Do There Exist Complete Sets for Promise Classes?*

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In this paper we investigate the following two questions:

Q1: Do there exist optimal proof systems for a given language \( L \)?
Q2: Do there exist complete problems for a given promise class \( C \)?

For concrete languages \( L \) (such as TAUT or SAT) and concrete promise classes \( C \) (such as \( \text{NP} \cap \text{coNP} \), UP, BPP, disjoint NP-pairs etc.), these questions have been intensively studied during the last years, and a number of characterizations have been obtained. Here we provide new characterizations for Q1 and Q2 that apply to almost all promise classes \( C \) and languages \( L \), thus creating a unifying framework for the study of these practically relevant questions.

While questions Q1 and Q2 are left open by our results, we show that they receive affirmative answers when a small amount of advice is available in the underlying machine model. For promise classes with promise condition in \( \text{coNP} \), the advice can replaced by a tally \( \text{NP} \)-oracle.

1 Introduction

A general proof system in the sense of Cook and Reckhow [CR79] can be understood as a nondeterministic guess-and-verify algorithm. The question whether there exist optimal or \( p \)-optimal proof systems essentially asks whether there exists a best such verification procedure. For practical purposes, such an optimal proof system would be extremely useful, as both the search for good verification algorithms as well as the quest for lower bounds to the proof size could concentrate on the optimal system. Thus the following question is of great significance:

Q1: Do there exist (\( p \))-optimal proof systems for a given language \( L \)?

Posed by Krajíček and Pudlák [KP89], this question has remained unresolved for more than twenty years. Understanding question Q1 better through characterizations is an important problem with consequences to seemingly unrelated areas such as descriptive complexity: very recently, Chen and Flum [CF10] have shown the equivalence of Q1 to the open problem whether \( L_c \) is a \( P \)-bounded logic for \( P \). Other recent research concentrated on modified versions of Q1, where a number of surprising positive results have been obtained. Cook and Krajíček [CK07] have shown that optimal propositional proof systems exist under non-uniform information (advice), and even one bit of advice suffices. In another direction, Hirsch and Itsykson considered randomized proof systems and showed the existence of an optimal system in the class of all automatizable heuristic proof systems [HI10, Hir10]. Still another positive result was very recently obtained by Pitassi and Santhanam [PS10] who show that there exists an optimal quantified propositional proof system under a weak notion of simulation.

For the original problem Q1, sufficient conditions were established by Krajíček and Pudlák [KP89] by \( \text{NE} = \text{coNE} \) for the existence of optimal and \( \text{E} = \text{NE} \) for \( p \)-optimal propositional proof systems, and these conditions were subsequently weakened by Köbler, Messner, and Torán [KMT03]. Necessary conditions for a positive answer to Q1 are tightly linked to the following analogue of Q1 for promise complexity classes lacking an easy syntactic machine model:

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Q2: Do there exist complete problems for a given promise class C?

Like the first question also Q2 has a long research record, dating back to the 80’s when Kowalczyk [Kow84] and Hartmanis and Hemachandra [HH88] considered this question for \( \text{NP} \cap \text{coNP} \) and UP. This research agenda continues to recent days where, due to cryptographic and proof-theoretic applications, disjoint NP-pairs have been intensively studied (cf. [GSS05, GSSZ04, GSSZ07, Bey07] and [GSZ06] for a survey). Very recently, Itsykson has shown the surprising result that AvgBPP, the average-case version of BPP, has a complete problem [Its09].

As many computational tasks are formulated as function problems [Sel96], it is also interesting to extend Q2 to function classes. In this formulation Q1 becomes a special case of Q2 because all proof systems for a given language can be understood as a promise function class in which complete functions correspond to p-optimal proof systems. In fact, Köbler, Messner, and Torán [KMT03] have shown that, with respect to Q2, proof systems provide the most difficult instances among all promise classes, i.e., a positive answer to Q1 implies a positive answer for Q2 for many choices of \( L \) and \( C \).

In the present paper we continue this line of research. While Köbler, Messner, and Torán [KMT03] focused on the implication \( Q1 \Rightarrow Q2 \), we provide new characterizations for both Q1 and Q2. In fact, from these characterizations we can also easily read off the implication \( Q1 \Rightarrow Q2 \) (under suitable assumptions), thus in addition, we provide alternative proofs for some results of [KMT03]. Köbler, Messner, and Torán used the notion of a test set to measure the complexity of the promise. Here we pursue a different but related approach by representing the promise in a language \( L \) and then using a proof system for \( L \) to verify the promise. On the propositional level, such representations have been successfully used to express the consistency of propositional proof systems (known as the reflection principle, cf. [Coo75, KP89]) or the disjointness of NP-pairs [KP98, Bey07, Bey08]. We create a unifying framework which generalizes these methods to arbitrary languages.

We will now describe in more detail our results and the organization of the paper. After developing the notion of representations in Sections 2 and 3 we examine Q1 in Section 4 where we prove that a language \( L \) has a p-optimal proof system if and only if all polynomial-time computable subsets of \( L \) are recursively enumerable. A similar characterization also holds for the existence of optimal proof systems. This widely generalizes previous results from [Sad02] for propositional proof systems and provides interesting characterizations for a number of applications like the graph isomorphism and automorphism problems.

In Section 5 we proceed with question Q2 where we discuss a characterization of Q2 in terms of uniform enumerations of promise obeying machines. Section 6 then contains our results on the connections between Q1 and Q2. We show that, under suitable assumptions, a promise class \( C \) has complete problems if and only if there exists a proof system for some language \( L \) in which \( C \) is representable. This also yields a general method to show the equivalence of reductions of varying strength with respect to Q2. In addition, we obtain that \( L \) has a p-optimal proof system if and only if every promise class expressible in \( L \) has a complete set or function. Different versions of these results hold for both optimality and p-optimality. We also apply these general theorems to concrete promise classes like UP, \( \text{NP} \cap \text{coNP} \), and disjoint NP-pairs.

In Section 7 we show that the relation between proof systems and promise classes also holds in the presence of advice. Employing recent advances of Cook and Krajíček [CK07] who show that optimal propositional proof systems exist which use only one bit of advice, we obtain hard sets for a large number of promise classes when advice is available. In Section 8 we replace the advice by a very weak oracle and show that promise classes with a coNP-promise (such as disjoint NP-pairs) admit a hard set under a tally NP-oracle. We conclude in Section 9 with a discussion on the status of Q1 and Q2 and future directions for research.

2 Preliminaries

We assume basic familiarity with complexity classes (cf. [BDG88]). Our basic model of computation are polynomial-time Turing machines and transducers. Tacitly we assume these machines to be suitably encoded by strings. We also assume that they always have a polynomial-time clock attached bounding their running time such that this running time is easy to detect from the code of the machine.

For a language \( L \) and a complexity class \( C \), the set of all \( C \)-easy subsets of \( L \) consists of all sets \( A \subseteq L \) with \( A \in C \). A class \( C \) of languages has a recursive \( P \)-presentation (resp. \( N \)-presentation) if there exists a recursively enumerable list \( N_1, N_2, \ldots \) of (non-)deterministic polynomial-time clocked Turing machines such that \( L(N_i) \in C \) for \( i \in \mathbb{N} \), and, conversely, for each \( A \in C \) there exists an index \( i \) with \( A \subseteq L(N_i) \). We remark
that this definition is useful for classes $C$ that do not contain the set of all words (otherwise there always exists the trivial P-presentation $N, N, N, \ldots$ where $N$ accepts all inputs). In this definition, it would also be natural to replace $A \subseteq L(N)$ by the stronger requirement $A = L(N)$, but the weaker concept suffices for our purpose.

