

Ch. 5 Computation

1 Memory

1.1 Definition

Define Memory; a set of elements

$$\mathcal{M} := \{b_1, b_2, \dots, b_M\}$$

1.2 Silicon Computation

In silicon based computation memory is represented with bits 0/1

$$\begin{aligned}\mathcal{M} &:= \{b_1, b_2, \dots, b_m\} \\ b_i &\in \{0, 1\} \quad \forall i\end{aligned}$$

2 Logical Operations

Working list; possibly any set operation/function

+

-

/

exp

\mathcal{C}

\leftarrow

delete

remove

insert

append

if

==

!

$f[X_i] \rightarrow Y_i$

3 Program

3.1 Logical Instructions

Define \mathcal{L} ; an ordered set of logical operations s_i

$$\mathcal{L} := \{s_1, s_2, \dots, s_N\}$$

3.2 State |

Define state; the memory required to perform program P

$$P := \{s_1, s_2, \dots, s_N | b_1, b_2, \dots, b_M\} = \\ \{s_1, s_2, \dots, s_N, b_1, b_2, \dots, b_M\}$$

3.3 Boolean Programs

Define a boolean program; boolean programs can represent functions with inputs x_i and boolean output y_o

$$X = \{x_1, \dots, x_n\} \\ P = P[X] := \{s_1, s_2, \dots, s_N | b_1, b_2, \dots, b_M, X_i, y_o\} = \\ P[X] \rightarrow y_o \in \{\mathbb{T}, \mathbb{F}\}$$

3.4 Void Programs

Define a void program; a program with inputs x_i and no output

$$X = \{x_1, \dots, x_n\} \\ P = P[X] := \{s_1, s_2, \dots, s_N | b_1, b_2, \dots, b_M, X_i\}$$

3.5 Numerical Programs

Define a numerical program; a program with inputs x_i and real, rational output y_o

$$X = \{x_1, \dots, x_n\} \\ P = P[X] := \{s_1, s_2, \dots, s_N | b_1, b_2, \dots, b_M, X_i, y_o\} = \\ P[X] \rightarrow y_o \in \mathbb{Q} \ y_o \geq 0$$

3.6 System Programs

Naming convention to be formalized; a program that outputs one or more elements

$$\begin{aligned} X &= \{x_1, \dots, x_n\} \\ P = P[X] &:= \{s_1, s_2, \dots, s_N \mid b_1, b_2, \dots, b_M, X_i, Y_o\} = \\ P[X] &\rightarrow Y_o = \{y_1, y_2, \dots, y_K\} \end{aligned}$$

3.7 Mathematical Programs

Define a mathematical program; a program with inputs x_i and numerical output y_o

$$\begin{aligned} X &= \{x_1, \dots, x_n\} \\ P = P[X] &:= \{s_1, s_2, \dots, s_N \mid b_1, b_2, \dots, b_M, X_i, y_o\} = \\ P[X] &\rightarrow y_o \in \mathbb{Q} \end{aligned}$$

4 No-op ;

4.1 Definition

Define ";" , the no-op (The C representation of a null statement or no-op)
 $y_o = y_{o2} \forall$; insertions

4.2 Properties of No-op

1. Can be added to any solution S_i and remain a solution for all i anywhere in the order for all j

5 Decision Problems

5.1 Definition

Define decision problem; a function with inputs x_i and boolean output "answer" a_o

$$X_i = \{x_1, \dots, x_n\}$$

$$D := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i$$

6 Solutions

6.1 Definition

Program P is a solution s^+ if P outputs answer a_o for all inputs $X_i \quad \forall i$
 s^+ is a function of the number of inputs n

$$X_i = \{x_1, \dots, x_n\}$$

$$D := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i$$

$$s^+ = s^+[n] := P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i$$

$$P[X_i] = \{s_1, s_2, \dots, s_N | b_1, b_2, \dots, b_M, X_i, y_o\}$$

$$s^+ = P[X_i] = \{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} \quad \forall X_i$$

