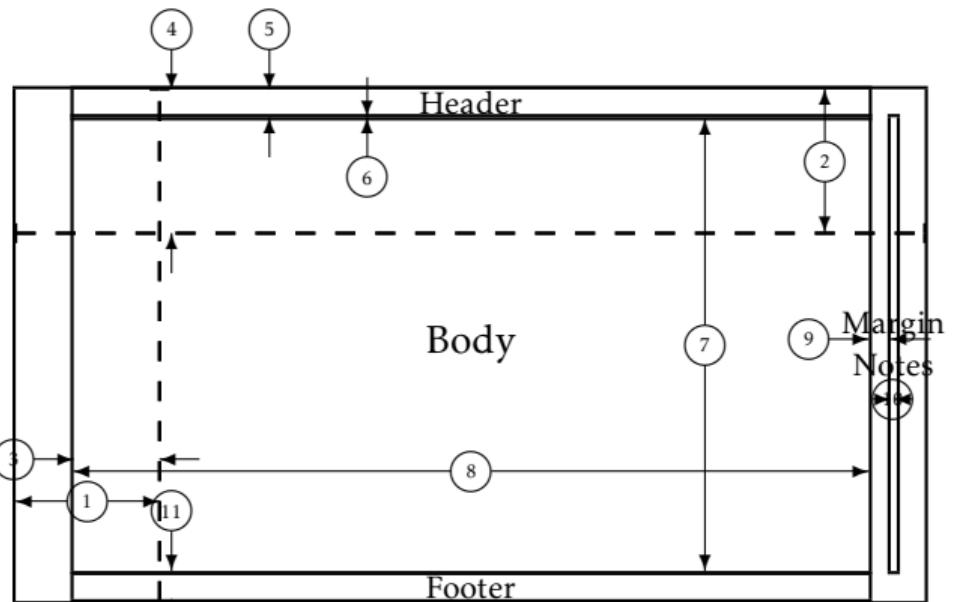


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Probability basics

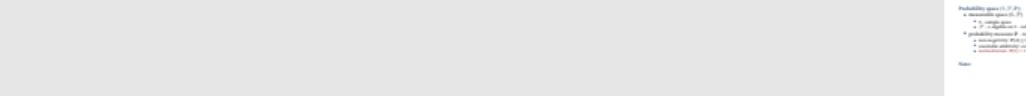
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1 one inch + \hoffset 2 one inch + \voffset
3 \oddsidemargin = -43pt 4 \topmargin = -72pt
5 \headheight = 14pt 6 \headsep = 0pt
7 \textheight = 227pt 8 \textwidth = 398pt
9 \marginparsep = 11pt 10 \marginparwidth = 3pt
11 \footskip = 14pt 11 \marginparpush = 5pt (not shown)

Outline

Blabla



1. σ -algebra \mathcal{S} - non-empty collection of subsets closed under complement and countable unions
 - if $A \in \mathcal{S}$, then $A^c \in \mathcal{S}$
 - if $A_i \in \mathcal{S}$ for $i \in I$ (countable index set), then $\bigcup_{i \in I} A_i \in \mathcal{S}$
 - if $A \in \mathcal{S}$ and $B \in \mathcal{S}$, then $A \cup B \in \mathcal{S}$
 - in consequence $\emptyset \in \mathcal{S}$ and $S \in \mathcal{S}$
2. (S, \mathcal{S}) forms a measurable space in this context called the sample space.
3. Probability measure is the same as probability distribution or probability law
4. More generally a positive measure on (S, \mathcal{S}) is a function $\mu: \mathcal{S} \rightarrow [0, \infty]$ satisfying non-negativity and countable additivity. A probability measure is a positive measure with total measure equal to 1.
5. The triplet (S, \mathcal{S}, μ) is a measure space. Probability space is a special case of a measure space where the total measure is 1.
6. Any finite positive measure μ on the sample space (S, \mathcal{S}) can be re-scaled into a probability measure as $\mathbb{P}(A) = \mu(A)/\mu(S)$, $A \in \mathcal{S} \Rightarrow$ link to energy models.

Probability space

Probability space $(S, \mathcal{S}, \mathbb{P})$:

- measurable space (S, \mathcal{S})
 - S - sample space
 - \mathcal{S} - σ -algebra on S - collection of subsets
- probability measure \mathbb{P} - real-valued function on sample space (S, \mathcal{S}) s.t.:
 - non-negativity: $\mathbb{P}(A) \geq 0$ for all $A \in \mathcal{S}$
 - countable additivity: countable disjoint $\{A_i : i \in I\} \in \mathcal{S} \Rightarrow \mathbb{P}(\bigcup_{i \in I} A_i) = \sum_{i \in I} \mathbb{P}(A_i)$
 - normalization: $\mathbb{P}(S) = 1$

Note: any finite positive measure μ on $(S, \mathcal{S}) \Rightarrow$ prob. measure $\mathbb{P}(A) = \mu(A)/\mu(S)$.

