

Artificial Intelligence I

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1 Foundations

1.1 Mathematical Language

Def 1.1. Natural Language: A natural language is any form of spoken or signed means of communication that has evolved naturally in humans through use and repetition without conscious planning or premeditation.

Def 1.2. Domain of Discourse: The [set](#) of entities, objects, or concepts that a language, discussion, or formal system refers to. It specifies the scope within which terms, statements, and reasoning apply.

Def 1.3. Jargon (Terminology): A jargon (or terminology) is a [set](#) of specialized words or phrases (called technical terms or just terms) relating to concepts from a particular [domain of discourse](#).

Def 1.4. Technical Language: A technical language is a [natural language](#) extended by a [terminology](#) and (possibly) special idioms, discourse markers, and notations.

Def 1.5. Stylized Language: Mathematicians use a stylized language that uses formulae to represent mathematical objects, uses math idioms for special situations (e.g. "iff", "hence", "let... be..., then..."), and classifies statements by role (e.g., definition, Lemma, Theorem, Proof, Example).

Def 1.6. Mathematical Vernacular: Specialized [technical language](#) used by mathematicians to express definitions, proofs, and reasoning. It combines natural language, standard mathematical idioms, and symbolic expressions into a coherent style for clear mathematical communication.

Def 1.7. Formulae: A mathematical formula can be a **mathematical statement** (clause) which can be true or false (e.g. $x > 5, 3 + 5 = 7$), or a **mathematical object** (e.g. $3, n, x^2 + y^2 + z^2, \int_1^0 x^{3/2} dx$)

Def 1.8. Sequence: [mathematical vernacular](#) uses the concept of sequences instead of lists. Sequences are usually finite, but can be infinite as well.

Def 1.9. Ellipses: [mathematical vernacular](#) uses ellipses (\dots) as a constructor for [sequences](#). The meaning of an ellipsis is usually considered "obvious" and left for interpretation by the reader.

[ellipses](#) allow to write down large [objects](#) easily, examples:

- $1, \dots, n \rightsquigarrow$ the [sequence](#) of natural numbers between 1 and n in order.
- $1, 4, 9, 16, \dots \rightsquigarrow$ the [sequence](#) of squares in orders.
- $e_1, \dots, e_n \rightsquigarrow$ a [sequence](#) of [objects](#) e_i for $1 \leq i \leq n$

Def 1.10. Varieties of Mathematical Statements: [Mathematical statements](#) come in different epistemic varieties:

- **Definitions:** [statements](#) that introduce new global identifier for important [objects](#)
- **Assertions:** [statements](#) that state properties of [mathematical objects](#)
- **Examples:** [statements](#) that exhibit a witness for some property
- **Axioms:** [statements](#) that characterize the [objects](#) of a certain [domain of discourse](#) or theory. An axiom (postulate) is a statement about [mathematical objects](#) that we *assume* to be true.

Def 1.11. Pragmatic Categories of Mathematical Assertions: [Mathematical assertions](#) are pragmatically classified into categories:

- **proofs** are arguments that justify the truth of [statements](#) beyond any doubt. [Proofs](#) are not really [statements](#), but we sometimes treat them together.
- **proposition** is a [statement](#) which is interesting in its own right.
- **lemma** is an easily proved [statement](#) which is helpful for proving other [propositions](#) and [theorems](#), but is usually not particularly interesting in its own right. A **lemma** is an intermediate [theorem](#) that serves as part of a [proof](#) of a larger [theorem](#).
- **theorem** is a more important [statement](#) than a [proposition](#) which says something definitive on the subject, and often takes more effort to prove than a [proposition](#) or [lemma](#). A **theorem** is a [statement](#) about [mathematical object](#) that we *know* to be true.
- **corollary** is a quick consequence of a [proposition](#) or [theorem](#) that was proven recently. It is a [theorem](#) that follows directly from another [theorem](#).

- **conjecture** is a **statement** that is thought to be provable, but has not been yet. A **conjecture** that is proven is a **theorem**

Def 1.12. Declaration: In **mathematical vernacular** we call a clause that introduce new identifiers together with some properties a declaration. For example Let $\epsilon, \delta > 0 \dots$

Def 1.13. Scope: The scope of an identifier is the part of an expression where the reference is valid; that is, where the identifier can be used to refer. In the other parts, the identifier may refer to a different entity, or to nothing at all (unbound)

Def 1.14. Definiendum: The definiendum is the symbol, expression, or term being introduced by a **definition**. It is the newly named **object** or concept.

Def 1.15. Definiens: The definiens is the symbol, expression, or construction that specifies the meaning of the **defineiendum**. It explains what the **defineiendum** denotes.

Def 1.16. Simple Definition: A **simple definition** introduces a new name (the **defineiendum**) for an existing **object** or concept (the **definiens**). The defineiendum must be new and must not occur in the definiens. We write simple definitions using $:=$ or \equiv , for example $\phi := 2$.

Def 1.17. Pattern Definition: A **pattern definition** introduces a new operator or **relation** by applying it to distinct pattern variables v_1, \dots, v_n . The **definiens** is an expression in these variables. Applying the operator to arguments a_1, \dots, a_n means substituting v_i by a_i in the definiens. We use $:=$ for operator definitions and $:\Leftrightarrow$ for relation definitions. Example:

$$A \cap B := \{x \mid x \in A \wedge x \in B\}.$$

Def 1.18. Implicit Definition: An **implicit definition** (definition by description) specifies a new name n (the **defineiendum**) by giving a clause A such that we can prove unique existence $\exists^1 n. A$. If such a **proof** exists, n is **well-defined**. Example: the empty set Φ is implicitly defined by $\forall x. x \notin \Phi$. Another example: the exponential function is the unique $f : \mathbb{R} \rightarrow \mathbb{R}$ satisfying $f' = f$ and $f(0) = 1$.

Def 1.19. Example: An example E is a **mathematical statement** that consists of:

- symbol p for the **exemplandum** (plural: exemplanda): the property to be exemplified,
- the **exemplans** (plural: exemplantia): an expression A denoting a **mathematical object** that acts as witness object for the property p , and
- (optionally) a **justification** of E , i.e. a **proof** π that $p(A)$ holds in the current context.

Def 1.20. Counter Example: An **example** for the complement of the property p where π is a **proof** that $p(A)$ does not hold.

Def 1.21. Running Example: An **example** that is recalled time and again during an argument, applying the newly discovered knowledge to it, presumably to show how things work.

Note

An **example** (or even several) for a **mathematical statement** B do not constitute a **proof** for B . On the other hand, one **counter example** for a **mathematical statement** B does show that $\neg B$ is true ($\exists x. \neg A \equiv \neg(\forall x. A)$).

Def 1.22. Proof by Contradiction: A **proof by contradiction** assumes the negation $\neg A$ of the claim A and derives a contradiction from that assumption. Since $\neg A$ cannot be true, we infer $\neg\neg A$ and therefore A .

Def 1.23. Proof by Local Hypothesis: A **proof by local hypothesis** is a proof of an implication $A \rightarrow B$ where we temporarily assume A and derive B from that assumption. After the subproof is closed, the assumption A may no longer be used.

Def 1.24. Chaining: **Chaining** applies a general implication $A \rightarrow B$ to a specific instance A' obtained from A by substituting variables with concrete values. This allows us to conclude the corresponding instance B' obtained from B by the same substitution.

Def 1.25. Without Loss of Generality (WLOG): An **inference** principle used in **proofs** when a **statement** $Q(x)$ must be shown for all cases of x , but the different cases are equivalent by symmetry or trivial reduction. **Formally**, if the domain of x can be partitioned into cases A_1, \dots, A_k and it can be shown that:

- proving $Q(x)$ under one representative case A_i suffices, because
- for every other case A_j there is either a trivial **proof** of $Q(x)$ or a direct reduction to the representative case,

then we may assume A_i holds “without loss of generality”. This allows us to simplify the argument by only proving the representative case.

Note

WLOG Example. if we want to prove the property $Q(p)$ for all primes p . We know that every prime is either 2 (the only even prime) or an odd prime. If the case $p = 2$ is easy, and the general odd prime case is the only real work, then we can say:

WLOG, assume p is odd

This does not exclude the even case; it means we have already checked that the even case adds no new difficulty.

Def 1.26. Greek Alphabet: In [mathematical vernacular](#), Greek letters are used extensively to convey precise meanings. [Table 1](#) lists the Greek letters you are expected to recognize and write fluently.

Note

In addition, [table 2](#) shows the curly Roman and Fraktur letters, which are also standard in mathematical writing.

α	A	alpha	β	B	beta	γ	Γ	gamma
δ	Δ	delta	ϵ	E	epsilon	ζ	Z	zeta
η	H	eta	θ, ϑ	Θ	theta	ι	I	iota
κ	K	kappa	λ	Λ	lambda	μ	M	mu
ν	N	nu	ξ	Ξ	xi	o	O	omicron
π, ϖ	Π	pi	ρ	P	rho	σ	Σ	sigma
τ	T	tau	υ	Υ	upsilon	φ, ϕ	Φ	phi
χ	X	chi	ψ	Ψ	psi	ω	Ω	omega

Table 1: Greek Alphabet

Roman	$\mathcal{A}, \mathcal{B}, \mathcal{C}, \mathcal{D}, \mathcal{E}, \mathcal{F}, \mathcal{G}, \mathcal{H}, \mathcal{I}, \mathcal{J}, \mathcal{K}, \mathcal{L}, \mathcal{M}, \mathcal{N}, \mathcal{O}, \mathcal{P}, \mathcal{Q}, \mathcal{R}, \mathcal{S}, \mathcal{T}, \mathcal{U}, \mathcal{V}, \mathcal{W}, \mathcal{X}, \mathcal{Y}, \mathcal{Z}$
Fraktur	$\mathfrak{A}, \mathfrak{B}, \mathfrak{C}, \mathfrak{D}, \mathfrak{E}, \mathfrak{F}, \mathfrak{G}, \mathfrak{H}, \mathfrak{I}, \mathfrak{J}, \mathfrak{K}, \mathfrak{L}, \mathfrak{M}, \mathfrak{N}, \mathfrak{O}, \mathfrak{P}, \mathfrak{Q}, \mathfrak{R}, \mathfrak{S}, \mathfrak{T}, \mathfrak{U}, \mathfrak{V}, \mathfrak{W}, \mathfrak{X}, \mathfrak{Y}, \mathfrak{Z}$

Table 2: Roman and Fraktur Letters

1.2 Sets

Def 1.27. Set: A collection of different things (called **elements** or **members** of the set). We can represent **sets** by listing the **elements** within curly brackets, e.g. $\{a, b, c\}$ or by describing the **elements** via a property, e.g. $\{x \mid x \bmod 2 = 0\}$ (the **set** of even numbers). We can state **element**-hood ($a \in S$) or not ($b \notin S$)

Def 1.28. Set Equality: $A \equiv B := \forall x. x \in A \Leftrightarrow x \in B$

Def 1.29. Subset: $A \subseteq B := \forall x. x \in A \Rightarrow x \in B$

Def 1.30. Proper Subset: $A \subset B := (A \subseteq B) \wedge (A \neq B)$

Def 1.31. Super Set: $A \supseteq B := \forall x. x \in B \Rightarrow x \in A$

Def 1.32. Proper Superset: $A \supset B := (A \supseteq B) \wedge (A \neq B)$

Def 1.33. Set Union: $A \cup B := \{x \mid x \in A \vee x \in B\}$

Def 1.34. Set Intersection: $A \cap B := \{x \mid x \in A \wedge x \in B\}$

Def 1.35. Disjoint: Two **sets** A, B are **disjoint** iff $A \cap B = \Phi$

Def 1.36. Set Difference: $A \setminus B := \{x \mid x \in A \wedge x \notin B\}$

Def 1.37. Power Set: $\mathcal{P}(A) := \{S \mid S \subseteq A\}$ (the set of all subsets)

Def 1.38. Cartesian Product: $A \times B := \{(a, b) \mid a \in A \wedge b \in B\}$, (a, b) is a pair

Def 1.39. Family: A **set** of **sets**

Def 1.40. Topology: A **family** of open **sets**

Def 1.41. Indexing Function: Let I and X be **sets**. A function $f : I \rightarrow X$ is called an indexing function. The set I is

called the index set. For each $i \in I$, we define $x_i := f(i)$, here i is called the index (or parameter) of x_i .

Def 1.42. Indexed Family: Given an **indexing function** $f : I \rightarrow X$, the **set** of values $f(I) = \{f(i) : i \in I\}$ is called an indexed **family**. It is often written as $(x_i)_{i \in I}$, $\langle x_i \rangle_{i \in I}$, or $\{x_i\}_{i \in I}$.

Def 1.43. Union over a Family: Let $(X_i)_{i \in I}$ be an **indexed family**, then $\bigcup_{i \in I} X_i := \{x \mid \exists i \in I. x \in X_i\}$

Def 1.44. Intersection over a Family: Let $(X_i)_{i \in I}$ be an **indexed family**, then $\bigcap_{i \in I} X_i := \{x \mid \forall i \in I. x \in X_i\}$

Def 1.45. n -fold Cartesian Product: $\prod_{i=1}^n X_i := X_1 \times \cdots \times X_n := \{\langle x_1, \dots, x_n \rangle \mid \forall i. 1 \leq i \leq n \Rightarrow x_i \in X_i\}$ where $\langle x_1, \dots, x_n \rangle$ is called an n -tuple.