### 2.1 Proof Systems

Cook and Reckhow [CR79] defined the notion of a proof system for a language $L$ quite generally as a polynomial-time computable function $f$ with range $L$. A string $w$ with $f(w) = x$ is called an $f$-proof for $x \in L$. By $f \vdash_{\leq m} x$ we indicate that $x$ has an $f$-proof of size $\leq m$. For a subset $A \subseteq L$ we write $f \vdash A$ if there is a polynomial $p$ such that $f \vdash_{\leq p(|x|)} x$ for all $x \in A$.

Proof systems are compared by simulations [CR79, KP89]. If $f$ and $g$ are proof systems for $L$, we say that $g$ simulates $f$ (denoted $f \leq g$), if there exists a polynomial $p$ such that for all $x \in L$ and $f$-proofs $w$ of $x$ there is a $g$-proof $w'$ of $x$ with $|w'| \leq p(|w|)$. If such a proof $w'$ can even be computed from $w$ in polynomial time, we say that $g$ p-simulates $f$ (denoted $f \leq_p g$). A proof system for $L$ is called (p-)optimal if it (p-)simulates all proof systems for $L$.

### 2.2 Promise Classes

Following the approach of Köbler, Messner, and Torán [KMT03], we define promise classes in a very general way. A promise $R$ is described as a binary predicate between nondeterministic polynomial-time Turing machines $N$ and strings $x$, i.e., $R(N,x)$ means that $N$ obeys promise $R$ on input $x$. A machine $N$ is called an $R$-machine if $N$ obeys $R$ on any input $x \in \Sigma^*$. Given a promise predicate $R$, we define the language class $C_R = \{L(N) \mid N$ is an $R$-machine $\}$ and call it the promise class generated by $R$. Instead of $R$-machines we will also speak of $C_R$-machines. Similarly, we define function promise classes by replacing $L(N)$ by the function computed by $N$ (cf. [KMT03]). For functions we use the following variant of many-one reductions (cf. [KMT03]): $f \leq g$ if there exists a polynomial-time computable function $t$ such that $f(x) = g(t(x))$ for all $x$ in the domain of $f$.

In this general framework it is natural to impose further restrictions on promise classes. One assumption which we will make throughout the paper is the presence of universal machines, i.e., we only consider promise conditions $R$ such that there exists a universal machine $U_R$ which, given an $R$-machine $N$, input $x$, and time bound $0^m$, efficiently simulates $N(x)$ for $m$ steps such that $U_R$ obeys promise $R$ on $(N,x,0^m)$.

Occasionally, we will need that C-machines can perform nondeterministic polynomial-time computations without violating the promise. We make this precise via the following notion from [KMT03]: for a complexity class $A$ and a promise class $C$ defined via promise $R$, we say that $A$-assertions are useful for $C$ if for any language $A \in A$ and any nondeterministic polynomial-time Turing machine $N$ the following holds: if $N$ obeys promise $R$ on any $x \in A$, then there exists a language $C \subseteq C$ such that $C \cap A = L(N) \cap A$. A similar definition also applies for function classes. Namely, $A$-assertions are useful for a function class $C$ if for any language $A \in A$ and any polynomial-time clocked Turing transducer $N$ it holds: if $N$ obeys promise $R$ on any input $x \in A$, then there exists a function $f \in C$ such that $N(x) = f(x)$ for any $x \in A$. Throughout this paper we will only consider promise classes $C$ for which $C$-assertions are useful. If also NP-assertions are useful for $C$, then we say that $C$ can use nondeterminism.

The set of all proof systems for a language $L$ is an example for a promise function class, where the promise for a given function $f$ is $\text{rng}(f) = L$. We define a larger class $PS(L)$ where we only concentrate on correctness but not on completeness of proof systems. This is made precise in the following definition.

**Definition 2.1** For a language $L$, the promise function class $PS(L)$ contains all polynomial-time computable functions $f$ with $\text{rng}(f) \subseteq L$.

### 3 Representations

In order to verify a promise, we need appropriate encodings of promise conditions. In the next definition we explain how a promise condition for a machine can be expressed in an arbitrary language.

**Definition 3.1** A promise $R$ is expressible in a language $L$ if there exists a polynomial-time computable function $corr : \Sigma^* \times \Sigma^* \times 0^* \rightarrow \Sigma^*$ such that the following conditions hold:
1. **Correctness:** For every (encoding of a) polynomial-time clocked Turing machine \( N \), for every \( x \in \Sigma^* \) and \( m \in \mathbb{N} \), if \( \text{corr}(x, N, 0^m) \in L \), then \( N \) obeys promise \( R \) on input \( x \).

2. **Completeness:** For every \( R \)-machine \( N \), the set

\[
\text{Correct}(N) = \{ \text{corr}(x, N, 0^p(|x|)) \mid x \in \Sigma^* \}
\]

is a subset of \( L \), where \( p \) is the polynomial time bound (the clock) inscribed in \( N \).

3. **Local recognizability:** For every fixed polynomial-time clocked Turing machine \( N \), the set \( \text{Correct}(N) \) is polynomial-time decidable.

We say that the promise class \( C \) generated by \( R \) is expressible in \( L \) if \( R \) is expressible in \( L \). If the elements \( \text{corr}(x, N, 0^m) \) only depend on \( |x| \), \( N \), and \( m \), but not on \( x \), we say that \( C \) is expressible in \( L \) by a **length-depending** promise.

This definition applies to both language and function promise classes. One of the most important applications for the above concept of expressibility is to choose \( L \) as the set of propositional tautologies \( \text{TAUT} \). Expressing promise conditions by propositional tautologies is a well known approach with a long history. For propositional proof systems, leading to the promise function class \( PS(\text{TAUT}) \), propositional expressions are constructed via the reflection principle of the proof system (cf. [Coo75, KP89]). Propositional expressions have also been used for other promise classes like disjoint \( \text{NP} \)-pairs and its generalizations [Bey07, BKM09]. Typically, these expressions are even length depending. We remark that Köbler, Messner, and Torán [KMT03] have used a related approach, namely the notion of a test set, to measure the complexity of promise conditions.

As a first example, consider the set of all \( \text{P} \)-easy subsets of a language \( L \). The next lemma shows that this promise class is always expressible in \( L \).

**Lemma 3.2** For every language \( L \), the \( \text{P} \)-easy subsets of \( L \) are expressible in \( L \).

**Proof.** Fix an element \( x_0 \notin L \). Then we define the function \( \text{corr}(x, N, 0^m) \) as

\[
\text{corr}(x, N, 0^m) = \begin{cases} 
  x & \text{if } N(x) \text{ accepts in } \leq m \text{ steps} \\
  x_0 & \text{otherwise.}
\end{cases}
\]

Using expressibility of a promise class in a language \( L \), we can verify the promise for a given machine with the help of short proofs in some proof system for \( L \). This leads to the following concept:

**Definition 3.3** Let \( C \) be a promise class which is expressible in a language \( L \). Let further \( A \) be a language from \( C \) and \( P \) be a proof system for \( L \). We say that \( A \) is **representable in** \( P \) if there exists a \( C \)-machine \( N \) for \( A \) such that \( P \vdash_{\text{corr}} \text{Correct}(N) \). If these \( P \)-proofs of \( \text{corr}(x, N, 0^p(|x|)) \) can even be constructed from input \( x \) in polynomial time, then we say that \( A \) is \( p \)-representable in \( P \).