1. Definition of discrete and continuous space is a bit tricky and depends on the definition of topology but simply speaking in discrete space the set S is countable.
 2. For d -dimensional Euclidean space $(\mathbb{R}^d, \mathcal{R}^d)$, $\mathbf{A} = A_1 \times A_2 \times \dots \times A_d \in \mathcal{R}^d$, $A_1, A_2, \dots, A_d \in \mathcal{R}$, $\lambda(\mathbf{A}) = \lambda(A_1) \times \lambda(A_2) \times \dots \times \lambda(A_d)$

Positive measure on (S, \mathcal{S}) -function $\mu : \mathcal{S} \rightarrow [0, \infty]$ s.t.:

- $\mu(\emptyset) = 0$
 - countable additivity: countable disjoint $\{A_i : i \in I\} \in \mathcal{S} \Rightarrow \mu(\bigcup_{i \in I} A_i) = \sum_{i \in I} \mu(A_i)$
 - \Rightarrow measure space (S, \mathcal{S}, μ)

Note: if $\mu(S) < \infty \Rightarrow ((S, \mathcal{S}, \mu))$ finite measure space.

Examples of measures:

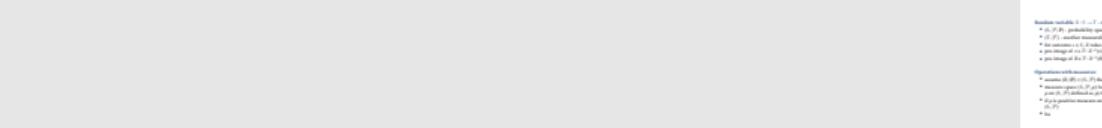
- counting measure: for discrete (S, \mathcal{S}) , $\#(A) =$ number of elements in $A \in \mathcal{S}$
 - Lebesgue measure: for Euclidean $(\mathbb{R}, \mathcal{R})$, interval $I = [a, b] \in \mathcal{R}$, $\lambda(I) = b - a$ - length
Euclidean $(\mathbb{R}^d, \mathcal{R}^d)$, $\mathbf{A} \in \mathcal{R}^d$, $\lambda(\mathbf{A}) = \lambda(A_1) \times \lambda(A_2) \times \dots \times \lambda(A_d)$ - area, volume
 - probability measure: positive finite measure on (S, \mathcal{S}) s.t. $\mathbb{P}(S) = 1$

1. sets $A, B \in \mathcal{S}$ are **equivalent** if $\mu(A \Delta B) = 0$, where $A \Delta B = (A \setminus B) \cup (B \setminus A)$
2. In probability space the almoste everywhere (a.e.) is equivalent to almost surely (a.s.) with respect to the probability measure \mathbb{P} .

Null sets and equivalence

Null set $A \in \mathcal{S}$ if $\mu(A) = 0$

- if statement holds for all $s \in S$ except for a null set, it holds **almost everywhere** (a.e.)
- sets $A, B \in \mathcal{S}$ are **equivalent** if $\mu(A \Delta B) = 0$, where $A \Delta B = (A \setminus B) \cup (B \setminus A)$
- measureble funcs $f, g : S \rightarrow T$ are **equivalent** if $\mu\{s \in S : f(s) \neq g(s)\} = 0$



Random variables

Random variable $X : S \rightarrow T$ - measurable function S to T

- (S, \mathcal{S}, μ) - probability space
- (T, \mathcal{T}) - another measurable space
- for outcome $s \in S$, X takes value $x = X(s) \in T$ - realization of r.v. X
- pre-image of $x \in T$: $X^{-1}(x) = \{s \in S : X(s) = x\} \in \mathcal{S}$
- pre-image of $B \in \mathcal{T}$: $X^{-1}(B) = \{s \in S : X(s) \in B\} \in \mathcal{S}$

Operations with measures:

- assume $(R, \mathcal{R}) \subset (S, \mathcal{S})$ then μ restricted to \mathcal{R} is measure on (R, \mathcal{R})
- measure space (S, \mathcal{S}, μ) func $f : S \rightarrow T$ and measure ν on $(T, \mathcal{T}) \Rightarrow$ **pullback measure** μ on (S, \mathcal{S}) defined as $\mu(A) = \nu(f(A))$ for $A \in \mathcal{S}$
- if μ is positive measure on (S, \mathcal{S}) then $c\mu$ for $c \in (0, \infty)$ is also positive measure on (S, \mathcal{S})
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