Def 1.46. n -dim Cartesian Space: $X^n := \{\langle x_1, \dots, x_n \rangle \mid 1 \leq i \leq n \Rightarrow x_i \in X\}$ where $\langle x_1, \dots, x_n \rangle$ is called a vector.

Def 1.47. Size of a Set: The size $\Gamma(A)$ (or $|A|$) of a **set** A is the number of **elements** in A .

1.3 Relations

Def 1.48. Relation: $R \subseteq A \times B$ is a (binary) relation between A and B

Def 1.49. Relation on: a **relation** $R \subseteq A \times B$ where $A = B$ is called a *relation on* A

Def 1.50. Total: A **relation** $R \subseteq A \times B$ is called *total* (*left total*) iff $\forall x \in A. \exists y \in B. (x, y) \in R$

Def 1.51. Converse Relation: $R^{-1} \subseteq B \times A := \{(y, x) \mid (x, y) \in R\}$ is the converse **relation** of R

Def 1.52. Relation Composition: The composition of $R \subseteq A \times B$ and $S \subseteq B \times C$ is defined as $R \circ S := \{(a, c) \in A \times C \mid \exists b \in B. (a, b) \in R \wedge (b, c) \in S\}$

A **relation on** A : $R \subseteq A \times A$ is called:

- **reflexive** on A , iff $\forall a \in A. (a, a) \in R$
- **irreflexive** on A , iff $\forall a \in A. (a, a) \notin R$
- **symmetric** on A , iff $\forall a, b \in A. (a, b) \in R \Rightarrow (b, a) \in R$
- **asymmetric** on A , iff $\forall a, b \in A. (a, b) \in R \Rightarrow (b, a) \notin R$
- **antisymmetric** on A , iff $\forall a, b \in A. (a, b) \in R \wedge (b, a) \in R \Rightarrow a = b$
- **transitive** on A , iff $\forall a, b, c \in A. (a, b) \in R \wedge (b, c) \in R \Rightarrow (a, c) \in R$
- **equivalence relation** on A , iff R is **reflexive symmetric**, and **transitive**

Def 1.53. Equality Relation: The equality relation is an **equivalence relation** on any **set**.

Def 1.54. Divides Relation: The *divides relation* on the integers is defined as $\mid \subseteq \mathbb{Z} \times \mathbb{Z}$, where $\mid = \{(a, b) \in \mathbb{Z} \times \mathbb{Z} \mid \exists k \in \mathbb{Z} : b = a \cdot k\}$. We write $a \mid b$ to denote $(a, b) \in \mid$ (read as " a divides b ").

Def 1.55. Congruence Modulo: For a fixed $n \in \mathbb{N}$ with $n \geq 1$, the *congruence modulo n* **relation on** the integers is defined as

$$\equiv_n \subseteq \mathbb{Z} \times \mathbb{Z}, \quad \equiv_n = \{(a, b) \in \mathbb{Z} \times \mathbb{Z} \mid n \mid (a - b)\}.$$

We write $a \equiv b \pmod{n}$ to denote $(a, b) \in \equiv_n$ (read as " a is congruent to b modulo n ").

Def 1.56. Partial Ordering (\preceq): A **relation** $R \subseteq A \times A$ is called a (non-strict) **partial ordering** on A , iff R is **reflexive**, **antisymmetric**, and **transitive** on A .

Def 1.57. Strict Partial Ordering (\prec): A **relation** $R \subseteq A \times A$ is called a **strict partial ordering** on A , iff R is **irreflexive**, and **transitive** on A .

Def 1.58. Comparable: In a **non-strict partial ordering**, two elements a, b are comparable if either $a \preceq b$ or $b \preceq a$. If neither holds, they are incomparable.

Def 1.59. Linear Order: A **partial ordering** is called linear (or total) on A , iff all **elements** in A are **comparable**, i.e. if $(x, y) \in R$ or $(y, x) \in R$ for all $x, y \in A$. For example, the \leq **relation** is a **linear order** on \mathbb{N} . However, the "**divides**" (\mid) **relation** is a **non-strict partial ordering** on \mathbb{N} that is not **linear** (e.g., neither $2 \mid 3$ nor $3 \mid 2$).

1.4 Functions

Def 1.60. Partial function: A [relation](#) $R \subseteq X \times Y$ is a *partial function* from X to Y ($f : X \rightarrow Y$) iff $\forall x \in X, \forall y_1, y_2 \in Y. ((x, y_1) \in R \wedge (x, y_2) \in R \Rightarrow y_1 = y_2)$ (i.e. there is at most one $y \in Y$ with $(x, y) \in f$)

Def 1.61. dom, codom: We call X the domain ($\text{dom}(f)$), and Y the codomain ($\text{codom}(f)$)

Def 1.62. Function Application: Instead of writing $(x, y) \in f$ we write $f(x) = y$

Def 1.63. Undefined: we call a [partial function](#) $f : X \rightarrow Y$ undefined at $x \in X$, iff $\forall y \in Y. (x, y) \notin f$ ($f(x) = \perp$).

Def 1.64. Function: A [partial function](#) $f : X \rightarrow Y$ is called a *function* (or *total function*) from X to Y iff it is a [total relation](#). ($\forall x \in X. \exists^1 y \in Y : (x, y) \in f$)

Note

A [partial function](#) guarantees uniqueness, but not existence. So each x has at most one y (some might have none). A [total function](#) guarantees both uniqueness and existence (at most and at least \rightarrow exactly one)

A [function](#) $f : X \rightarrow Y$ is called

- **injective** iff $\forall x_1, x_2 \in X. f(x_1) = f(x_2) \Rightarrow x_1 = x_2$
- **surjective** iff $\forall y \in Y. \exists x \in X. f(x) = y$
- **bijective** iff f is [injective](#) and [surjective](#)

Def 1.65. Function Composition: If $f : A \rightarrow B$ and $g : B \rightarrow C$ are [functions](#), then we call $g \circ f : A \rightarrow C; x \mapsto g(f(x))$ the composition of g and f (read g after f).

Def 1.66. Image and Preimage: Let $f : X \rightarrow Y$ be a [function](#), $X' \subseteq X$ and $Y' \subseteq Y$, then we call

- $f(X') := \{y \in Y \mid \exists x \in X'. (x, y) \in f\}$ the **image** of X' under f ,
- $\text{Im}(f) := f(X)$ the **image** of f , and
- $f^{-1}(Y') := \{x \in X \mid \exists y \in Y'. (x, y) \in f\}$ the **preimage** of Y' under f .

Def 1.67. Cardinality: We say that a [set](#) A is **finite** and has **cardinality** $\Gamma(A) \in \mathbb{N}$, iff there is a [bijective function](#) $f : A \rightarrow \{n \in \mathbb{N} \mid n < \Gamma(A)\}$.

Def 1.68. Countably Infinite: We say that a [set](#) A is countably infinite, iff there is a [bijective function](#) $f : A \rightarrow \mathbb{N}$.

Def 1.69. Countable: A [set](#) A is called countable, iff it is [finite](#) or [countably infinite](#).

Def 1.70. Curried Function Type: Let X, Y, Z be sets. An *uncurried function* is written as

$$f : X \times Y \rightarrow Z, \quad f(x, y) = E(x, y).$$

The same [function](#) can equivalently be written in *curried* form as

$$f : X \rightarrow Y \rightarrow Z, \quad f(x)(y) = E(x, y).$$

Thus in the curried notation the [function](#) takes one argument at a time: for $x \in X$, the value $f(x)$ is itself a [function](#) $Y \rightarrow Z$, and applying it to $y \in Y$ yields $f(x)(y) \in Z$.

Def 1.71. Invertible: Let A and B be [sets](#) and $f : A \rightarrow B$ a [function](#), then f is called invertible, iff there is a [function](#) $g : B \rightarrow A$, such that $f \circ g = \text{Id}_B$. g is called the inverse function (or just inverse) of f and is written as f^{-1} .

1.5 Mathematical Structures

Def 1.72. Formulae: A mathematical formula can be a **mathematical statement** (clause) which can be true or false (e.g. $x > 5, 3 + 5 = 7$), or a **mathematical object** (e.g. $3, n, x^2 + y^2 + z^2, \int_1^0 x^{3/2} dx$)

Def 1.73. Mathematical Structure: A mathematical structure combines multiple [mathematical objects](#) (the components) into a new [object](#). The components usually have names by which they can be referenced. Given a [definition](#) of a mathematical structure S , we say that any [object](#) that conforms to that is an instance of S .

Def 1.74. Group: A group is a [mathematical structure](#) $\langle G, \circ, e, \cdot^{-1} \rangle$ that consists of:

- a base [set](#) G of [objects](#),

- an operation $\circ : G \times G \rightarrow G$, such that $\forall a, b \in G. a \circ b \in G$ and $\forall a, b, c \in G. (a \circ b) \circ c = a \circ (b \circ c)$,
- a unit $e \in G$, such that $\forall a \in G. e \circ a = a \circ e = a$, and
- the inverse function $.^{-1} : G \rightarrow G$, such that $\forall a \in G. a \circ a^{-1} = a^{-1} \circ a = e$

An example of a **group** is the **set** $G = \mathbb{Z}$, with the operation $+: \mathbb{Z} \times \mathbb{Z} \rightarrow \mathbb{Z}$; then $e = 0$ and $.^{-1}$ is $\lambda x \in \mathbb{Z}. -x$. We write it as $\langle \mathbb{Z}, +, 0, - \rangle$.

Question: can we do the same for multiplication where $e = 1$? If we think about it, we would be stuck on finding the $.^{-1}$ component of the group. What can we always multiply with any integer and get $e = 1$ back? Take for example 1, we can multiply it with $\frac{1}{1}$ (great!). Take -1 , we can multiply it with $\frac{1}{-1}$ (also great!). But take any other integer, e.g. 2, we must multiply it with $\frac{1}{2}$ to get 1 back, however, $\frac{1}{2} \notin \mathbb{Z}$, it is however $\in \mathbb{Q}$! The multiplicative group would only make sense iff $G = \mathbb{Q} \setminus \{0\}$. We write that as $\langle \mathbb{Q} \setminus \{0\}, *, 1, a \mapsto \frac{1}{a} \rangle$.

But we need some multiplication **structure** that works for \mathbb{Z} !

Def 1.75. Monoid: A monoid is a **mathematical structure** $\langle M, \circ, e \rangle$ that consists of:

- A base set M of **objects**,
- An operation $\circ : M \times M \rightarrow M$, such that for all $a, b \in M$, we have $a \circ b \in M$ (closure),
- Associativity: for all $a, b, c \in M$, $(a \circ b) \circ c = a \circ (b \circ c)$, and
- A unit element $e \in M$, such that for all $a \in M. e \circ a = a \circ e = a$.

Hence, the **mathematical structure** $\langle \mathbb{Z}, *, 1 \rangle$ is a **monoid** (a **group** missing an inverse function).

We saw that $\langle \mathbb{Z}, +, 0, - \rangle$ is a **group** under addition. And $\langle \mathbb{Z}, *, 1 \rangle$ is a **monoid** under multiplication. Now, we want to do everyday arithmetics with integers; we want expressions where we have additions and multiplication (e.g. $a * (b + c)$) which evaluates to $(a * b) + (a * c)$.

For that, we need a **mathematical structure** where both addition and multiplication live together.

Def 1.76. Ring: A ring is a **mathematical structure** $\langle R, +, 0, -, *, 1 \rangle$ consisting of:

- A base **set** R .
- An addition operation $+: R \times R \rightarrow R$ such that $\langle R, +, 0, - \rangle$ is a **group** (called abelian **group**).
- A multiplication operation $*: R \times R \rightarrow R$ such that $\langle R, *, 1 \rangle$ is a **monoid**.
- Distributivity: for all $a, b, c \in R, a * (b + c) = (a * b) + (a * c), (a + b) * c = (a * c) + (b * c)$

Hence, mixing components from the **group** $\langle \mathbb{Z}, +, 0, - \rangle$ and the **monoid** $\langle \mathbb{Z}, *, 1 \rangle$ leaves us with a **ring** $\langle \mathbb{Z}, +, 0, -, *, 1 \rangle$

Def 1.77. Magma: A magma is a **mathematical structure** $\langle M, \circ \rangle$ consisting of:

- A base **set** M , and
- A binary operation $\circ : M \times M \rightarrow M$.

Example: $\langle \mathbb{Z}, - \rangle$ (integers under subtraction). $\langle \mathbb{N}, + \rangle$ (natural numbers under addition)

Def 1.78. Semigroup: A semigroup (**magma** with associativity) is a **mathematical structure** $\langle S, \circ \rangle$ consisting of:

- A base **set** S ,
- A binary operation $\circ : S \times S \rightarrow S$, and
- Associativity: for all $a, b, c \in S, (a \circ b) \circ c = a \circ (b \circ c)$

Example: $\langle \mathbb{Z}, + \rangle$ (integers under addition).