Furthermore, if every language \( A \in C \) is \( (p) \)-representable in \( P \), then we say that \( C \) is \( (p) \)-representable in \( P \).

Intuitively, representability of \( A \) in \( P \) means that we have short \( P \)-proofs of the promise condition of \( A \) (with respect to some \( C \)-machine for \( A \)). Given a proof system \( P \) for \( L \) and a promise class \( C \) which is expressible in \( L \), it makes sense to consider the subclass of all languages or functions from \( C \) which are representable in \( P \). This leads to the following definition:

**Definition 3.4** For a promise class \( C \) expressible in a language \( L \) and a proof system \( P \) for \( L \), let \( C(P) \) denote the class of all \( A \in C \) which are representable in \( P \).

Note that for each \( A \in C \) there exists some proof system \( P \) for \( L \) such that \( A \in C(P) \), but in general \( C(P) \) will be a strict subclass of \( C \) which enlarges for stronger proof systems. It is, of course, interesting to ask whether these subclasses \( C(P) \) have sufficiently good properties. In particular, it is desirable that \( C(P) \) is closed under reductions. Therefore, we make the following definition:

**Definition 3.5** A promise class \( C \) is **provably closed under a reduction** \( \leq_R \) in \( L \) if \( C \) is expressible in \( L \) and for each proof system \( P \) for \( L \) there exists a proof system \( P' \) for \( L \) such that \( P \leq P' \) and for all \( A \in C \) and \( B \in C(P') \), \( A \leq_R B \) implies \( A \in C(P') \).
We remark that provable closure of \( C \) under \( \leq_R \) is a rather weak notion as it does not even imply closure of \( C \) under \( \leq_R \) in the ordinary sense (because of the restriction \( A \in C \)). Also we do not require each subclass \( C(P) \) to be closed under \( \leq_R \), but that for each proof system \( P \) this holds for some stronger system \( P' \). This is a sensible requirement, because proof systems for \( L \) can be defined quite arbitrarily, and closure of \( C(P) \) typically requires additional assumptions on \( P \) (cf. [Bey07] where provable closure of the class of disjoint NP-pairs under different reductions is shown). In fact, it is not difficult to construct counterexamples:

**Proposition 3.6** Let \( C \) be a promise class which is expressible in a language \( L \) and let \( \leq_R \) be a reduction for \( C \). Let further \( P \) be a proof system for \( L \) such that there exist \( A, B \in C \ \setminus C(P) \) with \( A \leq_R B \). Then there exists a proof system \( P' \geq P \) such that \( C(P') \) is not closed under \( \leq_R \).

**Proof.** Under the hypotheses of the proposition, we construct the proof system \( P' \) as follows. We choose a \( C \)-machine \( N \) for \( B \) and define

\[
P'(y) = \begin{cases} x & \text{if } y = 0x \text{ and } x \in \text{Correct}(N) \\ P(x) & \text{if } y = 1x \\ x_0 & \text{otherwise} \end{cases}
\]

where \( x_0 \) is a fixed element from \( L \). As \( P' \models \text{Correct}(N) \) we have \( C(P') = C(P) \cup \{ B \} \). In particular, \( A \) is not contained in \( C(P') \), but \( B \in C(P') \) and \( A \leq_R B \). Therefore \( C(P') \) is not closed under \( \leq_R \).

\[ \square \]

### 4 Optimal Proof Systems and Easy Subsets

In this section we search for characterizations for the existence of optimal or even \( p \)-optimal proof systems for arbitrary languages \( L \) (Question Q1) and apply these results to concrete choices for \( L \). We start with a criterion for the existence of \( p \)-optimal proof systems.

**Theorem 4.1** Let \( L \) be a language such that \( PS(L) \) is expressible in \( L \). Then \( L \) has a \( p \)-optimal proof system if and only if the \( P \)-easy subsets of \( L \) have a recursive \( P \)-presentation.

**Proof.** Let \( f \) be a \( p \)-optimal proof system for \( L \) and let \( A \) be a polynomial-time computable subset of \( L \). We can define a proof system \( f_A \) for \( L \) as follows:

\[
f_A(x) = \begin{cases} f(y) & \text{if } x = 0y \\ a & \text{if } x = 1a \text{ and } a \in A \\ b & \text{otherwise} \end{cases}
\]

where \( b \) is a fixed element in \( L \). Because \( f \) is \( p \)-optimal, \( f_A \) is \( p \)-simulated by \( f \) via some polynomial-time computable function \( t_A \).

As this can be done for all \( P \)-easy subsets \( A \) of \( L \), we get a recursive \( P \)-presentation of \( L \) as follows. Let \( (t_i)_{i \in \mathbb{N}} \) be an enumeration of all deterministic polynomial-time clocked Turing transducers. For \( i \in \mathbb{N} \) consider the following set of algorithms \( M_i \):

1. **Input:** \( x \)
2. **IF** \( f(t_i(1|x)) = x \) **THEN** accept **ELSE** reject

Apparently, these algorithms \( M_i \) can be computed by deterministic polynomial-time Turing machines. Further, each \( M_i \) only accepts inputs from \( L \) because if \( M_i \) accepts \( x \), then we have an \( f \)-proof for \( x \).

Now for each \( P \)-easy subset \( A \) of \( L \), some machine computing the above function \( t_A \) appears in the enumeration \( t_i \), and therefore \( A \) is accepted by \( M_i \) for the appropriate index \( i \) such that \( t_i \) computes \( t_A \). Therefore \( M_i \) is a recursive \( P \)-presentation of the class of all \( P \)-easy subsets of \( L \).

For the converse direction, let \( (M_i)_{i \in \mathbb{N}} \) be a recursive \( P \)-presentation of the \( P \)-easy subsets of \( L \). We construct a \( p \)-optimal proof system \( P_{\text{opt}} \) for \( L \) as follows. Inputs for \( P_{\text{opt}} \) are tuples

\[
\langle \pi, P, 0^m, i, 0^n \rangle.
\]
On such an input, \( P_{opt} \) first checks whether \( P \) is the encoding of a Turing transducer with a polynomial-time bound attached. If this is not the case, then \( P_{opt} \) outputs some fixed element \( x_0 \in L \). Otherwise, \( P_{opt} \) spends \( n \) steps to compute the machine \( M_i \) from the enumeration \( M_1, M_2 \ldots \). If \( n \) steps do not suffice to construct \( M_i \), we output again \( x_0 \in L \). Otherwise, \( P_{opt} \) computes \( \text{corr}(\pi, P, 0^m) \) and checks whether \( M_i \) accepts \( \text{corr}(\pi, P, 0^m) \) in \( n \) steps. Again, if \( M_i \) does not stop in \( \leq n \) steps, then we output \( x_0 \). If \( P \) and \( \pi \) pass the test, then \( P_{opt} \) simulates \( P \) on input \( \pi \) and outputs \( P(\pi) \).

Apparently, \( P_{opt} \) can be computed in polynomial time. For each Turing transducer \( N \) with running time \( p \) and each input \( x \) with \( N(x) \in L \), the element \( \text{corr}(x, N, 0^{|\pi|}) \) is contained in some polynomial-time computable subset of \( L \). Therefore, \( P_{opt} \) is a proof system for \( L \), because by the correctness and completeness conditions from Definition 3.1, the range of \( P_{opt} \) is exactly \( L \).