6.2 Definition of S^+

Define S^+ ; the set of solutions to decision problem D

$$X_i = \{x_1, \dots, x_n\}$$

$$D := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i$$

$$s_j^+ := P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i$$

$$S^+ := \{s_j^+, \dots\} \quad \forall j$$

6.3 Definition of Solvable

Define solvable

$$X_i = \{x_1, \dots, x_n\}$$

$$D := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i$$

$$\begin{aligned} solvable &:= solvable[D] \rightarrow b_o \in \{\mathbb{T}, \mathbb{F}\} = \\ \exists s^+ : s^+ &= P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i \end{aligned}$$

7 The set of all Decision Problems \mathbb{D}

7.1 Definition

Define the set of decision problems \mathbb{D}

$$\begin{aligned} X_i &= \{x_1, \dots, x_n\} \\ D_j &:= f_j[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\ \mathbb{D} &:= \{D_j, \dots\} \quad \forall j \end{aligned}$$

8 Instruction and Memory Notation

Define \mathcal{L} a set of logical operations

Define \mathcal{M} a set of bits "memory"

$$\begin{aligned} P[X_i] \rightarrow y_o &= \{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} \\ \mathcal{L} &:= \{s_1, s_2, \dots, s_{O_T[n]}\} \\ \mathcal{M} &:= \{b_1, b_2, \dots, b_{O_S[n]}\} \\ P[X_i] &= \{\mathcal{L}, \mathcal{M}, X_i, y_o\} \end{aligned}$$

9 Complexity

9.1 Time Complexity of a Decision Problem $O_T[n]$

Define Time Complexity $O_T[n]$ of Decision Problem D with solution s^+

$$\begin{aligned} X_i &= \{x_1, \dots, x_n\} \\ D &:= f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\ s^+ &:= P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i = \\ \{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} &= \{\mathcal{L}, \mathcal{M}, X_i, y_o\} \\ O_T[n] &:= |\mathcal{L}| = N \end{aligned}$$

9.2 Space Complexity $O_S[n]$

Define Space Complexity $O_S[n]$ of Decision Problem D with solution s^+

$$\begin{aligned}
X_i &= \{x_1, \dots, x_n\} \\
D &:= f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\
s^+ &:= P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i = \\
&\{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} = \{\mathcal{L}, \mathcal{M}, X_i, y_o\} \\
O_S[n] &:= |\mathcal{M}| = M
\end{aligned}$$

10 Definition of Complexity

Define Complexity $O[n]$ as a vector of dimension C

$$\mathbf{O}[n] := \langle O_T[n], O_S[n], O_3[n], O_4[n], \dots, O_C[n] \rangle$$

11 Total Complexity

$$O[n] := O_T[n] + O_S[n] + \sum_{i=3}^C O_i[n]$$

12 Simple Computational Complexity

The remainder of this chapter assumes simple computational complexity of dimension 2

12.1 Definition

Define simple computational complexity of dimension 2

$$\mathbf{O}[n] := \langle O_T[n], O_S[n] \rangle$$

12.2 Time Complexity

Restate definition of Time Complexity $O_T[n]$

$$s^+ = \{\mathcal{L}, \mathcal{M}, X_i, y_o\}$$

$$O_T[n] := |\mathcal{L}| = N$$

12.3 Space Complexity

Restate definition of Time Complexity $O_S[n]$

$$s^+ = \{\mathcal{L}, \mathcal{M}, X_i, y_o\}$$

$$O_S[n] := |\mathcal{M}| = M$$

12.4 Total Complexity

$$O[n] := O_T[n] + O_S[n]$$

$$= |\mathcal{L}| + |\mathcal{M}| = N + M$$

13 Optimal Complexity

13.1 Definition

Define Optimal Complexity; the minimum total complexity required to solve a decision problem

$$O_{opt}[n] := \\ \# \hat{O}[n] : \hat{O}[n] < O_{opt}[n] \quad \forall n$$

13.2 Proof of Existence

Prove the existence of at least one $O_{min}[n]$ by induction/contradiction

14 Optimal solution

Define an optimal solution s_{opt}^+

14.1 Definition

$$X_i = \{x_1, \dots, x_n\} \\ D_j := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\ s^+ := P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i \\ s_{opt}^+ := s^+ : \\ \# \hat{O}[n] < O_{opt}[n] \quad \forall n, s^+ \in S_j^+$$