1.6 Formal Languages and Grammars

Def 1.79. Alphabet: An alphabet is a **finite set**; we call each element $a \in A$ a **character**, and an n -tuple $s \in A^n$ a **string** (of length n over A). We often write a string $\langle c_1, \dots, c_n \rangle$ as " $c_1 \dots c_n$ ", for instance "**abc**" for $\langle a, b, c \rangle$

Def 1.80. Empty String: Let A be an **alphabet**, then $A^0 = \{\langle \rangle\}$, where $\langle \rangle$ is the unique 0-tuple. We consider $\langle \rangle$ as the

string of length 0 and call it the **empty string** and denote it with ϵ .

Def 1.81. String Length: Given a string s , we denote its length with $|s|$.

Def 1.82. String Concatenation: The concatenation $\text{conc}(s, t)$ of two strings $s = \langle s_1, \dots, s_n \rangle \in A^n$ and $t = \langle t_1, \dots, t_m \rangle \in A^m$ is defined as $s + t$ or simply st

Def 1.83. Kleene Plus/Star: Let A be an **alphabet**, then we define the **sets**:

- $A^+ := \bigcup_{i \in \mathbb{N}^+} A^i$ (*nonempty strings*), and
- $A^* := A^+ \cup \{\epsilon\}$ (*strings*).

Def 1.84. Formal Language: A **set** $L \subseteq A^*$ is called a *formal language* over A . For example, If $A = \{a, b\}$ then:

$$A^* = \bigcup_{n=0}^{\infty} A^n$$

where $A^0 = \epsilon$, $A^1 = \{a, b\}$, $A^2 = \{a, b, ab, ba\}$, and so on ...

$$\begin{aligned} A^* &= \{\epsilon, a, b, aa, ab, ba, bb, aaa, aab, aba, abb, \dots\} \\ A^+ &= \{a, b, aa, ab, ba, bb, aaa, aab, aba, abb, \dots\} \end{aligned}$$

A **formal language** might be $L = \{a, b, aab, bbb, \dots\}$

Def 1.85. Repeating Chars: We use $c^{[n]}$ for the string that consists of the character c repeated n times. **Examples:**

- Let $A = \{a, b\}$, $a^{[5]} = \text{"aaaaa"}$.
- The **set** $M := \{ba^{[n]} \mid n \in \mathbb{N}\}$ of strings that start with character b followed by an arbitrary number of a 's is a **formal language** over $A = \{a, b\}$

Note

A **formal language** might be infinite and even undecidable even if the **alphabet** A is **finite**. For example, let $A = \{a, b\}$ then the **formal language** $L = \{a, ab, bba\}$ is **finite** but the **formal language** $L = \{a^{[n]} \mid n \in \mathbb{N}\}$ is infinite.

Since we cannot list all strings in a **formal language** L because there are infinitely many, we need a **finite description** that can generate all strings in L . We need **Grammars**.

Def 1.86. Phrase Structure Grammar: A phrase structure grammar (*type 0 grammar, unrestricted grammar, grammar*) is a tuple $\langle N, \Sigma, P, S \rangle$ where

- N is a **finite set** of **nonterminal symbols**,
- Σ is a **finite set** of **terminal symbols**. (members of $N \cup \Sigma$ are called symbols).
- P is a **finite set** of **production rules**: pairs $p := h \rightarrow b$ (also written as $h \Rightarrow b$), where $h \in (\Sigma \cup N)^* N (\Sigma \cup N)^*$ and $b \in (\Sigma \cup N)^*$. The string h is called the **head** of p and b the **body**.
- $S \in N$ is a distinguished symbol called the **start symbol** (also **sentence symbol**).

The **sets** N and Σ are assumed to be **disjoint**. Any word $w \in \Sigma^*$ is called a **terminal word**.

Note

Production rules map strings with at least one **nonterminal** to arbitrary other strings

Notation: If we have n **production rule** $h \rightarrow b_i$ sharing a head, we often write $h \rightarrow b_1 \mid \dots \mid b_n$ instead.

Example: A simple **grammar** for english sentences:

$S ::= \text{NP Vi}$
 $\text{NP} ::= \text{Article N}$
 $\text{Article} ::= \text{the} \mid \text{a} \mid \text{an}$
 $\text{N} ::= \text{dog} \mid \text{teacher} \mid \dots$
 $\text{Vi} ::= \text{sleeps} \mid \text{smells} \mid \dots$

Def 1.87. Lexical: A [production rule](#) whose head is a single [nonterminal](#) and whose body consists of a single [terminal](#) is called **lexical** or a **lexical insertion rule**.

Def 1.88. Lexicon of a Grammar: The [subset](#) of [lexical rules](#) of a [grammar](#) G is called the **lexicon** of G .

Def 1.89. Vocabulary: The [set](#) of body symbols are called the vocabulary (or the alphabet).

Def 1.90. Lexical Categories of a Grammar: The [nonterminals](#) that appear in the heads of [lexical](#) rules are called lexical categories of the [grammar](#).

Def 1.91. Structural Rules: The non [lexicon production rules](#) are called structural.

Def 1.92. Phrasal categories: The [nonterminals](#) in the heads of [production rules](#) that expand into other [nonterminals](#) are called phrasal (syntactic) categories.

Def 1.93. G -Derivation: Given a [phrase structure grammar](#) $G := \{N, \Sigma, P, S\}$, we say G **derives** $t \in (\Sigma \cup N)^*$ from $s \in (\Sigma \cup N)^*$ in one step, iff there is a [production rule](#) $p \in P$ with $p = h \rightarrow b$ and there are $u, v \in (\Sigma \cup N)^*$, such that $s = uhv$ and $t = ubv$. We write $s \rightarrow_G^p t$ (or $s \rightarrow_G t$ if p is clear from the context) and use \rightarrow_G^* for the [reflexive transitive](#) closure of \rightarrow_G . We call $s \rightarrow_G^* t$ a G -derivation of t from s .

Def 1.94. Sentential Form: Given a [phrase structure grammar](#) $G := \{N, \Sigma, P, S\}$, we say that $s \in (\Sigma \cup N)^*$ is a **sentential form** of G , iff $S \rightarrow_G^* s$.

Def 1.95. Sentence: A [sentential form](#) that does not contain [nonterminals](#) is called a sentence of G , we also say that G **accepts** s . We say that G **rejects** s , iff it is not a sentence of G .

Def 1.96. Language Generation: The language $L(G)$ of G is the [set](#) of its [sentences](#). We say that $L(G)$ is **generated** by G .

Def 1.97. Equivalent Grammars: We call two [grammars](#) **equivalent**, iff they have the same [languages](#).

Def 1.98. Universal Grammar: A [grammar](#) G is said to be **universal** if $L(G) = \Sigma^*$.

Def 1.99. Syntactic Analysis: Syntactic analysis (parsing / syntax analysis) is the process of analyzing a string of symbols, either in a [formal](#) or a [natural language](#) by means of a [grammar](#).

Note

The shape of the [grammar](#) determines the size of its [language](#)

Def 1.100. Context-Sensitive: We call a [grammar](#) context-sensitive (or type 1), if the bodies of [production rules](#) have no less symbols than the heads.

Def 1.101. Context-Free: We call a [grammar](#) context-free (or type 2), iff the heads have exactly one symbol.

Def 1.102. Regular: We call a [grammar](#) regular (or type 3, or REG), if additionally the bodies are empty or consist of a [nonterminal](#), optionally followed by a [terminal symbol](#).

Note

By extension, a [formal language](#) L is called *context-sensitive* / *context-free* / *regular*, iff it is the [language](#) of a respective [grammar](#). [Context-free grammars](#) are sometimes CFGs and context-free languages CFLs.

1.7 Graphs and Trees

Def 1.103. Undirected Graph: An undirected graph is a pair $\langle V, E \rangle$ such that

- V is a [set](#) of **vertices (nodes)**, and
- $E \subseteq \{ \{v, \bar{v}\} \mid v, \bar{v} \in V \wedge v \neq \bar{v} \}$ is the [set](#) of its **undirected edges**.

Def 1.104. Directed Graph: A directed graph (digraph) is a pair $\langle V, E \rangle$ such that

- V is a [set](#) of **vertices (nodes)**, and
- $E \subseteq V \times V$ is the [set](#) of its **directed edges**.

Def 1.105. Indegree: Given a **directed graph** $\langle V, E \rangle$, the indegree $\text{indeg}(v)$ of a vertex $v \in V$ is defined as $\text{indeg}(v) = |\Gamma(\{w \mid (w, v) \in E\})|$.

Def 1.106. Outdegree: Given a **directed graph** $\langle V, E \rangle$, the outdegree $\text{outdeg}(v)$ (branching factor) of a vertex $v \in V$ is defined as $\text{outdeg}(v) = |\Gamma(\{w \mid (v, w) \in E\})|$.

Note

directed graphs are nothing else than **relations**.

Def 1.107. Initial vs Terminal Node: Let $G = \langle V, E \rangle$ be a **directed graph**, then we call a node $v \in V$

- **initial** (source of G), iff there is no $w \in V$ such that $(w, v) \in E$ (no predecessor)
- **terminal** (sink of G), iff there is no $w \in V$ such that $(v, w) \in E$ (no successor)

Def 1.108. Graph Isomorphism: Iff we can find a **bijection** $\psi : V \rightarrow \bar{V}$ between two graphs $G = \langle V, E \rangle$ and $\bar{G} = \langle \bar{V}, \bar{E} \rangle$ then we call them isomorphic. The **bijection** $\psi : V \rightarrow \bar{V}$ is defined as $(a, b) \in E \Leftrightarrow (\psi(a), \psi(b)) \in \bar{E}$ for **directed graph**. And for **undirected graph** it is defined as $\{a, b\} \in E \Leftrightarrow \{\psi(a), \psi(b)\} \in \bar{E}$

Def 1.109. Equivalent Graphs: Two graphs G and \bar{G} are **equivalent** iff there is an **isomorphism** ψ between G and \bar{G} .

Def 1.110. Labeled Graph: A labeled graph G is a quadruple $\langle V, E, L, l \rangle$ where $\langle V, E \rangle$ is a graph and $l : V \cup E \rightarrow L$ is a **partial function** into a **set** L of labels.

Def 1.111. Paths in Graphs: Given a graph $G := \langle V, E \rangle$ we call a $n + 1$ -tuple $p = \langle v_0, \dots, v_n \rangle \in V^{n+1}$ a **path** in G iff $(v_{i-1}, v_i) \in E$ for all $1 \leq i \leq n$ and $n > 0$.

- We say that v_i are nodes on p and that v_0 and v_n are **linked** by p .
- v_0 and v_n are called the **start** and **end** of p (write $\text{start}(p)$ and $\text{end}(p)$), the other v_i are called **inner nodes** of p .
- n is called the **length** of p (write $\text{len}(p)$).
- We denote the **set** of paths in G with $\Pi(G)$.

Def 1.112. Cyclic Graphs: Given a **directed graph** $G = \langle V, E \rangle$, a **path** p is called **cyclic** iff $\text{start}(p) = \text{end}(p)$. A cycle $\langle v_0, \dots, v_n \rangle$ is called **simple**, iff $v_i \neq v_j$ for $1 \leq i, j \leq n$ with $i \neq j$ (all inner nodes are distinct).

Def 1.113. DAG: A **directed graph** with no **cycles** is called **directed acyclic graph (DAG)**.

Def 1.114. Node Depth: Let $G = \langle V, E \rangle$ be a **directed graph**, then the depth $\text{dp}(v)$ of a vertex $v \in V$ is defined to be 0, iff v is a source of G and the supremum $\sup(\{\text{len}(p) \mid \text{indeg}(\text{start}(p)) = 0 \wedge \text{end}(p) = v\})$ otherwise, i.e. the length of the longest **path** from a source of G to v (can be infinite).

Def 1.115. Graph Depth: Given a **directed graph** $G = \langle V, E \rangle$, the **depth** ($\text{dp}(G)$) of G is defined as the supremum $\sup(\{\text{len}(p) \mid p \in \Pi(G)\})$, i.e. the maximal path length in G .

Def 1.116. Tree: A tree is a **DAG** $G = \langle V, E \rangle$ such that

- There is exactly one **initial node** $v_r \in V$ (called the **root**)
- All nodes but the root have **indegree** of 1.

We call v the **parent** of w , iff $(v, w) \in E$ (w is a child of v). We call a node v a **leaf** of G , iff it is **terminal**, i.e. if it does not have children.

Note

For any node $v \in V$ except the root v_r , there is exactly one **path** $p \in \Pi(G)$ with $\text{start}(p) = v_r$ and $\text{end}(p) = v$.

2 Introduction

Def 2.1. Artificial intelligence (AI): The capability of computational systems to perform tasks typically associated with human intelligence, such as learning, reasoning, problem-solving, perception, and decision-making

Def 2.2. Symbolic AI: A subfield of [Artificial Intelligence \(AI\)](#) based on the assumption that many aspects of intelligence can be achieved by the manipulation of symbols, combining them into meaningful structures (expressions) and manipulating them (using processes) to produce new expressions.

Def 2.3. Statistical AI: Remedies the two shortcomings of [symbolic AI](#) approaches: that all concepts represented by symbols are crisply defined, and that all aspects of the world are knowable/representable in principle. Statistical AI adopts sophisticated mathematical models of uncertainty and uses them to create more accurate world models and reason about them.

Def 2.4. Subsymbolic AI (a.k.a connectionism/neural AI): A subfield of [AI](#) that posits that intelligence is inherently tied to brains, where information is represented by a simple sequence pulses that are processed in parallel via simple calculations realized by neurons, and thus concentrates on neural computing.