To prove the \( p \)-optimality of \( P_{opt} \), let \( P \) be a proof system for \( L \). Because by assumption \( PS(L) \) is expressible in \( L \), the set \( \text{Correct}(P) \) is a \( P \)-easy subset of \( L \) (by the local recognizability condition from Definition 3.1). Hence there exists an index \( i \) such that \( M_i \) decides \( \text{Correct}(P) \). Let \( c \) be a constant such that \( M_i \) can be computed from \( i \) in time \( c \) and let \( p \) and \( q \) be polynomial time bounds for \( P \) and \( M_i \), respectively. Then \( P \) is easily seen to be \( p \)-simulated by

\[
\pi \mapsto \langle \pi, P, 0^{|\pi|}, i, 0^{|\text{corr}(\pi, P, 0^{|\pi|})|+c} \rangle
\]

which completes the proof.

By a similar argument we can provide two characterizations for the existence of optimal proof systems.

**Theorem 4.2** Let \( L \) be a language such that \( PS(L) \) is expressible in \( L \). Then the following conditions are equivalent:

1. There exists an optimal proof system for \( L \).
2. The \( \text{NP} \)-easy subsets of \( L \) have a recursive \( \text{NP} \)-presentation.
3. The \( \text{P} \)-easy subsets of \( L \) have a recursive \( \text{NP} \)-presentation.

Given these general results, it is interesting to ask for which languages \( L \) the set \( PS(L) \) of all proof systems for \( L \) is expressible in \( L \). Our next lemma provides sufficient conditions:

**Lemma 4.3** Let \( L \) be a language fulfilling the following two conditions:

1. Natural numbers can be encoded by elements of \( L \), i.e., there exists an injective function \( \text{Num} : \mathbb{N} \to L \) which is both computable and invertible in polynomial time.
2. \( L \) possesses an \( \text{AND} \)-function, i.e., there exists a function \( \text{AND} : \Sigma^* \times \Sigma^* \to \Sigma^* \) which is both polynomial-time computable and polynomial-time invertible such that for all \( x, y \in \Sigma^* \), \( \text{AND}(x, y) \in L \) if and only if \( x \in L \) and \( y \in L \).

Then \( PS(L) \) is expressible in \( L \).

**Proof.** We have to define the function \( \text{corr} \) according to Definition 3.1. Given a string \( x \), an encoding of a polynomial-time computable Turing transducer \( N \), and a number \( m \in \mathbb{N} \), we first simulate \( N(x) \) for \( \leq m \) steps. Let \( y \) be the output of \( N(x) \), if the simulation succeeded. Otherwise, we choose a fixed string \( y \notin L \).

Next we interpret the binary encoding of \( N \) as a natural number (which we again denote by \( N \)) and compute \( \text{Num}(N) \). We then define the function \( \text{corr} \) as

\[
\text{corr}(x, N, 0^m) = \text{AND}(y, \text{AND}(\text{Num}(N), \text{Num}(m)))
\]

Clearly, \( \text{corr} \) is polynomial-time computable. To verify the conditions of Definition 3.1 for \( \text{corr} \), we observe that correctness and completeness of \( \text{corr} \) follow because the string \( w := \text{AND}(\text{Num}(N), \text{Num}(m)) \) is contained in \( L \) for all \( N, m \in \mathbb{N} \), and therefore \( \text{AND}(y, w) \in L \) if and only if \( y \in L \).

Local recognizability for \( \text{corr} \) follows as \( \text{AND} \) and \( \text{Num} \) are invertible in polynomial time and therefore for each polynomial-time Turing transducer \( N \), the set \( \text{Correct}(N) \) is in \( P \).
Using this lemma we can show L-expressibility of \( PS(L) \) for many interesting choices of \( L \):

**Proposition 4.4** For any of the following languages \( L \), the set \( PS(L) \) is expressible in \( L \):

- \( SAT_i \) for \( i \in \mathbb{N} \) (the satisfiability problem for quantified propositional formulas with \( i \) quantifier alternations, starting with existential quantifiers),
- \( TAUT_i \) for \( i \in \mathbb{N} \) (quantified propositional tautologies with \( i \) quantifier alternations, starting with universal quantifiers),
- \( QBF \) (quantified propositional tautologies),
- the graph isomorphism problem \( GI \), its complement \( \overline{GI} \), and the complement \( \overline{GA} \) of the graph automorphism problem.

**Proof.** We have to check the conditions from the previous lemma. For languages consisting of formulas like \( SAT_i \), \( TAUT_i \), or \( QBF \), the AND-function is provided by the Boolean connective \( \land \). The function \( Num \) can be defined for example by

\[
    n \mapsto (p_{\lceil \log n-1 \rceil} \lor \neg p_{\lceil \log n+1 \rceil}) \land \cdots \land (p_{n_1} \lor \neg p_{n_1}),
\]

where \( p_0, p_1 \) are fixed propositional variables and \( n_{\lceil \log n+1 \rceil} \ldots n_1 \) are the bits of the binary representation of \( n \).

For \( GI \), \( Num(n) \) can be implemented by pairs \((K_i, K_{i+1})\) of cliques of size \( i \) which indicate the positions \( i \) where the binary representation of \( n \) contains a 1. For \( GI \) we take \((K_i, K_{i+1})\), and for \( GA \) we take an easy rigid graphs with \( i \) vertices. It is well known that \( GI \) has an AND-function (cf. [KST93]). For the AND-functions of \( GI \) and \( GA \) we can take the OR-functions of \( GI \) and \( GA \) (cf. [KST93]).

For \( GI \), which like any problem in \( NP \) has an optimal proof system, we obtain the following characterization on the existence of a p-optimal proof system.

**Corollary 4.5** \( GI \) has a p-optimal proof system if and only if there exists a recursive \( P \)-presentation of all polynomial-time computable subsets of \( GI \).

Let us remark that in Lemma 4.3, instead of an AND-function we could also use a padding function for \( L \). In this way we obtain a similar result as Corollary 4.5 for \( GA \) (which is not known to possess an AND-function).

## 5 Complete Sets and Enumerations

In this section we consider the question Q2, asking whether language or function promise classes have complete sets or functions. There is a long history of equating complete sets and recursive enumerations of machines. The following result essentially stems from [KMT03], but particular cases of the theorem have been been previously obtained, namely for \( NP \land coNP \) by Kowalczyk [Kow84], for \( UP \) by Hartmans and Hemachandra [HH88], and, more recently, for disjoint \( NP \)-pairs by Glaßer, Selman, and Sengupta [GSS05]. We just formulate the theorem for language classes, but a similar result also holds for promise function classes. The theorem is already included in [KMT03], but for the benefit of the reader we include a full proof (our argument is more direct than the proof given in [KMT03]).

**Theorem 5.1** (Köbler, Messner, Torán [KMT03]) Let \( C \) be a promise class which is closed under many-one reductions. Then \( C \) has a many-one complete problem if and only if there exists a recursive enumeration \((N_i)_{i \geq 1}\) of polynomial-time clocked Turing machines such that \( C = \{L(N_i) \mid i \geq 1\} \) (a uniform enumeration of \( C \)).

**Proof.** For the proof of the forward implication, let \( A_C \) be a many-one complete problem for \( C \) and let \( M \) be a \( C \)-machine for \( A_C \). We fix an enumeration \( N_1, N_2, \ldots \) of all polynomial-time Turing transducers (by our convention from Section 2 all Turing transducers are clocked). By assumption, the class \( C \) is closed under many-one reductions, and hence \((M \circ N_i)_{i \geq 1}\) is a uniform enumeration of \( C \).