14.2 Optimal Time Complexity Solution

$$X_i = \{x_1, \dots, x_n\} \\ D_j := f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\ s^+ := P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i = \\ \{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} = \{\mathcal{L}, \mathcal{M}, X_i, y_o\} \\ O_T[n] := |\mathcal{L}| = N \\ s_T^+ := s^+ : \\ \# \hat{O}_T[n] < O_T[n] \quad \forall n, s^+ \in S_j^+$$

14.3 Optimal Space Complexity Solution

$$\begin{aligned}
X_i &= \{x_1, \dots, x_n\} \\
D_j &:= f[X_i] \rightarrow a_o \in \{\mathbb{T}, \mathbb{F}\} \quad \forall X_i \\
s^+ &:= P[X_i] \rightarrow y_o : y_o = a_o \quad \forall X_i = \\
&\{s_1, s_2, \dots, s_{O_T[n]}, b_1, b_2, \dots, b_{O_S[n]}, X_i, y_o\} = \{\mathcal{L}, \mathcal{M}, X_i, y_o\} \\
O_S[n] &:= |\mathcal{M}| = M \\
s_S^+ &:= s^+ : \\
\# \hat{O}_S[n] &< O_S[n] \quad \forall n, s^+ \in S_j^+
\end{aligned}$$

15 Polynomial Complexity

15.1 Definition

Decision problem D with solution s^+ has (optimal) total complexity $O[n]$ bounded by polynomial complexity if

$$\begin{aligned}
&\exists K, C, \lambda_1 \dots \lambda_K : \\
O_{opt}[n] &< (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C \quad \forall n
\end{aligned}$$

15.2 Polynomial Problems

Define \mathbb{P} , the set of Decision Problems that can be solved with Polynomial Complexity

$$\begin{aligned}
\mathbb{P} &:= \{D_1, D_2, \dots\} : \\
&\exists K, C, \lambda_1 \dots \lambda_K : \\
O_{opt}[n] &< (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C \quad \forall n, D_j \in \mathbb{P}
\end{aligned}$$

15.3 Polynomial Order of Complexity

MINIMUM K SATISFYING THE CONDITION; Need additional information on K_{opt} ; K_{opt} could be too high as shown

Total complexity $O[n]$ is said to be of order K_{opt}

$$\begin{aligned}
O[n] &\sim K_{opt} \\
O_{opt}[n] &:= O[n] : \\
\# \hat{O}[n] &< O_{opt}[n] \quad \forall n \\
O_{opt}[n] &< (\lambda_{K_{opt}} n)^{K_{opt}} + (\lambda_{K_{opt}-1} n)^{K_{opt}-1} \dots + \lambda_1 n + C \quad \forall n
\end{aligned}$$

16 Corrolary of Optimal Complexity

$$\nexists s^+ \in S^+ :$$

$$O_T[n] < (\lambda_{\hat{K}} n)^{\hat{K}} + (\lambda_{\hat{K}-1} n)^{\hat{K}-1} \dots + \lambda_1 n + C \quad \forall n, \hat{K} < K_{opt}$$

16.1 Proof

Proof by contradiction; definition of optimal complexity

17 Polynomial Time Complexity

17.1 Definition

Decision problem D with (optimal) Time Complexity $O_T[n]$ is bound by polynomial time complexity if

$$\exists K, C, \lambda_1 \dots \lambda_K :$$

$$O_T[n] < (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C \quad \forall n$$

17.2 Polynomial Problems

Define \mathbb{P} , the set of Decision Problems that can be solved with polynomial time complexity

$$\mathbb{P} := \{D_1, D_2, \dots\} :$$

$$\exists K, C, \lambda_1 \dots \lambda_K :$$

$$O_T[n] < (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C \quad \forall n, D_j \in \mathbb{P}$$

17.3 Total Polynomial Complexity Implies Time bounded Polynomial Complexity

Proof Trivial

17.4 Time bounded Polynomial Complexity implies Total Polynomial Complexiy

Proof harder use $O[n] = |L| + |M|$

17.5 Polynomial Time Complexity iff Polynomial Complexity

Follows from previous 2 sections

17.6 Order of Complexity?

Time complexity $O_T[n]$ is said to be of order ?