Def 2.5. Embodied AI: Posits that intelligence cannot be achieved by [reasoning](#) about the state of the world ([symbolically](#), [statistically](#), or [sub-symbolically](#)), but must be embodied i.e. situated in the world, equipped with a "body" that can interact with it via sensors and actuators. Here, the main method for realizing intelligent behavior is by learning from the world.

Def 2.6. Reasoning: The process of producing valid arguments and predictive world models. There are three forms of reasoning:

- **deductive reasoning** to produce new knowledge from existing knowledge. *Example:* All humans are mortal. Socrates is a human. Therefore, Socrates is mortal.
- **inductive reasoning** to produce knowledge from perception. *Example:* The sun has risen every morning in recorded history. Therefore, the sun will rise tomorrow.
- **abductive reasoning** to produce explanations for observations and given knowledge. *Example:* The grass is wet this morning. If it rained last night, that would explain it. Therefore, it probably rained.

Def 2.7. Inference: The act or process of reaching a **conclusion** about something from known facts or evidence (jointly called **premises**)

Def 2.8. Formal Logic: The science of [deductively](#) valid inferences, meaning arguments whose [conclusions](#) necessarily follow from their [premises](#) by virtue of their form (their structure), regardless of the topic.

Def 2.9. Agent: An agent is a [structure](#) $A := \langle \mathcal{P}, \mathcal{A}, f \rangle$ where:

- \mathcal{P} is a [set](#) of percepts
- \mathcal{A} is a [set](#) of actions
- f is a [function](#) $f : \mathcal{P}^* \rightarrow \mathcal{A}$ that maps from percepts to actions.

In other words, an agent is anything that perceives its environment via sensors and acts on it with actuators.

Def 2.10. Performance Measure: A [function](#) that evaluates a sequence of environments.

Def 2.11. Rational: An [agent](#) is called **rational**, if it chooses whichever action maximizes the expected value of the [performance measure](#) given the percept sequence to date.

Def 2.12. PEAS: To design [rational agents](#), we must specify the PEAS components: [Performance Measure](#), Environment, Actuators, and Sensors.

Def 2.13. Environment Types: For an [agent](#) a we classify the environment e of a by its **type**, which is one of the following. We call e

1. *fully observable*, iff the a 's sensors give it access to the complete state of the environment at any point in time; otherwise, we call it *partially observable*.
2. *deterministic*, iff the next state of the environment is completely determined by the current state and a 's actions; otherwise, *stochastic*.
3. *episodic*, iff a 's experience is divided into atomic episodes, where it perceives and then performs a single action, and the next episode does not depend on previous ones. Otherwise, *sequential*.
4. *dynamic*, iff the environment can change without an action performed by a ; otherwise, *static*. If the environment does not change but a 's performance measure does, we call e *semidynamic*.

5. *discrete*, iff the **sets** of e 's state and a 's actions are countable; otherwise, *continuous*.

6. *single-agent*, iff only a acts on e ; otherwise *multi-agent*.

Def 2.14. Reflex: An **agent** $\langle \mathcal{P}, \mathcal{A}, f \rangle$ is called a **reflex agent**, iff it only takes the last percept into account when choosing an action, i.e. $f(p_1, \dots, p_k) = f(p_k) \forall p_1 \dots p_k \in \mathcal{P}$

Def 2.15. Model-Based Agent: A model-based agent $\langle \mathcal{P}, \mathcal{A}, \mathcal{S}, \mathcal{T}, s_0, S, a \rangle$ is an **agent** $\langle \mathcal{P}, \mathcal{A}, f \rangle$ whose actions depend on:

- a **world model**: a **set** \mathcal{S} of possible *states*, and a start state $s_0 \in \mathcal{S}$.
- a **transition model** \mathcal{T} that predicts a new state $\mathcal{T}(s, a)$ from a state s and an action a .
- a **sensor model** S that given a state s and a percept p determines a new state $S(s, p)$.
- an **action function** $a : \mathcal{S} \rightarrow \mathcal{A}$ that given a state $s \in \mathcal{S}$ selects the next action $a \in \mathcal{A}$.

Note

If the agent is in state s then it took action a and now perceives p , then the agent state will become $s' = S(p, \mathcal{T}(s, a))$ and accordingly take action $a' = a(s')$.

Def 2.16. Goal Based Agent: A goal based agent $\langle \mathcal{P}, \mathcal{A}, \mathcal{S}, \mathcal{T}, s_0, S, a, \mathcal{G} \rangle$ is a **model based agent** with an explicit set of goals. It consists of:

- a set of internal states \mathcal{S} and an initial state $s_0 \in \mathcal{S}$,
- a transition model $\mathcal{T} : \mathcal{S} \times \mathcal{A} \rightarrow \mathcal{S}$,
- a state update function $S : \mathcal{P} \times \mathcal{S} \rightarrow \mathcal{S}$,
- a set of goals \mathcal{G} ,
- a goal conditioned action function $a : \mathcal{S} \times \mathcal{G} \rightarrow \mathcal{A}$ selecting an action given a state and the goals.

Def 2.17. Utility-Based Agent: A utility-based agent uses a world model along with a utility function that models its preferences among the states of that world. It chooses the action that maximizes the expected utility.

Def 2.18. Learning Agent: A learning agent is an **agent** that can improve its own behavior through experience. It is composed of four main components:

- the performance element, which selects actions based on the agent's current percepts and represents its existing knowledge or behavior,
- the learning element, which improves the performance element over time by analyzing feedback and identifying how to make better decisions,
- the critic, which evaluates the agent's performance according to a given performance standard and provides feedback to the learning element,
- the problem generator, which suggests new and informative actions or experiences that help the agent explore and learn more effectively.

Def 2.19. State Representation: We call a state representation **atomic**, iff it has no internal structure (black box). However, iff each state is characterized by attributes and their values then the representation is **factored**. A **structured** state representation is when include representations of objects, their properties and relationships.

3 General Problem Solving

Def 3.1. Search Problem: A search problem $\Pi := \langle \mathcal{S}, \mathcal{A}, \mathcal{T}, \mathcal{I}, \mathcal{G} \rangle$ consists of a **set** \mathcal{S} of states, a **set** \mathcal{A} of actions, and a transition model $\mathcal{T} : \mathcal{A} \times \mathcal{S} \rightarrow \mathcal{P}(\mathcal{S})$ that assigns to any action $a \in \mathcal{A}$ and state $s \in \mathcal{S}$ a **set** of successor states. Certain states in \mathcal{S} are designated as goal (terminal) states ($\mathcal{G} \subseteq \mathcal{S}$ with $\mathcal{G} \neq \emptyset$) and initial states $\mathcal{I} \subseteq \mathcal{S}$.

Def 3.2. Action Application: We say that an action $a \in \mathcal{A}$ is applicable in state $s \in \mathcal{S}$, iff $\mathcal{T}(a, s) \neq \emptyset$ and that any $s' \in \mathcal{T}(a, s)$ is a result of applying action a to state s . We call $\mathcal{T}_a : \mathcal{S} \rightarrow \mathcal{P}(\mathcal{S})$ with $\mathcal{T}_a(s) := \mathcal{T}(a, s)$ the result relation for a and $\mathcal{T}_\mathcal{A} := \bigcup_{a \in \mathcal{A}} \mathcal{T}_a$ the result relation of Π .

Def 3.3. State Space: The graph $\langle \mathcal{S}, \mathcal{T}_\mathcal{A} \rangle$ is called the state space induced by Π .

Def 3.4. Solution: A solution for Π consists of a sequence a_1, \dots, a_n of actions such that for all $1 < i \leq n$:

- a_i is **applicable** to state s_{i-1} where $s_o \in \mathcal{I}$, and
- $s_i \in \mathcal{T}_{a_i}(s_{i-1})$, and $s_n \in \mathcal{G}$

Def 3.5. Cost Function: Often we add a cost function $c : \mathcal{A} \rightarrow \mathbb{R}_0^+$ that associates a step cost $c(a)$ to an action $a \in \mathcal{A}$. The cost of a **solution** is the sum of the step costs of its actions.

Note

- For **deterministic** environments, we have $|\mathcal{T}(a, s)| \leq 1$
- For **fully observable** ones, we have $\mathcal{I} = \{s_0\}$

Def 3.6. Successor Function/State: In a **search problem**, \mathcal{T}_a induces a **partial function** $S_a : \mathcal{S} \rightarrow \mathcal{S}$ whose natural domain is the **set** of states where a is **applicable**: $S_a(s) := s'$ if $\mathcal{T}_a = \{s'\}$ and undefined at s otherwise. We call S_a the **successor function** for a and $S_a(s)$ the **successor state** of s .

Note

- A **search problem** is called a **single-state problem** iff it is **fully observable**, **deterministic**, **static**, and **discrete**.
- A **search problem** is called a **multi-state problem** iff it is **partially observable**.
- A **search problem** is called a **contingency problem** iff the environment is **non deterministic** and the **state space** is unknown.

Def 3.7. Tree Search: Given a **search problem** $\Pi := \langle \mathcal{S}, \mathcal{A}, \mathcal{T}, \mathcal{I}, \mathcal{G} \rangle$, the **tree search algorithm** consists of the simulated exploration of the **state space** $\langle \mathcal{S}, \mathcal{T}_\mathcal{A} \rangle$ in a search **tree** formed by successively expanding already explored states.

Def 3.8. Path Cost: We define the path cost of a node n in a **search tree** T to be the sum of the **step costs** on the **path** from n to the root of T .

The general structure of **Tree Search** algorithms is as follows:

```

TREE_SEARCH(start, INSERT):
    fringe ← empty list
    append start to fringe
    while fringe is not empty:
        node ← remove first element of fringe
        if is_goal(node):
            return node
        children ← expand(node)
        INSERT(fringe, children)
    return failure

```

Def 3.9. Strategy: A strategy is a **function** that picks a node from the fringe of a search tree

Note

Every algorithm has its own **strategy**, hence the implementation of the INSERT differs based on that **strategy**.

Def 3.10. Properties of Strategies:

- completeness: does it always find a solution if one exists?
- optimality: does it always find the optimal solution (least cost)?

- time complexity: number of nodes the algorithm explores (expands)
- space complexity: maximum number of nodes in memory

Note

Time and space complexity measured in terms of:

- b : maximum branching factor of the [search tree](#)
- d : minimal [graph depth](#) of a [solution](#) in the [search tree](#)
- m : maximum [graph depth](#) of the [search tree](#)

Def 3.11. Breadth-First Search (BFS): The BFS [strategy](#) treats the fringe as a FIFO queue, i.e. [successors](#) go in at the end (back) of the fringe.

BFS Properties Analysis

- Time Complexity:
assume we have a binary [search tree](#) ($b = 2$) and we found [solution](#) in [depth](#) = 3. In the 0th level we explored $2^0 = 1$ nodes. In the 1st level, $2^1 = 2$ nodes. In the 2nd level, $2^2 = 4$ nodes. And finally $2^3 = 8$ nodes in the third level. Number of nodes the algorithm explored will be $2^0 + 2^1 + 2^2 + 2^3$.
– Generally, for arbitrary b, d , we have $1 + b + b^2 + \dots + b^d$, which means, worst time complexity is $\mathcal{O}(b^d)$ (exponential in d).
- The space complexity is also $\mathcal{O}(b^d)$.
- [BFS](#) is complete if: (i) b is finite, and (ii) the [state space](#) is finite or has a solution.
- [BFS](#) is optimal if we have a uniform constant cost for all actions.

Def 3.12. Uniform-Cost Search (UCS): UCS is the [strategy](#) where the fringe is ordered by increasing [path cost](#). (*expands least cost first*)

Note

UCS it is equivalent to BFS if all costs are equal

UCS Properties Analysis

- Time and space complexity $\approx \mathcal{O}(b^d)$
- [UCS](#) is complete if:
 - b is finite
 - all actions costs are $\geq \epsilon > 0$
 - [state space](#) is finite or has a solution.

Def 3.13. Depth-First Search (DFS): DFS is the [strategy](#) where the fringe is organized as a LIFO stack, i.e. [successors](#) go in at the front of the fringe.

Def 3.14. Backtracking: Every node that is pushed to the stack is called a **backtrack point**. Popping a non-goal node from the stack and continuing the search with the new top element is called **backtracking**. For this reason, the [DFS algorithm](#) is also referred to as a **backtracking search**.

DFS Properties Analysis

- Time complexity is $\mathcal{O}(b^m)$ (exponential in the maximum depth)
- Space complexity is linear. While [BFS](#) keeps track of all nodes at the current level in memory, [DFS](#) needs to only keep the current path (from root to m), and also needs to keep the remaining unexplored sibling root node for each node along that path (to backtrack). This means, for each level, we store the siblings, i.e. b nodes. We explore until we hit the max depth m and in each level we store b . Hence we have a linear space $\mathcal{O}(b \cdot m)$
- [DFS](#) is complete if the [search tree](#) is [finite](#) and [acyclic](#).
- [DFS](#) is not optimal.

Def 3.15. Depth-Limited Search (DLS): DLS is a [DFS](#) with a depth limit l . We treat all nodes at depth l as if they have no children.