For the converse implication, let \( f \) be a recursive function computing an enumeration \((N_i)_{i \in \mathbb{N}}\) of \( C \)-machines such that \( C = \{L(N_i) \mid i \geq 1\} \). We construct a many-one complete problem \( A_C \) for \( C \) as follows: elements of \( A_C \) are of the form

\[
    \langle x, i, 0^m \rangle.
\]
On such an input, \(A_C\) first simulates \(f\) on input \(i\) for at most \(m\) steps. If this simulation does not terminate, then we reject. Otherwise, \(f(i)\) is guaranteed to output a \(C\)-machine \(N_i\). Now we simulate this machine \(N_i\) on input \(x\) for at most \(m\) steps. Again, if the simulation does not stop, we reject. Otherwise, \(A_C\) answers according to the answer obtained in the simulation of \(N_i(x)\).

By our general assumption on promise classes we have a universal \(C\)-machine, and therefore \(A_C \in C\). To prove the hardness of \(A_C\) for \(C\), let \(A \in C\). Then there exists a \(C\)-machine \(N_i\) in the enumeration computed by \(f\) such that \(N_i\) accepts \(A\). Let \(p\) be a polynomial bounding the running time of \(N_i\) and let \(c\) be the time that \(f(i)\) spends to compute \(N_i\). Then the polynomial-time computable function

\[
x \mapsto (x, i, 0^{p(|x|)}+c)
\]

many-one reduces \(A\) to \(A_C\).

Let us note that in the proof of the forward implication of Theorem 5.1, the hypothesis that \(C\) is closed under many-one reductions seems indeed crucial. Namely, if \(C\) consists of all \(P\)-easy subsets of \(TAUT\), then \(C\) trivially contains a many-one complete set. On the other hand, a recursive enumeration of \(C\)-machines as in Theorem 5.1 is rather unlikely to exist, as this would imply the existence of a \(p\)-optimal propositional proof system by Theorem 4.1. But of course, the \(P\)-easy subsets of \(TAUT\) are not closed under many-one reductions.

### 6 Optimal Proof Systems and Complete Sets

Now we are ready to analyse the relations between our central questions \(Q1\) and \(Q2\) on the existence of optimal proof systems for languages \(L\) and the existence of complete sets for promise classes \(C\). While Köbler, Messner, and Torán [KMT03] have shown that for many natural choices of \(L\) and \(C\), a positive answer to \(Q1\) implies a positive answer to \(Q2\), we will provide here a number of characterizations involving both questions. In particular, these characterizations will also yield the above mentioned relation between \(Q1\) and \(Q2\) for concrete applications.

Our first result characterizes the existence of complete sets for a promise class \(C\) by the representability of \(C\) in a proof system.

**Theorem 6.1** Let \(C\) be a promise language (or function) class which can use nondeterminism and let \(L\) be a language such that \(C\) is provably closed under many-one reductions in \(L\). Then \(C\) has a many-one complete language (or function) if and only if there exists a proof system for \(L\) in which \(C\) is representable.

**Proof.** For the proof of the forward implication, let \(C\) be a promise complexity class with a many-one complete language \(A\). Let \(L\) be a language such that \(C\) is provably closed under reductions in \(L\). Let \(P\) be an arbitrary proof system for \(L\) and let \(N\) be a \(C\)-machine for \(A\). We construct a proof system \(P'\) with \(A \in C(P')\) as follows:

\[
P'(y) = \begin{cases} 
  x & \text{if } y = 0x \text{ and } x \in \text{Correct}(N) \\ 
  P(x) & \text{if } y = 1x \\ 
  x_0 & \text{otherwise}
\end{cases}
\]

where \(x_0\) is a fixed element from \(L\). Because \(C\) is provably closed under reductions in \(L\), there exists a proof system \(P''\) for \(L\) such that \(C(P'')\) is closed under many-one reductions. As \(A \in C(P'')\) and \(A\) is many-one complete for \(C\), it follows that \(C(P'') = C\).

For the proof of the converse implication, let \(P\) be a proof system for \(L\) in which every language \(A \in C\) is representable. We construct a complete set \(A_C\) for \(C\) by specifying a \(C\)-machine \(M_A\) accepting \(A_C\). Elements of \(A_C\) are of the form

\[
\langle x, N, 0^m, 0^n \rangle
\]

On such inputs \(M_A\) performs the following operations. \(M_A\) guesses a string \(\pi \in \Sigma^{\leq n}\) and checks whether \(P(\pi) = \text{corr}(x, N, 0^n)\). At this point we need that \(C\) can use nondeterminism. If this test fails, then \(M_A\) rejects the input. Otherwise, \(M_A\) simulates the machine \(N\) on input \(x\) for \(m\) steps and answers according to the answer obtained in this simulation. If the simulation does not terminate in \(m\) steps, \(M_A\) rejects the input.
Because $C$ can use nondeterminism and has a universal machine, $M_A$ is a C-machine and hence $A_C \in C$. To verify the hardness of $A_C$ for $C$, let $A$ be a language from $C$. Because $A$ is representable in $P$, there exists a C-machine $N'$ accepting $A$ such that $P \vdash$, $\text{Correct}(N')$. Let $p$ and $q$ be polynomials bounding the running time of $N$ and the proof size of $\text{Correct}(N)$ in $P$, respectively. Then the polynomial-time computable function

$$x \mapsto \langle x, N, 0^p(|x|), 0^q(|\text{corr}(x, N, 0^p(|x|))|) \rangle$$

many-one reduces $A$ to $A_C$. 

Without making the assumption that the promise classes can use nondeterminism we obtain the following result:

**Theorem 6.2** Let $C$ be a promise language (or function) class which is closed under many-one reductions and let $L$ be a language such that $C$ is expressible in $L$. Then $C$ has a many-one complete language (or function) if and only if $L$ has a proof system in which $C$ is $p$-representable.

**Proof.** For the forward direction, assume that $C$ has a many-one complete language (or function). It follows from Theorem 5.1 that $C$ has a uniform enumeration $\{N_i\}_{i \in \mathbb{N}}$ of C-machines with polynomial running times $p_{N_i}$. Let $G$ be a Turing machine generating the codes of the machines $N_1, N_2, \ldots$

We say that a string $v \in \Sigma^*$ is in good form if

$$v = \langle w_G, x, 0^{p_{N_i}(|x|)} \rangle$$

where $x \in \Sigma^*$ and $w_G$ is a computation of the machine $G$ eventually producing the code of the C-machine $N_i$. Apparently, we can check in polynomial time whether a given string is in good form.

We define $g : \Sigma^* \rightarrow L$ in the following way. If $v = \langle w_G, x, 0^{p_{N_i}(|x|)} \rangle$ is in good form, then $g(v) = \text{corr}(x, N_i, 0^{p_{N_i}(|x|)})$ (where $N_i$ is the machine produced by $G$ during the computation $w_G$), otherwise $g(v) = 0v$, where $0v$ is a certain fixed string from $L$.

This polynomial-time computable function can be extended to a proof system $P$ for $L$ in which $C$ is $p$-representable. Let $P'$ be any proof system for $L$ ($L$ is recursively enumerable). We define the proof system $P$ as follows:

$$P(y) = \begin{cases} g(v) & \text{if } y = 0v \\ P'(v) & \text{if } y = 1v. \end{cases}$$

Let $A$ be any language from $C$. There exists a machine $N_i$ from the uniform enumeration of the class $C$ such that $L(N_i) = A$, and $p_{N_i}$ is its polynomial-time bound. Let $w_G$ be the computation of $G$ producing the code of the machine $N_i$. The function $\lambda(x) = 0\langle w_G, x, 0^{p_{N_i}(|x|)} \rangle$ produces $P$-proofs of $\text{Correct}(N_i)$ in polynomial-time in $|x|$.