$$O_T[n] \sim ?$$

18 Theorem of Optimal (Polynomial?) Complexity

Formally define dynamic programming, Optimal complexity minimizes the difference between time and space complexity order

$$D \in \mathbb{P}$$

$$O[n] := O_T[n] + O_S[n]$$

$$O_{opt}[n] := O[n] :$$

$$\nexists \hat{O}[n] < O[n] \quad \forall n$$

$$O_T^+[n] := |\mathcal{L}| = N :$$

$$\nexists \hat{O}_T[n] < O_T^+[n] \quad \forall n$$

$$O_S^+[n] := |\mathcal{M}| = M$$

$$\nexists \hat{O}_S[n] < O_S^+[n] \quad \forall n$$

18.1 Proof

18.2 Even ordered decision problems

18.3 Odd ordered decision problems

19 Theorem of Polynomial Duality

19.1 polynomial in time iff polynomial in space

19.2 $D \in \mathbb{P}$ iff polynomial in Time

19.3 $D \in \mathbb{P}$ iff polynomial in Space

20 Proof of the existence of P

Trivial

21 Definition of Non-Polynomial Problems

Define \mathcal{N} , the set of Decision Problems that cannot be solved with Polynomial Time Complexity

$$\begin{aligned}\mathcal{N} &:= \{D_1, D_2, \dots\} \\ &\quad \nexists K, C, \lambda_1 \dots \lambda_K : \\ O_T[n] &< (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C \quad \forall n, D_j \in \mathcal{N}\end{aligned}$$

22 Divergent Problems

$$\begin{aligned}\mathcal{D} &:= \{D_1, D_2, \dots\} \\ \hat{O}_T[n] &\geq n^n \quad \forall n, D_j \in \mathcal{D}\end{aligned}$$

23 Fundamental Theorem of Computation

Prove no convergent solution exists for Divergent Problems

24 Proof of the existence of \mathcal{N}

Non-trivial; Formalize the traveling salesman problem as a decision problem (any optimization problem)

25 The Traveling Salesman Problem

English description

25.1 Formal Definition

$$X_i = \{c_1, c_2, \dots, c_n\} :$$

$$\dim[c_i] = C > 1$$

$$\bar{X}_i = \{c_1, c_2, \dots, c_n, \bar{P}, \bar{f}[c_i, c_j]\}$$

$$\bar{P} := \{c_k, \dots\} :$$

$$\exists c_k \in \bar{P} \quad \forall c_k \in X_i$$

26 Divergent Problems

Define $\hat{\mathbb{D}}$ the set of decision problems with no convergent?/finite? solution \hat{D}_j

$$\begin{aligned}\hat{\mathbb{D}} &:= \{\hat{D}_j, \dots\} \\ \hat{D}_j &\in \mathbb{D} \quad \forall j \\ \nexists \hat{s}^+ \in S^+ : \hat{s}^+ \text{ solves } \hat{D}_j \quad \forall \hat{D}_j \in \hat{\mathbb{D}} &\iff \\ \nexists \hat{s}^+ \in S^+ : O_j[n] < n^n \quad \forall n, j\end{aligned}$$

There exists no such solution such that $O[n] < n^n$, but there is a right and wrong answer

Either here or in the next chapter we'll prove you can only solve to a certain degree

!!! There exists no such solution such that $O[n] < n^n \quad \forall n$

26.1 Definition

$$\begin{aligned}\hat{O}[n] &:= n^n \\ \lim_{n \rightarrow \infty} \frac{O_{opt}[n]}{\hat{O}[n]} &\text{ diverges} \\ \text{diverges}[O_{opt}[n], \hat{O}[n]] &\rightarrow \mathbb{T}\end{aligned}$$

26.2 Theorem of Divergent Programs

Prove that Divergent implies not in polynomial (trivial)

Prove that Divergent implies Non-polynomial (trial after proving above)

Show that there exists at least one member of Divergent

27 Properties of Solvable and Divergent problems

27.1 Solvable and Divergent are disjoint

Prove by contradiction

28 "Theorem of Divergent Programs"