DLS Properties Analysis

- Time complexity $\rightsquigarrow \mathcal{O}(b^l)$
- Space complexity $\rightsquigarrow \mathcal{O}(b \cdot l)$
- Not complete (solution maybe beyond chosen l)
- Not optimal

Def 3.16. Iterative Deepening Search (IDS): IDS is a [DLS](#) with an ever-increasing depth limit. We call the difference between successive depth limits the step size.

Note

- [IDS](#) solves the problem of choosing a good l in [DLS](#).
- [IDS](#) tries all values until either a [solution](#) is found (depth is returned to [DLS](#), we call that the cutoff value) or we return failure.
- [IDS](#) combines benefits of [DFS](#) and [BFS](#).

IDS Properties Analysis

It is similar to [DFS](#) in memory requirements (assuming [finite](#) and [acyclic search tree](#)).

- $\mathcal{O}(b \cdot d)$ when there is a [solution](#)
- $\mathcal{O}(b \cdot m)$ when there is no [solution](#)

It is similar to [BFS](#) in terms of optimality, completeness, and time complexity. Optimal for problems where all actions have the same cost. Complete if b is [finite](#) and the [state space](#) is finite or has a solution.

With regard to time complexity, consider [IDS](#) finds a solution at depth d , then:

- nodes at the very bottom (d) are visited only once in the final iteration.
- nodes at $d - 1$ are visited twice, once in the search with depth $d - 1$ and once in the search with depth d .
- nodes at $d - 2$ are visited three times
- and so on, until root node ($d = 0$) which is visited d times.

Generally, we have $(d)b^1 + (d-1)b^2 + (d-2)b^3 + \dots + b^d$. However, the majority of nodes are at the max depth where there are b^d nodes. Those are only visited once and the extra cost of visiting shallower nodes does not significantly increase the overall count. Therefore, time complexity is bounded by b^d ($\mathcal{O}(b^d)$) because b^d dominates the total nodes count in a large [search space](#).

Def 3.17. Graph Search: A graph search algorithm is a variant of a [tree search algorithm](#) that prunes nodes whose state has already been considered (duplicate pruning), essentially using a [DAG](#) data structure.

The general structure of [Graph Search](#) algorithms is as follows:

```
GRAPH_SEARCH(start, INSERT):
    fringe ← empty list
    visited ← empty set
    append start to fringe
    add state(start) to visited
    while fringe is not empty:
        node ← remove first element of fringe
        if is_goal(node):
            return node
        children ← expand(node)
        for each child in children:
            if state(child) not in visited:
                add state(child) to visited
                INSERT(fringe, child)
    return failure
```


Def 3.18. Search Algorithm: We speak of a **search algorithm** when we do not want to distinguish whether it is a **tree** or a **graph search algorithm** - *difference considered as an implementation detail*.

Def 3.19. Informed: A **search algorithm** is called **informed**, iff it uses some form of external information to guide the search.

Def 3.20. Evaluation Function: An evaluation function assigns a desirability value to each node of the **search tree**. It is not part of the **search problem**, but must be added externally.

Def 3.21. Best-First Search: In best-first search, the fringe is a queue sorted in decreasing order of desirability.

Def 3.22. Heuristic: A heuristic is an **evaluation function** h on states that estimates the **cost** from n to the nearest goal state. We speak of **heuristic search** if the **search algorithm** uses a heuristic in some way.

Def 3.23. Greedy Search: A **best-first search strategy** where the fringe is organized as a queue sorted based on h value.

Greedy Search Properties

- Time and space complexity are both exponential in m (max depth) $\sim \mathcal{O}(b^m)$
- It is not optimal. But it is complete if the **state space** is finite.

Def 3.24. Heuristic Function: A heuristic function for **search problem II** is a **function** $h : \mathcal{S} \rightarrow \mathbb{R}_0^+ \cup \{\infty\}$ so that $h(s) = 0 \quad \forall s \in \mathcal{G}$.

Def 3.25. Goal Distance Function: The **function** $h^* : \mathcal{S} \rightarrow \mathbb{R}_0^+ \cup \{\infty\}$, where $h^*(s)$ is the **cost** of a cheapest **path** from s to a goal state, or ∞ if no such path exists. (*the perfect heuristic which we do not know and cannot compute*)

Def 3.26. Admissible: For a **search problem II** with states \mathcal{S} and actions \mathcal{A} . We say that a **heuristic** h for II is **admissible** if

$$h(s) \leq h^*(s) \quad \forall s \in \mathcal{S}$$

Note

Admissible heuristics never overestimates; our guess should be always optimistic or exact, never too high

Def 3.27. Consistent: For a **search problem II** with states \mathcal{S} and actions \mathcal{A} . We say that a **heuristic** h for II is **consistent** if

$$h(s) - h(s') \leq c(a) \quad \forall s \in \mathcal{S}, a \in \mathcal{A}, s' \in \mathcal{T}(s, a)$$

Note

A **consistent heuristic** never drops faster than the real cost you pay when moving through the state space. Formally, for every transition $s \xrightarrow{a} s'$,

$$h(s) \leq c(a) + h(s')$$

This means the heuristic respects a triangle inequality with respect to the actual **step costs**. **Consistency** ensures that the estimated **total cost** along a **path** never decreases.

Def 3.28. A* Evaluation Function: Given a **path cost** function g and a **heuristic** h , the **A* evaluation function** is

$$f(n) = g(n) + h(n).$$

It estimates the total cost of a cheapest path from the **start state** to a goal via n .

Def 3.29. A* Search: A* search is the **best-first search** strategy that uses the **A* evaluation function** $f = g + h$ to order the fringe.

Theorem 3.1. A* Search with **admissible heuristic** is optimal

Def 3.30. Dominance: Let h_1 and h_2 be two **admissible heuristic**, we say that h_2 **dominates** h_1 if $h_2(s) \geq h_1(s) \quad \forall s \in \mathcal{S}$.

Theorem 3.2. If h_2 **dominates** h_1 , then h_2 is better for **search** than h_1

Proof. If h_2 dominates h_1 , then h_2 is closer to h^* than h_1 □

Def 3.31. Adversarial Search: An adversarial search problem is a [search problem](#) $\langle \mathcal{S}, \mathcal{A}, \mathcal{T}, \mathcal{I}, \mathcal{G} \rangle$, where:

- $\mathcal{S} = \mathcal{S}^{\text{Max}} \uplus \mathcal{S}^{\text{Min}} \uplus \mathcal{G}$
- $\mathcal{A} = \mathcal{A}^{\text{Max}} \uplus \mathcal{A}^{\text{Min}}$
- For $a \in \mathcal{A}^{\text{Max}}$, if $s \xrightarrow{a} s'$, then $s \in \mathcal{S}^{\text{Max}}$ and $s' \in (\mathcal{S}^{\text{Min}} \cup \mathcal{G})$
- For $a \in \mathcal{A}^{\text{Min}}$, if $s \xrightarrow{a} s'$, then $s \in \mathcal{S}^{\text{Min}}$ and $s' \in (\mathcal{S}^{\text{Max}} \cup \mathcal{G})$

together with a **game utility function** $u : \mathcal{G} \rightarrow \mathbb{R}$

Def 3.32. Strategy: Let Θ be an [adversarial search problem](#), and let $X \in \{\text{Max}, \text{Min}\}$. A **strategy** for X is a [function](#) $\sigma^X : \mathcal{S}^X \rightarrow \mathcal{A}^X$ so that $a \in \mathcal{A}^X$ is [applicable](#) to $s \in \mathcal{S}^X$ whenever $\sigma^X(s) = a$.

Def 3.33. Optimal Strategy: A [strategy](#) is called **optimal** if it yields the best possible [utility](#) for X assuming perfect opponent play.

Note

Assumptions for Adversarial Games:

- Two players acting in strictly alternating turns
- Fully observable and deterministic game states
- Finite state space so the game tree is finite
- Terminal states have a real-valued utility
- The game is zero-sum: Max maximizes the utility, Min minimizes the same utility

Def 3.34. Minimax: Let Θ be an [adversarial search problem](#) with [utility function](#) $u : \mathcal{G} \rightarrow \mathbb{R}$. The **minimax value** is the [function](#) $\hat{u} : \mathcal{S} \rightarrow \mathbb{R}$ defined by:

$$\hat{u}(s) = \begin{cases} u(s) & \text{if } s \in \mathcal{G}, \\ \max_{s' \in \mathcal{T}(a,s)} \hat{u}(s') & \text{if } s \in \mathcal{S}^{\text{Max}}, \\ \min_{s' \in \mathcal{T}(a,s)} \hat{u}(s') & \text{if } s \in \mathcal{S}^{\text{Min}}, \end{cases}$$

where a ranges over all actions [applicable](#) to s . The **minimax decision** at a root state r is any action whose [successor state](#) achieves $\hat{u}(r)$.

Def 3.35. Minimax Algorithm: The minimax algorithm is given by the following recursive function whose arguments are a node (state) and whether it is Max's turn or Min's.

```
function MINIMAX(node, maximizing):
    if TERMINAL(node):
        return UTILITY(node)

    if maximizing:
        best = -infinity
        for each child in SUCCESSORS(node):
            val = MINIMAX(child, false)
            if val > best:
                best = val
        return best

    else:
        best = +infinity
        for each child in SUCCESSORS(node):
            val = MINIMAX(child, true)
            if val < best:
                best = val
        return best
```

Def 3.36. Evaluation Function: In practice, full-depth [minimax](#) is infeasible because the [search tree](#) is too large. We therefore impose a fixed *depth limit* / *horizon* d and stop the search at all states $\{s \in \mathcal{S} \mid \text{dp}(s) = d\}$, called *cut-off states*.

An evaluation function is a **function** $f : \mathcal{S} \rightarrow \mathbb{R}$ that assigns a numerical estimate of the true **minimax value** $\hat{u}(s)$ to any nonterminal cut-off state s . If a cut-off state is terminal, its utility $u(s)$ is used instead of $f(s)$.

Def 3.37. Alpha Value: For each node n in a **minimax search tree**, the alpha value $\alpha(n)$ is the highest Max-node utility that search has encountered on its **path** from the **root** to n .

Def 3.38. Beta Value: For each node n in a **minimax search tree**, the beta value $\beta(n)$ is the highest Min-node utility that search has encountered on its **path** from the **root** to n .

Def 3.39. Alpha-Beta Pruning: Alpha-beta pruning is a variant of **minimax** that avoids exploring branches that cannot change the **minimax value**.

- At a Max-node n , if a successor s is found with utility $\hat{u}(s) \geq \beta(n)$, then no remaining successors of n can affect its value and they are pruned.
- At a Min-node n , if a successor s is found with utility $\hat{u}(s) \leq \alpha(n)$, then no remaining successors of n can affect its value and they are pruned.

The algorithm returns the same **minimax value** as full search while expanding fewer nodes.

```
function ALPHABETA(node, alpha, beta, maximizing):
    if TERMINAL(node):
        return UTILITY(node)

    if maximizing:
        best = -infinity
        for each child in SUCCESSORS(node):
            val = ALPHABETA(child, alpha, beta, false)
            if val > best:
                best = val
            if best > alpha:
                alpha = best
            if beta <= alpha:
                break
        return best

    else:
        best = +infinity
        for each child in SUCCESSORS(node):
            val = ALPHABETA(child, alpha, beta, true)
            if val < best:
                best = val
            if best < beta:
                beta = best
            if beta <= alpha:
                break
        return best
```

4 Constraint Satisfaction Problems

Def 4.1. Constraint Satisfaction Problem (CSP): A triple $\gamma := \langle V, D, C \rangle$ where

- V is a **finite set** of variables,
- D is an V -**indexed family** of domains: $(D_v)_{v \in V}$
- C is the **set** of constraints. For a **subset** $\{v_1, \dots, v_k\} \subseteq V$, a constraint $C_{\{v_1, \dots, v_k\}} \subset D_{v_1} \times \dots \times D_{v_k}$

Def 4.2. Variable Assignment: We call a **partial function**:

$$\varphi : V \rightarrow \bigcup_{v \in V} D_v$$

a **variable assignment** if $\varphi(v) \in D_v \quad \forall v \in \text{dom}(\varphi)$

Def 4.3. Satisfying Assignment: a **variable assignment** φ **satisfies** a constraint $C_{\{v_1, \dots, v_k\}}$ iff $(\varphi(v_1), \dots, \varphi(v_k)) \in C_{\{v_1, \dots, v_k\}}$.

Def 4.4. Consistent Assignment: a **variable assignment** φ is called **consistent** iff it **satisfies** all constraints in C .

Def 4.5. Legal Assignment: A value $d \in D_v$ is called **legal** for a variable $v \in V$ iff $v \mapsto d$ is a **consistent assignment**; otherwise, **illegal**.

Def 4.6. Conflicted: A variable with an **illegal value** under **assignment** φ is called **conflicted**.

Def 4.7. CSP Solution: A **variable assignment** that is **total** (i.e. **function**) and **consistent** is a **solution** for the **CSP**.

Def 4.8. Satisfiable: A **CSP** γ is called **satisfiable** iff it has a **solution** (a **total variable assignment** φ that **satisfies** all constraints).

Def 4.9. Discrete: We call a **CSP discrete** iff all of the variables have **countable domains**; otherwise, **continuous**.

Def 4.10. Boolean CSP: A **discrete CSP** is called **boolean** iff $|D_v| = 2 \quad \forall v \in V$.