For the converse direction, assume that there exists a proof system $P$ for $L$ such that $C$ is $p$-representable in it. Let $R$ be the promise condition for $C$. Consider the language

$$A_C = \{ \langle x, N, 0^{p_N(|x|)}, w \rangle \mid x \in L(N), P(w) = \text{corr}(x, N, 0^{p_N(|x|)}) \}$$

where $N$ is a nondeterministic Turing machine with polynomial time bound $p_N$ (the clock) and $w$ is a $P$-proof of $\text{corr}(x, N, 0^{p_N(|x|)})$. We claim that $A_C$ is the desired C-complete language.

Let us first argue that $A_C \in C$. We say that a string $v$ is in good form if and only if $v = \langle x, N, 0^{p_N(|x|)}, w \rangle$ and $P(w) = \text{corr}(x, N, 0^{p_N(|x|)})$, where $x, N, p_N$, and $w$ mean the same as above. Let us notice that if $v$ is in good form, then $\text{corr}(x, N, 0^{p_N(|x|)}) \in L$. From the correctness condition it follows that $N$ obeys promise $R$ on input $x$.

There exists a polynomial-time Turing machine $V$ which verifies whether a given input string $v$ is in good form. By our general assumption on promise classes we have also a universal machine $U$ with respect to the promise $R$. Using the machines $V$ and $U$ we can construct an $R$-machine $K$ accepting $A_C$. The machine $K$, first runs $V$ on input $v$, checking if $v$ is in good form. If this test fails, then $K$ rejects $v$, otherwise $K$ runs $U$ on input $\langle N, x, 0^{p_N(|x|)} \rangle$. Since $U$ is a universal machine for $C$, the machine $K$ is an $R$-machine.
To prove hardness of $A_C$ for $C$, let $A$ be any language in $C$. Since $A$ is p-representable in $P$, there exists a $C$-machine $N$ such that $L(N) = A$, and the $P$-proofs of $\text{Correct}(N)$ can be constructed in polynomial time in $|x|$. The function $f : \Sigma^* \rightarrow \Sigma^*$ defined by

$$f(x) = (x, N, 0^{p_N(|x|)}, w)$$

performs a polynomial-time many-one reduction from $A$ to $A_C$. The string $w$ from the definition of $f$ is the $P$-proof of $\text{corr}(x, N, 0^{p_N(|x|)})$.

Let us mention some applications of this result. The promise class $\text{DisjNP}$ of disjoint $\text{NP}$-pairs and the class $\text{UP}$ are expressible in $\text{TAUT}$, and the class $\text{NP} \cap \text{coNP}$ is expressible in $\text{QBF}$ (cf. [Bey07, KMT03, Sad97, Sad08]). Hence we obtain the following corollary exemplifying our theorem.

**Corollary 6.3**

1. Complete disjoint $\text{NP}$-pairs exist if and only if $\text{TAUT}$ has a proof system in which $\text{DisjNP}$ is p-representable (if and only if $\text{TAUT}$ has a proof system in which $\text{DisjNP}$ is representable).

2. $\text{UP}$ has a complete language if and only if $\text{TAUT}$ has a proof system in which $\text{UP}$ is p-representable.

3. $\text{NP} \cap \text{coNP}$ has a complete language if and only if $\text{QBF}$ has a proof system in which $\text{NP} \cap \text{coNP}$ is p-representable.

**Theorem 6.1** also allows to derive results which show that the question of the existence of complete problems for $C$ does not depend on the strength of the underlying reduction. This can be done as in the following corollary:

**Corollary 6.4** Let $\leq$ and $\leq'$ be two reductions which are refined by many-one reductions. Assume further that $C$ can use nondeterminism and is both provably closed under $\leq$ and $\leq'$ in some language $L$. Then $C$ has a $\leq$-complete problem if and only if $C$ has a $\leq'$-complete problem.

**Proof.** It suffices to show that $C$ contains a many-one complete problem if and only if $C$ contains a $\leq$-complete problem. The forward implication is clear as $\leq$ is refined by many-one reductions.

Conversely, if $C$ has a $\leq$-complete problem and $C$ is provably closed in $L$ under $\leq$, then we obtain a proof system $P$ for $L$ with $C(P) = C$ as in the proof of the forward implication of Theorem 6.1. Now the converse implication of Theorem 6.1 together with the assumption that $C$ can use nondeterminism gives us a many-one complete problem for $C$.

In this way it can be shown, for example, that the question of the existence of complete disjoint $\text{NP}$-pairs is equivalent for reductions ranging from strong many-one reductions to smart Turing reductions (cf. [GSS05, Bey07]).

Our next result shows that question Q1 on the existence of p-optimal proof systems for a language $L$ can be characterized by a “universally quantified” version of the condition from Theorem 6.2. Further, Q1 is even equivalent to the existence of complete sets for all promise classes representable in $L$:

**Theorem 6.5** Let $L$ be a language such that $P^S(L)$ is expressible in $L$. Then the following conditions are equivalent:

1. There exists a p-optimal proof system for $L$.

2. There exists a proof system for $L$ in which any promise class which is expressible in $L$ is p-representable.

3. There exists a proof system for $L$ in which the class of all $P$-easy subsets of $L$ is p-representable.

4. Every promise language and function class which is expressible in $L$ has a many-one complete language or function.
**Proof.** The proof is structured into the implications $1 \Rightarrow 2 \Rightarrow 3 \Rightarrow 1$ and $2 \Rightarrow 4 \Rightarrow 1$.

For the proof of the direction $1 \Rightarrow 2$, let $P$ be a p-optimal proof system for $L$ and let $A$ be a language from $C$. We choose a $C$-machine $N$ accepting $A$. Because $C$ is expressible in $L$, the set $\text{Correct}(N)$ is a polynomial-time decidable subset of $L$. Therefore we can devise a proof system $P_A$ with polynomial-size proofs of $\text{Correct}(N)$ as follows:

$$P_A(y) = \begin{cases} x & \text{if } y = 0x \text{ and } x \in \text{Correct}(N) \\ P(x) & \text{if } y = 1x \\ x_0 & \text{otherwise} \end{cases}$$

where $x_0$ is a fixed element from $L$. Apparently, $P_A$ is a proof system for $L$ such that $P_A$-proofs of $\text{Correct}(N)$ can be constructed in polynomial time, i.e., $A$ is p-representable in $P_A$. By the p-optimality of $P$ we infer $P_A \leq_P P$. Thus $P_A$-proofs of $\text{Correct}(N)$ can be efficiently translated into $P$-proofs of these formulas and therefore $A$ is also p-representable in $P$.

Implication $2 \Rightarrow 3$ holds because by Lemma 3.2, the $P$-easy subsets of $L$ are expressible in $L$.

For the proof of $3 \Rightarrow 1$ we show that item 3 yields a recursive $P$-presentation of all $P$-easy subsets of $L$. As both Theorem 4.1 and Theorem 6.5 use the assumption of expressibility of $PS(L)$ in $L$, we obtain a p-optimal proof system for $L$ by Theorem 4.1.