28.1 Divergence Test

1. Let $d_j \in D$
2. $d_j = (d_j \in \hat{\mathcal{D}}) \cup (d_j \in \text{set of solvable problems})$ by disjoint condition of solvable and divergent
3. Let $O_{opt}[n]$, the optimal complexity of d_j
4. $\rightarrow s_i^+$ that solve d_j have larger complexity $\forall i$
5. 2 implies $O_{opt}[n]$ is either bounded by n^n or not
6. $\hat{O}[n] \equiv n^n$
7. Easy Suppose $d_j \in \text{solvable}$ $\lim_{n \rightarrow \infty} \frac{O_{opt}[n]}{\hat{O}[n]} = 0$
8. Suppose $d_j \in \hat{\mathcal{D}}$ $\lim_{n \rightarrow \infty} \frac{O_{opt}[n]}{\hat{O}[n]} \neq 0$ (by disjoint condition)

$$\lim_{n \rightarrow \infty} \frac{O_{opt}[n]}{\hat{O}[n]} = 1$$

29 Connection to verification in polynomial time

30 Fundamental Theorem of Computation

n^n or $\lambda n^n + C$ the universal bound to solvable computational complexity
 $(\lambda n)^n + C$?

30.1 Time Complexity Argument

Suppose decision problem d with optimal time complexity $O_{T_{min}}[n]$ and solution s^+ , an arbitrary decision problem in P with polynomial complexity

Assumptions

1. $d \in P, s^+ \in S^+$

Assertions

2. $\exists K, C, \lambda_1 \dots \lambda_K : O_{T_{min}}[n] < (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C, \quad \forall n$
3. Define $f[K, C, \lambda_1, \dots, \lambda_K] \equiv (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C$
4. $\exists K, C, \lambda_1 \dots \lambda_K : O_{T_{min}}[n] < f[K, C, \lambda_1, \dots, \lambda_K] \quad \forall n$
5. Let $\hat{s}^+ \equiv \mathcal{C}^n[s^+]$
6. $O_T[n] \leq \hat{O}_T[n]$ (by definition of nested loop)

7. $\hat{O}_{T_{min}}[n] < (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} \dots + \lambda_1 n + C$

8. $\hat{O}_{T_{min}}[n] < \lim_{n \rightarrow \infty} \mathcal{C}^n[s^+]$ (by definition of limit + definition of nested loop, expand to show full derivation, valid because this is a series, probably need to show limit applies)

9. $\therefore O_{T_{min}}[n] < \hat{O}_{T_{min}}[n] < n^n = \lim_{n \rightarrow \infty} \mathcal{C}^n[s^+]$

I want to say for all n but seems refutable for $n = 1, 2, \dots$ but as n approach infinity it's a contradiction to say a solvable problem in P $\hat{O}_{T_{min}} = n^n \quad \forall n$

10. For "sufficiently large n "

$$\nexists \hat{s}^+ \in S^+ : |\hat{s}^+| \equiv O_{T_{min}}[n] < n^n, \quad \forall n$$

$$\hat{O}[n] \equiv n^n$$

30.2 Space Argument

Similar but additional notation required?

31 Divergence Criterion

Necessary condition for divergent program, iff
or you can show there exists no lambda, C such that $O[n]$ is n^n is bounded
by $\lambda n^n + C$ for all n

$$\lim_{n \rightarrow \infty} \text{div} / \text{solvable} > 1$$

Assumptions

1. Define the "Null Space of \mathcal{D} " or "Null Set" O_\perp

$$O_\perp = \{\hat{d}_1, \hat{d}_2, \dots, \hat{d}_j\}, \quad j > 0$$

$$\hat{O}_j[n] \equiv (O[n])^n, \forall j$$

2. $O_P \cup O_N = \mathcal{D}$ (by definition)