Note

Discrete CSPs with **domain size** d and n variables has a **search space** of size d^n , so a naive solve (brute-force) is worst case $\mathcal{O}(d^n)$. In general, deciding solvability of a **finite-domain CSP** is NP-complete.

Def 4.11. Constraint Order (Arity): Let $\gamma = \langle V, D, C \rangle$ be a **CSP**. For a constraint

$$C_{\{v_1, \dots, v_k\}} \subseteq D_{v_1} \times \dots \times D_{v_k},$$

its **order** (or *arity*) is

$$\text{ord}(C_{\{v_1, \dots, v_k\}}) := k.$$

The order of γ is

$$\max_{C_{\{v_1, \dots, v_k\}} \in C} k.$$

A constraint of order 1 is *unary*, order 2 is *binary*, and any constraint with order > 2 is *higher order*.

Def 4.12. Binary: A **binary CSP** is a **CSP** where each constraint is **unary** or **binary**.

Def 4.13. Constraint Graph: A **binary CSP** forms a **graph** called the **constraint graph** whose nodes are variables, and whose edges represent the constraints.

Def 4.14. Constraint Network: A constraint network is a **CSP** $\gamma := \langle V, D, C \rangle$ of **order** 2, hence, we write a **unary constraint** as C_v (representing $C_{\{v\}}$) and a **binary constraint** as C_{uv} (representing $C_{\{u, v\}}$). Note that $C_{uv} = C_{vu}$.

Note

The **constraint graph** where all constraints are **binary** is the **undirected graph**:

$$\langle V, \{(u, v) \in V^2 \mid C_{uv} \neq D_u \times D_v\} \rangle$$

CSP as Search

We can induce a **search problem** $\Pi_\gamma := \langle \mathcal{S}_\gamma, \mathcal{A}_\gamma, \mathcal{T}_\gamma, \mathcal{I}_\gamma, \mathcal{G}_\gamma \rangle$ from the **constraint network** $\gamma := \langle V, D, C \rangle$.

We define the **set** of states \mathcal{S}_γ as the **set** of all **partial assignments**:

$$\mathcal{S}_\gamma := \left\{ \varphi : V \rightharpoonup \bigcup_{v \in V} D_v \mid \varphi(v) \in D_v \quad \forall v \in \text{dom}(\varphi) \right\}$$

We define goal states \mathcal{G}_γ as the **set** of **total** and **consistent assignments**

$$\mathcal{G}_\gamma := \left\{ \varphi : V \rightarrow \bigcup_{v \in V} D_v \mid (\varphi(v) \in D_v \quad \wedge \quad (\varphi(u), \varphi(v)) \in C_{uv}) \quad \forall u, v \in V \right\}$$

The initial state is when we have no **assignments**:

$$\mathcal{I}_\gamma := \left\{ \varphi : V \rightharpoonup \bigcup_{v \in V} D_v \mid \text{dom}(\varphi) = \emptyset \right\}$$

Actions represent **assignments** as pairs:

$$\mathcal{A}_\gamma := \{(v, d) \mid v \in V \wedge d \in D_v\}$$

Transition model, given a current state s and an action $a = (v, d)$ gives us the successor state s' .

$$\mathcal{T}_\gamma : \mathcal{A}_\gamma \times \mathcal{S}_\gamma \rightarrow \mathcal{P}(\mathcal{S}_\gamma)$$

Concretely: $\mathcal{T}_\gamma((v, d), s) := \{s'\}$ where s' is the successor state (we are updating the **variable assignment**) which will evaluate as follows:

$$s'(w) := \begin{cases} d, & \text{if } w = v \\ s(w), & \text{if } w \in \text{dom}(s) \text{ and } w \neq v \end{cases} \quad \text{and} \quad \text{dom}(s') := \text{dom}(s) \cup \{v\}.$$

Def 4.15. Backtracking Search: Backtracking search is the basic uninformed algorithm for solving CSPs. It is DFS with two improvements, namely:

1. One variable at a time
 - **variable assignments** are commutative, fixing the order saves us time!
 - i.e. starting from $\text{state}(\text{WA}=\text{red}) \rightarrow \text{state}(\text{WA}=\text{red}, \text{NT}=\text{green})$ is the same as starting from $\text{state}(\text{NT}=\text{green}) \rightarrow \text{state}(\text{NT}=\text{green}, \text{WA}=\text{red})$ ¹
 - We only need to consider **assignments** to a single variable at each step
2. Check constraints as you go
 - Consider only values which do not conflict previous **assignments**
 - That is not a goal test! We can think about it as incremental goal test.

The pseudocode for backtracking search is shown in [Algorithm 1](#)

Algorithm 1 Backtracking Search

```
1: function BACKTRACKING-SEARCH(csp)
2:   return RECURSIVE-BACKTRACKING({}, csp)
3: end function
4: function RECURSIVE-BACKTRACKING(assignment, csp)
5:   if assignment is complete then
6:     return assignment
7:   end if
8:   var  $\leftarrow$  SELECT-UNASSIGNED-VARIABLE(Variables[csp], assignment, csp)
9:   for each value in ORDER-DOMAIN-VALUES(var, assignment, csp) do
10:    if value is consistent with assignment given Constraints[csp] then
11:      add {var = value} to assignment
12:      result  $\leftarrow$  RECURSIVE-BACKTRACKING(assignment, csp)
13:      if result  $\neq$  failure then
14:        return result
15:      end if
16:      remove {var = value} from assignment
17:    end if
18:  end for
19:  return failure
20: end function
```

Note

Note that the exact behaviors of the function: SELECT-UNASSIGNED-VARIABLE and ORDER-DOMAIN-VALUES are not specified yet

Def 4.16. Forward Checking: Forward checking is a filtering (inference) technique that improves the general uninformed backtracking search. It propagates information about illegal values. Whenever a variable $v \in V$ is assigned by $d \in D_v$ ($\varphi(v) = d$), we delete all values that are **inconsistent** with $\varphi(v)$ from every D_u for all variables u connected with v by a constraint.

Note

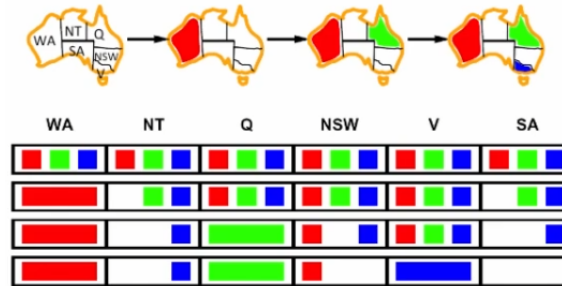
The idea is to keep track of **domains** for unassigned variables and cross off bad options. In **forward checking** we cross off values that violate a constraint when added to the existing assignment (look ahead).

forward checking propagates information from assigned to unassigned variables, but doesn't provide early detection for all failures. For example, (remember Australia map):

- Assume we chose to assign **WA=Red**
- Because **NT**, **SA** are neighbors, we remove **Red** from their domains

¹I am referring to the known Australia map example mentioned in the lecture notes

- Assume we pick next $Q=Green$
- Because NT, SA, NSW are neighbors, we remove **Green** from their domains
- At this point, forward checking thinks we are in good shape, but we are already doomed! *why?* Because NT, SA are neighbors and they both are left with only **Blue** in their domains!
- Assuming, next we choose $V=Green$, we cross off **Green** from NSW, SA, now SA is left with no colors, only then, we know we failed, so we backtrack.

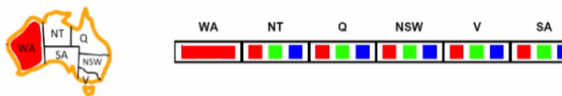


Def 4.17. Arc Consistency: Let $\gamma := \langle V, D, C \rangle$ be a **constraint network**, a variable $u \in V$ is **arc consistent** relative to another variable $v \in V$ if either $C_{uv} \notin C$, or for every value $d \in D_u$ there exist a value $d' \in D_v$ such that $(d, d') \in C_{uv}$. The **constraint network** γ is **arc consistent** if every variable $u \in V$ is **arc consistent** relative to every other variable $v \in V$.

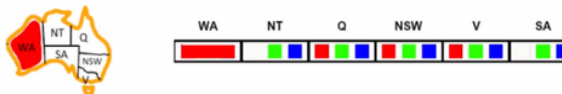
Note

Note that when we are checking some arc $x \rightarrow y$ and find it inconsistent, we remove values from D_x to make it consistent. When doing so, we need to check every other arc $z \rightarrow x$ (where the *tail* of the arc we checked is the *head* of other arcs) even if it was checked before (because now we have fewer options, it is not necessarily consistent anymore!)

Example:



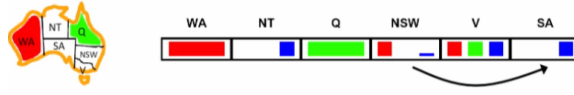
- Checking NT \rightarrow WA \rightsquigarrow removing **Red** from the domain of NT
- Checking SA \rightarrow WA \rightsquigarrow removing **Red** from the domain of SA
- All arcs are consistent now



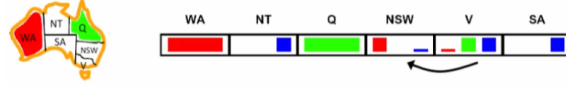
Assume we pick next $Q=Green$, and therefore remove **Green** from NT, NSW, SA:



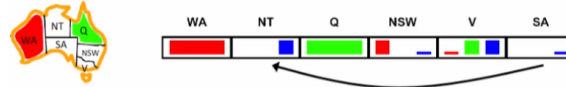
- Looking at V \rightarrow NSW \rightsquigarrow consistent
- Looking at SA \rightarrow NSW \rightsquigarrow consistent
- Looking at NSW \rightarrow SA \rightsquigarrow not consistent! If we choose NSW = **Blue**, nothing I can choose in SA that would not violate the constraint!
- We make NSW \rightarrow SA consistent by removing **Blue** from the domain of NSW



- Now, we check again the arc where NSW was the head.
- Looking at $V \rightarrow NSW \rightsquigarrow$ not consistent



- Looking at $SA \rightarrow NSW \rightsquigarrow$ consistent
- Continue checking arcs. Assume looking at $SA \rightarrow NT \rightsquigarrow$ not consistent



- Now SA domain is empty, we know we failed, so we backtrack!
- Notice that we detect failure earlier than forward checking!

The pseudocode for the arc consistency algorithm (called AC-3) is shown in [Algorithm 2](#)

Algorithm 2 AC-3

```

1: function AC-3(csp)
2:   inputs: csp, a binary CSP with variables  $\{X_1, X_2, \dots, X_n\}$ 
3:   local variables: queue, a queue of arcs, initially all the arcs in csp
4:   while queue is not empty do
5:      $(X_i, X_j) \leftarrow \text{REMOVE-FIRST}(\text{queue})$ 
6:     if REMOVE-INCONSISTENT-VALUES( $(X_i, X_j)$ ) then
7:       for each  $X_k$  in Neighbors[ $X_i$ ] do
8:         add  $(X_k, X_i)$  to queue
9:       end for
10:    end if
11:  end while
12:  return csp
13: end function
14: function REMOVE-INCONSISTENT-VALUES( $(X_i, X_j)$ )
15:   removed  $\leftarrow$  false
16:   for each  $x$  in Domain[ $X_i$ ] do
17:     if no value  $y$  in Domain[ $X_j$ ] allows  $(x, y)$  to satisfy the constraint  $X_i \leftrightarrow X_j$  then
18:       delete  $x$  from Domain[ $X_i$ ]
19:       removed  $\leftarrow$  true
20:     end if
21:   end for
22:   return removed
23: end function

```

- In the worst case, we have a fully connected [constraint network](#) of n variables, hence we will have a total of $n(n-1)$ arcs.
- For one arc $x \rightarrow y$, if the number of values each variable can take is d , to make this arc consistent, we have one variable that has d possibilities (*the tail*) and the *head* too (another d possibilities), and we want to make sure for everything in the *tail* there is some value in the *head* that we can assign. That requires $d \cdot d$ checks in the worst case $= d^2$.
- Moreover, it is not enough to check each arc once. We **enqueue** an already **dequeued** arc whenever we eliminate a value from the domain of the arc's *head*. In the worst case, we might eliminate all the values of a variable, i.e. we put an arc back d times.

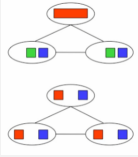
- Therefore, the total run time for AC-3 is $\mathcal{O}((n(n-1)) \cdot d^2 \cdot d) = \mathcal{O}(n^2 d^3)$

Note

Notice that the run time \approx polynomial and yet detecting all possible future problems is NP-Hard because of backtracking. Arc consistency does not guarantee finding a solution. After enforcing arc consistency, we might have:

- one solution left
- multiple solutions left
- no solutions left (and not know it!)

Moreover, the algorithm still runs inside [backtracking search](#).



Def 4.18. Minimum Remaining Values: The MRV heuristic for [backtracking search](#) (a.k.a *Most Constrained Variable First*) selects the unassigned variable with the fewest [legal values](#) given the current partial [assignment](#) φ . Formally, it chooses the variable v that minimizes $|\{d \in D_v \mid \varphi \cup \{v \mapsto d\} \text{ is consistent}\}|$.