For this let $P$ be a proof system for $L$ in which all $P$-easy subsets of $L$ are p-representable. Let $(N_i)_{i \in \mathbb{N}}$ be an enumeration of all deterministic polynomial-time clocked Turing transducers and let $(N_i, q_i)_{i \in \mathbb{N}}$ be an enumeration of all deterministic polynomial-time clocked Turing machines with their respective running times. The machines $N_i$ will be candidates for machines accepting $P$-easy subsets of $L$ where the promise condition $R_{easy}$ for $N_i$ is $L(N_i) \subseteq L$. Finally, let $(\cdot, \cdot)$ be a polynomial-time computable and polynomial-time invertible bijective pairing function on $\mathbb{N}$.

Now, for $i \in \mathbb{N}$, consider the following set of algorithms $M_i$:

1. Input: $x$
2. compute numbers $j$ and $k$ with $i = (j, k)$
3. compute $N_j$, $q_j$, and $t_k$
4. IF $P(t_k(x)) \neq \text{corr}(x, N_j, 0^{f_j(|x|)})$ THEN reject
5. ELSE
6. IF $N_j(x)$ accepts in $\leq q_j(|x|)$ steps THEN accept
7. ELSE reject

Apparently, these algorithms $M_i$ can be computed in deterministic polynomial time. Further, each $M_i$ only accepts inputs from $L$ because if $M_i$ accepts $x$, then we have an $f$-proof for $\text{corr}(x, N_j, 0^{f_j(|x|)})$, confirming that $N_j$ obeys promise $R_{easy}$ on $x$, i.e., $x \in L$.

On the other hand, each $P$-easy subset $A \subseteq L$ is p-representable in $P$ with respect to some machine $N_j$ and some transducer $t_k$ computing the $P$-proofs of $\text{corr}(x, N_j, 0^{f_j(|x|)})$ from input $x$. For $i = (j, k)$ we then have $L(M_i) = A$. Thence $M_i$ is a recursive $P$-presentation of all $P$-easy subsets of $L$.

For the direction $2 \Rightarrow 4$, let $C$ be a promise class which is expressible in $L$. By item 2 we have a proof system $P$ for $L$ in which every language $A \in C$ is p-representable. We will construct a complete set $A_C$ for $C$ as follows. If $C$ is a language class, then $A_C$ will be a many-one complete language for $C$, and if $C$ is a function promise class, then $A_C$ will be a many-one complete function for $C$. Elements of $A_C$ are of the form $(x, N, 0^n, M, 0^n)$.

On such inputs the $C$-machine $M_A$ for $A_C$ performs the following operations. $M_A$ computes in polynomial time the string $\text{corr}(x, N, 0^n)$. It then simulates the Turing transducer $M$ on input $x$ for at most $n$ steps. If the simulation does not terminate, then $M_A$ rejects (in case of a language class $C$) or outputs a fixed element (for a function class $C$).

Otherwise, let $\pi$ be the output of the simulation $M(x)$. Next $M_A$ checks whether $P(\pi) = \text{corr}(x, N, 0^n)$. If this test fails, then again $M_A$ rejects the input (or outputs some fixed element). Otherwise, $M_A$ simulates the machine $N$ on input $x$ for $m$ steps and answers according to the answer obtained in this simulation.
By our general assumption, \( C \) has a universal machine and can perform polynomial-time computations. Therefore \( M_A \) is a \( C \)-machine and thus \( A_C \in C \). To verify the hardness of \( A_C \) for \( C \), let \( A \) be a language (or function) from \( C \). Because \( A \) is \( p \)-representable in \( P \), there exists a \( C \)-machine \( N \) accepting \( A \) such that \( P \)-proofs of \( \text{Correct}(N) \) can be efficiently constructed. Let \( p \) be the running time of \( N \) and let \( M \) be a Turing transducer with polynomial running time \( q \) that computes \( P \)-proofs of \( \text{Correct}(N) \) from input \( x \). Then the polynomial-time computable function

\[
x \mapsto (x, N, 0^p(|x|), M, 0^q(|x|))
\]

many-one reduces \( A \) to \( A_C \).

For the final implication \( 4 \Rightarrow 1 \) we need the assumption of the expressibility of \( \text{PS}(L) \) in \( L \). Because \( \text{PS}(L) \) is a promise function class, item 4 together with this assumption guarantees the existence of a many-one complete function for \( \text{PS}(L) \), which coincides with the notion of a \( p \)-optimal proof system for \( L \).

The next theorem contains a similar statement for optimal proof systems.

**Theorem 6.6** Let \( L \) be a language such that \( \text{PS}(L) \) is expressible in \( L \). Then the following conditions are equivalent:

1. There exists an optimal proof system for \( L \).
2. \( L \) has a proof system \( P \) such that every promise class which is expressible in \( L \) is representable in the system \( P \).
3. \( L \) has a proof system in which all \( P \)-easy subsets of \( L \) are representable.

**Proof.** The proof of the direction \( 1 \Rightarrow 2 \) proceeds similarly as the proof of \( 1 \Rightarrow 2 \) in Theorem 6.5. Let \( P \) be an optimal proof system for \( L \) and let \( A \) be a language from \( C \). We choose a \( C \)-machine \( N \) accepting \( A \). Because \( C \) is expressible in \( L \), the set \( \text{Correct}(N) \) is a polynomial-time subset of \( L \). We construct a proof system \( P_A \) with polynomial-size proofs of \( \text{Correct}(N) \) as follows:

\[
P_A(y) = \begin{cases} 
  x & \text{if } y = 0x \text{ and } x \in \text{Correct}(N) \\
  P(x) & \text{if } y = 1x \\
  x_0 & \text{otherwise}
\end{cases}
\]

where \( x_0 \) is a fixed element from \( L \). Apparently, \( P_A \) is a proof system for \( L \) with \( P_A \vdash_\ast \text{Correct}(N) \). By the optimality of \( P \) we infer \( P_A \leq P \). Thus also \( P \vdash_\ast \text{Correct}(N) \) holds, i.e., \( A \) is representable in \( P \).

As in the previous proof, implication \( 2 \Rightarrow 3 \) holds because the \( P \)-easy subsets of \( L \) are expressible in \( L \) by Lemma 3.2. The remaining implication \( 3 \Rightarrow 1 \) follows by an analogous argument as in the proof of \( 3 \Rightarrow 1 \) in Theorem 6.5.

Combining Theorems 6.1 and 6.6 we obtain the following corollary which is essentially contained in [KMT03].

**Corollary 6.7** Let \( L \) be a language. If \( L \) has an optimal proof system, then any promise language or function class \( C \) which is expressible in \( L \) and which can use nondeterminism has a complete language or function.

As the proof of the backward implication of Theorem 6.1 does not use provable closure of \( C \) under reductions in \( L \), we can formulate Corollary 6.7 without this assumption.

Comparing Theorem 6.5 and Corollary 6.7, it is apparent that while we could prove the equivalence of the existence of \( p \)-optimal proof systems for \( L \) and complete problems for all promise classes expressible in \( L \) (Theorem 6.5), we did not obtain this equivalence for optimal proof systems (cf. Corollary 6.7). The reason is that \( \text{PS}(L) \), considered as a promise function class, does not seem to have the property to use nondeterminism, because otherwise, the existence of an optimal proof system for \( L \) would already imply the existence of a \( p \)-optimal proof system for \( L \). We can even obtain a slightly stronger result:

**Proposition 6.8** If \( \text{PS}(\text{SAT}) \) can use nondeterminism, then every language with an optimal proof system also has a \( p \)-optimal proof system.
Recently, Cook and Krajíček [CK07] have introduced the concept of propositional proof systems with advice which seems to yield a strictly more powerful model than the classical Cook-Reckhow setting. We recall the definition of proof systems with advice from [BKM] which generalizes the original definition from [CK07].