Assertions

3. $O_P \cap O_\perp = \emptyset$
4. Let $O_N \cap O_\perp = \hat{O} = \{\hat{O}_i, \dots\}, i > 0$
5. Consider $D_j \in O_N$
6. D_j has finite complexity by definition

$$O_j[n] = C$$

7. D_j has at least one optimal solution by the necessity of optimal solution (theorem Z)

$$O_j[n] = C$$

32 Proof of "P ≠ NP"

32.1 Proof N implies D

Is trivial by implication of Theorem x and Theorem y

$$\begin{aligned}\mathcal{N} &\equiv \{d_j, ..\} \quad \forall j, \mathcal{N} \in D \\ &\quad \nexists K, C, \lambda_1... \lambda_K : \\ O_T[n] &< (\lambda_K n)^K + (\lambda_{K-1} n)^{K-1} ... + \lambda_1 n + C, \quad \forall n, \forall d_j \in \mathcal{N}\end{aligned}$$

1. $\rightarrow \mathcal{P} \cap \mathcal{N} = \emptyset$ by definition of P,N
 2. $d_i \in \hat{D} \vee d_i \in \text{solvable}$
 3. $1 \rightarrow d_i \notin \text{solvable}$
 4. $\therefore d_i \in \hat{D} \quad \forall i$ (theorem y) Show that Definition of Non-Polynomial Problems automatically implies Divergent
1. We've proven Solvable Union are disjoint and complete set P 2. N not in P by definition 3. therefore N in divergence by set theory

Currently we have only defined solvable problems and divergent problems
Additionally polynomial problem which the existence of is trivial
Plus we defined non-polynomial complexity
Prove the existence of \mathcal{N} the set of non polynomial problems

32.2 Proof that D implies N

32.3 D iff N

Show $O[n]$ in the \emptyset the set of problems with $n^n > O[n] > n^k + c$
Proving there's Polynomial and Divergent, in the set of all decision problems

A neat follow up, tie in the definition of \mathcal{N} implies membership to divergent problems

33 Prove the existence of $D = N$, The Traveling Salesman Problem

Define the traveling salesman problem, prove it is divergent and has the same solution as current approaches

Consider proving with both definition and necessary condition

33.1 Compute every sub path or recursive subpaths in memory

Trade off between time and space, $O_{salesman}[n]$ diverges with a polynomial $O_P[n]$

34 Prove Polynomial and Divergent problems are Complements

Implied by the previous sections

35 Solvable Union Divergent = all decision problems

Trivial as a result of the previous section by definition of Ω

$$\mathbb{P} \cup \hat{\mathbb{D}} = \mathbb{D}$$

36 Theorem of Prime Numbers "Riemann Hypothesis"

Riemann Zeta Function

$$\zeta(s) \equiv \sum_{n=1}^{\infty} \frac{1}{n^s} \quad [2]$$

"The prime number theorem determines the average distribution of the primes. The Riemann hypothesis tells us about the deviation from the average. Formulated in Riemann's 1859 paper, it asserts that all the 'non-obvious' zeros of the zeta function are complex numbers with real part 1/2." [2]

Prove the problem is divergent

There fore it can only be proven to a certain degree

The limit as n approaches infinity implies a real part of one half

Connection with the real and imaginary part of $O[n]$

36.1 Prove O_{opt} is performing O_{opt} recursively for the ints less than square root of n

Testing the primes less than \sqrt{n} ? double check

1. Optimal solution for $n=1,2,3$, everything else is a recursive optimal proof by induction

Time Complexity seems to be on the order of $n \log n$... implies divergence or lack of bound? Add in the complexity of division.. probably approaches n^n

36.2 Since divergent, no s^+ exists.. only rules

Express as a limit

36.3 Show that the limit as $n \rightarrow \infty$ implies the real part is 1/2

$1/2 \pm 14.134725 i$ $1/2 \pm 21.022040 i$ $1/2 \pm 25.010858 i$ $1/2 \pm 30.424876 i$
 $1/2 \pm 32.935062 i$ $1/2 \pm 37.586178 i$

$$Z = \zeta(1/2 + it)$$

36.4 Notation, real imaginary parts of the problem

Even numbers and numbers ending in 5 are automatically convergent
Testing numbers ending in 1,3,7,9 results in divergent expression
we can continue to add rules to a certain degree

Citations

- [1] *<https://www.claymath.org/millennium-problems>*
- [2] *https://www.claymath.org/sites/default/files/official_problem_description.pdf*
- [3] *<http://www.math.uchicago.edu/may/VIGRE/VIGRE2011/REUPapers/Riffer-Reinert.pdf>*