Note

MRV \mapsto we want to fail early! Hence, MRV ordering is also called fail-fast ordering.

Def 4.19. Least Constraining Value: Given a variable v , the LCV heuristic chooses the least constraining value for v , i.e. the one that rules out the fewest values in the remaining variables. For the current [partial assignment](#) φ and a chosen variable v , we pick a value $d \in D_v$ that minimizes $|d' \in D_u \mid u \notin \text{dom}(\varphi), C_{uv} \in C, \text{ and } (d', d) \notin C_{uv}|$

Def 4.20. Independent Subproblems: Assume we have independent connected components of a [constraint graph](#). We treat each component as a separate CSP since it has no constraints outside its variables.

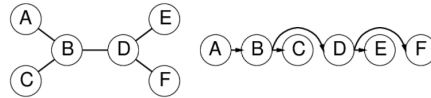
Remember, a naive solver (brute-force) solves a [discrete CSPs](#) with [domain size](#) d and n variables in $\mathcal{O}(d^n)$ (exponential in n). If that same problem has independent subproblems each with c variables, then we have $\frac{n}{c}$ subproblems where each one in the worst case is $\mathcal{O}(d^c)$. So the total time will be $\mathcal{O}(\frac{n}{c} d^c)$, that is, linear in n and exponential in c . That is very fast for small c .

Def 4.21. Decomposition: The process of decomposing a [constraint network](#) into [components](#).

Def 4.22. Tree-Structured CSP: We call a CSP [tree-structured](#) iff its [constraint graph](#) is [acyclic](#).

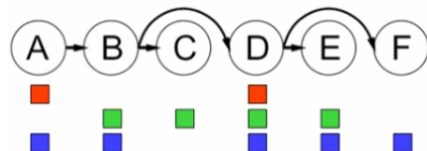
Theorem 4.1. A [Tree-Structured CSP](#) can be solved in $\mathcal{O}(nd^2)$

To solve a [Tree-Structured CSP](#), we choose a variable as [root](#) and order the variables from [root](#) to [leaves](#) such that every node's parent precedes it in the ordering:



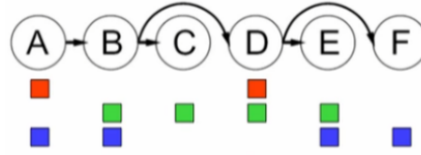
- Backward pass: For $i = n$ and down to 2, apply $\text{Remove-Inconsistent-Values}((\text{Parent}(X_i), X_i))$
- Forward pass: For $i = 1$ and up to n , [assign](#) X_i [consistently](#) with $\text{Parent}(X_i)$

For example, consider the following domains:

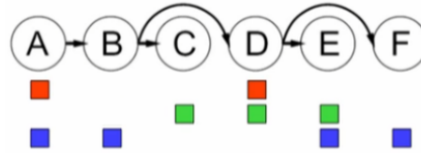


Backward Pass:

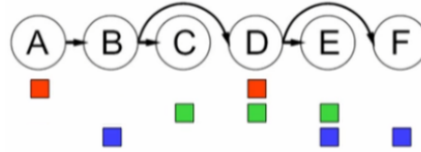
- Start with F , checking $D \rightarrow F \rightsquigarrow$ not consistent:



- Next is E , checking $D \rightarrow E \rightsquigarrow$ consistent.
- Next is D , checking $B \rightarrow D \rightsquigarrow$ consistent.
- Next is C , checking $B \rightarrow C \rightsquigarrow$ not consistent:



- Next is B , checking $A \rightarrow B \rightsquigarrow$ not consistent:



Forward Pass:

$\varphi(A) = \text{Red}$, $\varphi(B) = \text{Blue}$, $\varphi(C) = \text{Green}$, $\varphi(D) = \text{Green}$, $\varphi(E) = \text{Blue}$, $\varphi(F) = \text{Blue}$

Note

After the backward pass on a **tree-structured CSP**, every root-to-leaf arc becomes **consistent**, and no arc ever needs to be enqueued again. The key reason is the direction in which consistency is enforced. Starting from the leaves and moving upward, each arc $x \rightarrow y$ is processed only after all of x 's children have been handled. When enforcing consistency on $x \rightarrow y$, we only remove values from the domain of the tail x . Since all arcs of the form $v \rightarrow x$ were already processed earlier, there will be no later step that removes values from the head y of any previously processed arc. In contrast to general AC algorithms, where pruning the head forces re-checking incoming arcs, the tree structure guarantees that once an arc's head has been finalized, it is never modified again. This is why no arc needs to be re-enqueued: domain reductions always flow upward toward the root, never back down.

Note

If root-to-leaf arcs are **consistent**, the forward pass will not backtrack and will go straight to the solution!

Def 4.23. Conditioning: Instantiate a variable, prune its neighbors' domains

Def 4.24. Cutset conditioning: Instantiate in all ways a set of variables such that the remaining **constraint graph** is a **tree**.

Note

A cutset of size c gives runtime $\mathcal{O}(d^c(n - c)d^2)$:

- Because we are left with a tree of size $n - c$, we have $\mathcal{O}((n - c)d^2)$
- Because we have to look at all possible **assignments** of the cutset, we have $\mathcal{O}(d^c)$

and that is very fast for small c !

5 Formal Systems

Def 5.1. Logical System: A **logical system** (or simply a **logic**) is a triple $\mathcal{S} := \langle \mathcal{L}, \mathcal{M}, \models \rangle$, where

- \mathcal{L} is a **set** of **propositions**
- \mathcal{M} is a **set** of **models**,
- \models is a **relation** $\models \subseteq \mathcal{M} \times \mathcal{L}$ called the **satisfaction relation**. We read $\mathcal{M} \models A$ as \mathcal{M} **satisfies** A and correspondingly $\mathcal{M} \not\models A$ as \mathcal{M} **falsifies** A

Let $\langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system**, $M \in \mathcal{M}$ a model and $A \in \mathcal{L}$ a proposition. Then we say that A is

- **satisfied** by M iff $M \models A$
- **satisfiable** iff A is **satisfied** by some model
- **unsatisfiable** iff A is not **satisfiable**
- **falsified** by M iff $M \not\models A$
- **valid** or **unfalsifiable** (write $\models A$) iff A is **satisfied** by **every** model $M \in \mathcal{M}$
- **invalid** or **falsifiable** (write $\not\models A$) iff A is not **valid**

Def 5.2. Derivation Relation (\vdash): Let \mathcal{L} be a **formal language**. For any **set** of propositions $\mathcal{H} \subseteq \mathcal{L}$ (*called the context or hypotheses*) and any proposition $A \in \mathcal{L}$, we write $\mathcal{H} \vdash A$ to say that A is derivable from \mathcal{H} . We call a **relation** $\vdash \subseteq \mathcal{P}(\mathcal{L}) \times \mathcal{L}$ a **derivation relation** for \mathcal{L} , iff

- $\mathcal{H} \vdash A$, if $A \in \mathcal{H}$ (proof reflexive),
- $\mathcal{H} \vdash A$ and $(\mathcal{H}' \cup \{A\}) \vdash B$ imply $(\mathcal{H} \cup \mathcal{H}') \vdash B$ (proof transitive),
- $\mathcal{H} \vdash A$ and $\mathcal{H} \subseteq \mathcal{H}'$ imply $\mathcal{H}' \vdash A$ (monotonic admits weakening).

Def 5.3. Inference Rules: Let \mathcal{L} be a **formal language**, then an **inference rule** over \mathcal{L} is decidable $n + 1$ -ary **relation** on \mathcal{L} . Inference rules are traditionally written as:

$$\frac{A_1 \quad \dots \quad A_n}{C} \mathcal{N}$$

where A_1, \dots, A_n and C are schemata for words in \mathcal{L} and \mathcal{N} is a name. The A_i are called **assumptions** of \mathcal{N} , and C is called its **conclusion**.

Any $n + 1$ -tuple $\frac{a_1 \quad \dots \quad a_n}{c}$ in \mathcal{N} is called an **application** of \mathcal{N} and we say that we apply \mathcal{N} to a **set** M of words with $a_1, \dots, a_n \in M$ to obtain c .

Def 5.4. Axiom: An **inference rule** without **assumptions** is called an **axiom**

Def 5.5. Calculus: A **calculus** (or **inference system**) is a **formal language** \mathcal{L} equipped with a **set** \mathcal{C} of **inference rules** over \mathcal{L} .

Def 5.6. \mathcal{C} -derivation: Let $\mathcal{S} := \langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system**, and \mathcal{C} a **calculus** for \mathcal{L} , then a \mathcal{C} -derivation of a proposition $P \in \mathcal{L}$ from a **set** $\mathcal{H} \subseteq \mathcal{L}$ of hypotheses (write $\mathcal{H} \vdash_{\mathcal{C}} P$) is a sequence A_1, \dots, A_m of propositions where:

- $A_m = P$,
- for all $1 \leq i \leq m$, either $A_i \in \mathcal{H}$, or,
- there is an **inference rule** $\frac{A_{l_1} \quad \dots \quad A_{l_k}}{A_i} \mathcal{N}$ in \mathcal{C} with $l_j < i$ for all $j \leq k$

We can also see a **\mathcal{C} -derivation** as a **derivation tree**, where A_{l_j} are the children of the node A_i

Def 5.7. Derivation System: We call $\langle \mathcal{L}, \mathcal{M}, \models, \vdash \rangle$ a **derivation system**, iff $\langle \mathcal{L}, \mathcal{M}, \models \rangle$ is a **logical system**, and \vdash a **derivation relation** for \mathcal{L} .

Assertion Let $\langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system**, and \mathcal{C} a **calculus**, then $\vdash_{\mathcal{C}}$ is a **derivation relation** and thus $\langle \mathcal{L}, \mathcal{M}, \models, \vdash_{\mathcal{C}} \rangle$ is a **derivation system**

Def 5.8. Formal System: Let $\mathcal{S} := \langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system**, and let \mathcal{C} be a **calculus** over the language \mathcal{L} . Then the pair $\langle \mathcal{S}, \mathcal{C} \rangle$ is called a **formal system**.

Note

Let \mathcal{C} be a **calculus**, then a **\mathcal{C} -derivation** $\emptyset \vdash_{\mathcal{C}} A$ is called a **proof** of A and if one exists (write $\vdash_{\mathcal{C}} A$) then A is called a **\mathcal{C} -theorem**.

Def 5.9. Admissible: An **inference rule** \mathcal{J} is called **admissible** in a **calculus** \mathcal{C} , if the extension of \mathcal{C} by \mathcal{J} does not yield new **theorems**

Def 5.10. Derived Rules: An **inference rule** $\frac{A_1 \quad \dots \quad A_n}{C} \mathcal{N}$ is called **derived** / **derivable** in a **calculus** \mathcal{C} , if there is a **\mathcal{C} -derivation** $A_1, \dots, A_n \vdash_{\mathcal{C}} C$.

Note

Derivable inference rules are **admissible**, but not the other way around.

Theorem 5.1. It is not true that every **admissible inference rule** is **derivable**

Proof.

1. Construct an empty **calculus** \mathcal{C}_{\emptyset}
2. Since we have no **axioms** and no **inference rules**, then the **set** of **theorems** is also empty $\text{Th}(\mathcal{C}_{\emptyset}) = \emptyset$
3. Let \mathcal{J} be an arbitrary **inference rule**

$$\frac{A_1 \quad \dots \quad A_n}{B} \mathcal{J}$$

4. Now we extend our empty calculus by this rule $\mathcal{C}_{\emptyset} \cup \mathcal{J}$
5. Since we have no **axioms**, the rule \mathcal{J} has no *input* to operate on, therefore, the **set** of **theorems** stays empty $\text{Th}(\mathcal{C}_{\emptyset} \cup \mathcal{J}) = \emptyset$
6. Thus, \mathcal{J} is **admissible**.
7. Since \mathcal{C}_{\emptyset} has no **rule** and no **axioms**, it is impossible to construt steps from A_1, \dots, A_n to B
8. Hence, \mathcal{J} is not **derivable**
9. Since \mathcal{J} is **admissible** but not **derived**, the statement "*every admissible inference rule is derivable*" is False, hence, it is not true that every admissible inference rule is derivable.

□

6 Propositional Logic

Def 6.1. Propositional Logic: Propositional Logic (PL^0) is a **logical system** $\langle \mathcal{L}, \mathcal{M}, \models \rangle$ Where

- the **formal language** \mathcal{L} is defined as the well-formed formulae of propositional logic: $\text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0})$,
- the **set** \mathcal{M} of models is given by the **set** \mathcal{K}_0 of **total functions** $\varphi : \mathcal{V}_0 \rightarrow \mathcal{D}_0$ (i.e., propositional variable **assignments**),
- and $\varphi \models A$ iff $\mathcal{I}_\varphi(A) = T$ (a **proposition** is satisfiable by model φ if the interpretation of it -under that φ - is true)

Hence, propositional logic is a **logical system** $\langle \text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0}), \mathcal{K}_0, \models \rangle$

Def 6.2. Formulae of PL^0 : The **formulae** of PL^0 are made up from

- propositional variables $\mathcal{V}_{\text{PL}^0}$ (or just \mathcal{V}_0): $\mathcal{V}_0 := \{P, Q, R, \dots\}$ (**countably infinite**)
- propositional signature Σ_{PL^0} (or just Σ_0): $\Sigma_0 := \{\top, \perp, \neg, \vee, \wedge, \Rightarrow, \Leftrightarrow, \dots\}$ (called *connectives*)

We define the **set** $\text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0})$ of **well-formed propositional formulae (wffs)** as:

- propositional variables
- logical constants \top and \perp
- negations $\neg A$
- conjunctions $A \wedge B$
- disjunctions $A \vee B$
- implications $A \Rightarrow B$, and
- equivalence (biimplications) $A \Leftrightarrow B$

where $A, B \in \text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0})$ themselves.

Def 6.3. Grammar for PL^0 : The **set** of **propositional formulae** $\{A \mid A \in \text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0})\}$ is given by the following **grammar**:

$$\begin{aligned} X &::= P \mid Q \mid R \mid \dots \quad (\text{propositional variables}) \\ A &::= X \mid \top \mid \perp \mid \neg A \mid A \wedge A \mid A \vee A \mid A \Rightarrow A \mid A \Leftrightarrow A \end{aligned}$$

Def 6.4. Canonical Model: We call a **logical system** $\langle \mathcal{L}, \mathcal{M}, \models \rangle$ a **single-model** with **canonical model** M iff $\mathcal{M} = \{M\}$

Def 6.5. PL^0 Models: A model $\mathcal{M} := \langle \mathcal{D}_0, \mathcal{I} \rangle$ for PL^0 consists of

- the universe $\mathcal{D}_0 := \{T, F\}$ (truth values)
- the interpretation \mathcal{I} that **assigns** truth values to essential connectives
 1. $\mathcal{I}(\neg) : \mathcal{D}_0 \rightarrow \mathcal{D}_0; T \mapsto F, F \mapsto T$
 2. $\mathcal{I}(\wedge) : \mathcal{D}_0 \times \mathcal{D}_0 \rightarrow \mathcal{D}_0; \langle \alpha, \beta \rangle \mapsto T$, iff $\alpha = \beta = T$
 3. $\mathcal{I}(\top) = T$
 4. $\mathcal{I}(\perp) = F$

Note

PL^0 is a single-model **logical system** with **canonical model** $\langle \mathcal{D}_0, \mathcal{I} \rangle$

Def 6.6. Variable Assignment: PL^0 uses a **total variable assignment** $\varphi : \mathcal{V}_0 \rightarrow \mathcal{D}_0$ that assigns truth values to propositional variables.

Def 6.7. Value Function: PL^0 uses a value **function** $\mathcal{I}_\varphi : \text{wff}_0(\Sigma_{\text{PL}^0}, \mathcal{V}_{\text{PL}^0}) \rightarrow \mathcal{D}_0$ that assigns truth values to PL^0 **formulae**. It is defined as follows:

- $\mathcal{I}_\varphi(P) = \varphi(P)$
- $\mathcal{I}_\varphi(\neg A) = \mathcal{I}(\neg)(\mathcal{I}_\varphi(A))$

- $\mathcal{I}_\varphi(A \wedge B) = \mathcal{I}(\wedge)(\mathcal{I}_\varphi(A), \mathcal{I}_\varphi(B))$

Note

Value Function Example: Compute $\mathcal{I}_\varphi(P \vee Q)$ under the assignment $\varphi := \{P \mapsto F, Q \mapsto T\}$

$$\begin{aligned}
\mathcal{I}_\varphi(P \vee Q) &= \mathcal{I}_\varphi(\neg(\neg P \wedge \neg Q)) \\
\mathcal{I}_\varphi(\neg P) &= \mathcal{I}(\neg)(\mathcal{I}_\varphi(P)) \\
\mathcal{I}_\varphi(\neg Q) &= \mathcal{I}(\neg)(\mathcal{I}_\varphi(Q)) \\
\mathcal{I}_\varphi(\neg P \wedge \neg Q) &= \mathcal{I}(\wedge)(\mathcal{I}_\varphi(\neg P), \mathcal{I}_\varphi(\neg Q)) \\
\mathcal{I}_\varphi(P \vee Q) &= \mathcal{I}(\neg)(\mathcal{I}_\varphi(\neg P \wedge \neg Q)) \\
&= \mathcal{I}(\neg)(\mathcal{I}(\wedge)(\mathcal{I}_\varphi(\neg P), \mathcal{I}_\varphi(\neg Q))) \\
&= \mathcal{I}(\neg)(\mathcal{I}(\wedge)(\mathcal{I}(\neg)(\mathcal{I}_\varphi(P)), \mathcal{I}(\neg)(\mathcal{I}_\varphi(Q))))
\end{aligned}$$

We know that

- $\mathcal{I}_\varphi(P) = \varphi(P) = F$, and
- $\mathcal{I}_\varphi(Q) = \varphi(Q) = T$

Hence:

$$\begin{aligned}
&= \mathcal{I}(\neg)(\mathcal{I}(\wedge)(\mathcal{I}(\neg)(F), \mathcal{I}(\neg)(T))) \\
&= \mathcal{I}(\neg)(\mathcal{I}(\wedge)(T, F)) \\
&= \mathcal{I}(\neg)(F) \\
&= T
\end{aligned}$$

Def 6.8. Variable Occurrence: A **function** that maps a **PL⁰ formula** to the **set** of variable occurred in that formula. It is defined as:

- $\text{Var}(P) = \{P\}$
- $\text{Var}(\top) = \emptyset$
- $\text{Var}(\perp) = \emptyset$
- $\text{Var}(\neg A) = \text{Var}(A)$
- $\text{Var}(A \circ B) = \text{Var}(A) \cup \text{Var}(B)$ where $\circ \in \{\wedge, \vee, \Rightarrow, \Leftrightarrow\}$

Def 6.9. Ground Formula: A **PL⁰ formula** A is called **ground** iff $\text{Var}(A) = \emptyset$

Note

Alternative Notation: write $\llbracket A \rrbracket_\varphi$ for $\mathcal{I}_\varphi(A)$ (and $\llbracket A \rrbracket$ if A is **ground**)

Def 6.10. Equivalent: Two **formulae** A and B are called **equivalent**, iff $\mathcal{I}_\varphi(A) = \mathcal{I}_\varphi(B)$ for all **variable assignments** φ

Def 6.11. PL⁰ Identities: Table 3 shows the identities in **PL⁰**

Table 3: **PL⁰ Identities**

Name	for \wedge	for \vee
Idempotence	$\varphi \wedge \varphi = \varphi$	$\varphi \vee \varphi = \varphi$
Identity	$\varphi \wedge \top = \varphi$	$\varphi \vee \perp = \varphi$
Absorption 1	$\varphi \wedge \perp = \perp$	$\varphi \vee \top = \top$
Commutativity	$\varphi \wedge \psi = \psi \wedge \varphi$	$\varphi \vee \psi = \psi \vee \varphi$
Associativity	$\varphi \wedge (\psi \wedge \theta) = (\varphi \wedge \psi) \wedge \theta$	$\varphi \vee (\psi \vee \theta) = (\varphi \vee \psi) \vee \theta$
Distributivity	$\varphi \wedge (\psi \vee \theta) = (\varphi \wedge \psi) \vee (\varphi \wedge \theta)$	$\varphi \vee (\psi \wedge \theta) = (\varphi \vee \psi) \wedge (\varphi \vee \theta)$
Absorption 2	$\varphi \wedge (\varphi \vee \theta) = \varphi$	$\varphi \vee (\varphi \wedge \theta) = \varphi$
De Morgan rule	$\neg(\varphi \wedge \psi) = \neg\varphi \vee \neg\psi$	$\neg(\varphi \vee \psi) = \neg\varphi \wedge \neg\psi$
double negation	$\neg\neg\varphi = \varphi$	
Definitions	$\varphi \Rightarrow \psi = \neg\varphi \vee \psi$	$\varphi \Leftrightarrow \psi = (\varphi \Rightarrow \psi) \wedge (\psi \Rightarrow \varphi)$

Let $\mathcal{M} := \langle \mathcal{D}_0, \mathcal{I} \rangle$ be our **model**, then we say that a **formula** A is

- **true under** φ in \mathcal{M} , iff $\mathcal{I}_\varphi(A) = T$, (write $\mathcal{M} \models^\varphi A$)
- **falsifies** φ in \mathcal{M} , iff $\mathcal{I}_\varphi(A) = F$, (write $\mathcal{M} \not\models^\varphi A$)
- **satisfiable in** \mathcal{M} , iff $\mathcal{I}_\varphi(A) = T$ for some **assignment** φ
- **valid in** \mathcal{M} , iff $\mathcal{M} \models^\varphi A$ for all **variable assignments** φ
- **falsifiable in** \mathcal{M} , iff $\mathcal{I}_\varphi(A) = F$ for some **assignment** φ , and
- **unsatisfiable in** \mathcal{M} , iff $\mathcal{I}_\varphi(A) = F$ for all **variable assignments** φ .

Note

Since PL^0 is a **single-model logical system**, we can omit the explicit reference to the **canonical model** and write $\varphi \models A$ to mean "A is true under φ ", recovering the notation from $\mathcal{M} \models^\varphi A$.

Def 6.12. Entailment: We say that A **entails** B (write $A \models B$), iff $\mathcal{I}_\varphi(B) = T$ for all φ with $\mathcal{I}_\varphi(A) = T$ (i.e. all **variable assignments**) that makes A true also make B true

Theorem 6.1. $A \models B \Rightarrow A \wedge C \models B \wedge C$

Proof.

1. Assume that $A \models B$ (\mathcal{H}_1), we want to show that for all φ that makes $A \wedge C$ true also make $B \wedge C$ true
2. Assume there is some φ that makes $A \wedge C$ true $\rightsquigarrow \mathcal{I}_\varphi(A \wedge C) = T$ (\mathcal{H}_2)
3. This means that $\mathcal{I}_\varphi(A) = \mathcal{I}_\varphi(C) = T$
4. Given \mathcal{H}_1 we now know that our φ also make B true $\rightsquigarrow \mathcal{I}_\varphi(B) = T$
5. From 3 and 4 and by the definition of conjunction: $\mathcal{I}_\varphi(B \wedge C) = T$

□

Def 6.13. Soundness: Let $\mathcal{S} := \langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system** and $\mathcal{H} \subseteq \mathcal{L}$ a **set** of hypotheses, then we call a **calculus** \mathcal{C} over \mathcal{L} **sound** (correct), iff $\mathcal{H} \vdash_{\mathcal{C}} A$ whenever $\mathcal{H} \models_{\mathcal{C}} A$

Def 6.14. Completeness: Let $\mathcal{S} := \langle \mathcal{L}, \mathcal{M}, \models \rangle$ be a **logical system** and $\mathcal{H} \subseteq \mathcal{L}$ a **set** of hypotheses, then we call a **calculus** \mathcal{C} over \mathcal{L} **complete**, iff $\mathcal{H} \vdash_{\mathcal{C}} A$ whenever $\mathcal{H} \models A$

Def 6.15. Hilbert Calculus: The Hilbert calculus \mathcal{H}^0 is a **calculus** that consists of the following **inference rules**:

$$\frac{}{P \Rightarrow (Q \Rightarrow P)} K \quad \frac{}{(P \Rightarrow (Q \Rightarrow R)) \Rightarrow ((P \Rightarrow Q) \Rightarrow (P \Rightarrow R))} S$$

$$\frac{A \Rightarrow B \quad A}{B} \text{MP} \quad \frac{A}{[B/X](A)} \text{Subst}$$

Note

Subst **inference rule** is **admissible** but not **derivable**

Def 6.16. Hilbert Formal System: Using **propositional logic** and the **Hilbert Calculus**, we can build a simple **formal system**:

$$\langle \underbrace{\langle \mathcal{L}, \mathcal{M}, \models \rangle}_{\text{propositional logic}}, \underbrace{\mathcal{H}^0}_{\text{Hilbert Calculus}} \rangle$$

Example: $C \Rightarrow C$ is a \mathcal{H}^0 and here is its **proof**:

Proof. We show that $\emptyset \vdash_{\mathcal{H}^0} C \Rightarrow C$

1. **axiom** S gives us:

$$(P \Rightarrow (Q \Rightarrow R)) \Rightarrow ((P \Rightarrow Q) \Rightarrow (P \Rightarrow R))$$

2. We apply Subst with $[C/P], [C \Rightarrow C/Q], [C/R]$, we get:

$$(C \Rightarrow ((C \Rightarrow C) \Rightarrow C)) \Rightarrow ((C \Rightarrow (C \Rightarrow C)) \Rightarrow (C \Rightarrow C))$$

3. **axiom** K gives us:

$$P \Rightarrow (Q \Rightarrow P)$$

4. We apply Subst with $[C/P], [C \Rightarrow C/Q]$, we get:

$$C \Rightarrow ((C \Rightarrow C) \Rightarrow C)$$

5. We use MP on 4 and 2, we get:

$$(C \Rightarrow (C \Rightarrow C)) \Rightarrow (C \Rightarrow C)$$

6. **axiom** K gives us:

$$P \Rightarrow (Q \Rightarrow P)$$

7. We apply Subst with $[C/P], [C/Q]$, we get:

$$C \Rightarrow (C \Rightarrow C)$$

8. We apply MP on 7 and 5, we get:

$$C \Rightarrow C$$

□