Definition 7.1 ([BKM]) For a function \( k : \mathbb{N} \to \mathbb{N} \), a proof system \( f \) for \( L \) is a proof system with \( k \) bits of advice, if there exist a polynomial-time Turing transducer \( M \), an advice function \( h : \mathbb{N} \to \Sigma^* \), and an advice selector function \( \ell : \Sigma^* \to 1^* \) such that

1. \( \ell \) is computable in polynomial time,
2. \( M \) computes the proof system \( f \) with the help of the advice \( h \), i.e., for all \( \pi \in \Sigma^* \), \( f(\pi) = M(\pi, h([\ell(\pi)])) \), and
3. for all \( n \in \mathbb{N} \), the length of the advice \( h(n) \) is bounded by \( k(n) \).

Surprisingly, Cook and Krajíček [CK07] have shown that in the presence of advice, optimal propositional proof systems exist (cf. also [BKM, BM10] for a generalization to arbitrary languages and further results on proof systems with advice). Our next result shows that the relation between optimal proof systems and complete proof systems exist (cf. also [BKM, BM10] for a generalization to arbitrary languages and further results on optimal proof systems with advice).

Theorem 7.2 Let \( C \) be a promise complexity class and let \( L \) be a language such that \( C \) is expressible in \( L \) by a length-depending promise. Then \( C/1 \) contains a problem (or function) using one bit of advice which is many-one hard for \( C \).

Proof. We choose a tupling function \( \langle \cdot \rangle \) on \( \Sigma^* \) which is injective on lengths, i.e., for all sequences of strings \( x_1, \ldots, x_n, y_1, \ldots, y_n \in \Sigma^* \), \( \langle [x_1, \ldots, x_n] \rangle = \langle [y_1, \ldots, y_n] \rangle \) implies \( |x_i| = |y_i| \) for \( i = 1, \ldots, n \). Moreover, we choose \( \langle \cdot \rangle \) in such a way that for every fixed \( n \) the function \( \langle x_1, \ldots, x_n \rangle \) is polynomial-time computable and polynomial-time invertible.

We now define the problem (or function) \( A_C \) with one advice bit which will be many-one hard for \( C \). Inputs are of the form

\[
\langle x, 0^N, 0^m \rangle
\]

where \( x \) is the input, \( 0^N \) is the unary encoding of a Turing machine \( N \), and \( 0^m \) is the time bound for \( N \). On such an input, \( A_C \) first computes the string \( \text{corr}(x, N, 0^m) \). Then \( A_C \) uses its advice bit to verify whether or not \( \text{corr}(x, N, 0^m) \) is in \( L \) (for this step we could have also used the optimal proof system for \( L \) with one bit of advice, cf. [CK07, BKM]). If \( \text{corr}(x, N, 0^m) \in L \), then \( A_C \) simulates \( N \) on input \( x \) for at most \( m \) steps and produces the corresponding output (in case the simulation does not terminate it rejects or outputs some fixed element). As \( \langle \cdot, \cdot \rangle \) is length injective and \( \text{corr} \) is length depending, the element \( \text{corr}(x, N, 0^m) \) is uniquely determined by \( \langle x, 0^N, 0^m \rangle \) and therefore the advice bit of \( A_C \) can in fact refer to \( \text{corr}(x, N, 0^m) \).

If \( A \) is a problem (or function) from \( C \) and \( N \) is a \( C \)-machine for \( A \) with polynomial running time \( p \), then \( A \) many-one reduces to \( A_C \) via \( x \mapsto \langle x, 0^N, 0^{p(|x|)} \rangle \). Hence \( A_C \) is many-one hard for \( C \).
Corollary 7.3 There exist a disjoint pair \((A, B)\) and a sequence \((a_n)_{n \in \mathbb{N}}\) with the following properties:

1. \(A\) and \(B\) are computable in nondeterministic polynomial time with advice \(a_n\) for inputs of length \(n\).
2. The set \(\{(a_n, 0^n) \mid n \in \mathbb{N}\}\) is computable in \(\text{coNP}\).
3. Every disjoint \(\text{NP}\)-pair is strongly many-one reducible to \((A, B)\).

8 Hard Problems under a Tally \(\text{NP}\)-Oracle

We now show that for promise classes with a \(\text{coNP}\)-oracle to obtain similar results as in the previous section.

Recall that a set \(A \subseteq \Sigma^*\) is sparse if there exists a polynomial \(p\) such that for each \(n \in \mathbb{N}\), \(|A \cap \Sigma^n| \leq p(n)\). A sparse set \(A\) is called tally if \(A \subseteq \{1^n \mid n \in \mathbb{N}\}\). We denote the set of all sparse and tally sets by \(\text{Sparse}\) and \(\text{Tally}\), respectively. Sparse \(\text{NP}\)-sets appear to be very weak if used as oracles. For instance, \(\text{TAUT} \notin \text{NP}^S\) with a sparse \(\text{NP}\)-oracle \(S\), unless the polynomial hierarchy collapses to its second level \([\text{Kad89}]\).

Theorem 8.1 Let \(C\) be a promise language (or function) class which is representable in some language from \(\text{coNP}\). Then there exists a tally \(\text{NP}\)-oracle \(A\) such that \(C^A\) contains a language (or function) which is many-one hard for \(C\).

Proof. As in the proof of Theorem 7.2 we choose a tupling function \(\langle\ldots,\ldots\rangle\) which is injective on lengths and polynomial-time computable for sequences of fixed length.

The oracle set \(A\) now contains all machines which violate the promise condition on some given length, i.e.,

\[
A = \{ 1^m \mid m = |\langle 1^N, 1^n, 1^t \rangle| \text{ where } N, n, t \in \mathbb{N}, \text{ a nondeterministic Turing machine and there exists some } x \in \Sigma^n \text{ such that } \text{corr}(x, N, 1^t) \notin L \},
\]

and where \(L\) is the \(\text{coNP}\)-set in which the promise of \(C\) is expressible.

By definition, the set \(A\) is tally. Let us verify that \(A \in \text{NP}\). Because of the length injectivity of the tupling function, a number \(m \in \mathbb{N}\) already uniquely determines the tuple \(\langle 1^N, 1^n, 1^t \rangle\) with \(|\langle 1^N, 1^n, 1^t \rangle| = m\). Therefore, on input \(1^m\) we can first determine the entries \(N, n, t\) and then verify that \(N\) indeed encodes a nondeterministic Turing machine. Next we guess a string \(x \in \Sigma^n\) and compute \(\text{corr}(x, N, 1^t)\). As \(L \in \text{coNP}\), we can check \(\text{corr}(x, N, 1^t) \notin L\) in nondeterministic polynomial time.

The hard set for \(C\) will now contain elements \(\langle 1^N, x, 1^t \rangle\). On such input, we first query the string \(1^m\) with \(m = |\langle 1^N, 1^n, 1^t \rangle|\) to the oracle \(A\). If the answer is negative, then we simulate \(M\) on input \(x\) for at most \(t\) steps and answer according to the output of this simulation. If the answer is positive or if the simulation does not terminate in \(t\) steps, then we reject. It is easy to verify that this yields a hard set (or function) for \(C\). □

As disjoint \(\text{NP}\)-pairs have a \(\text{coNP}\)-promise (cf. [Bey07]), we obtain:

Corollary 8.2 There exists a strongly many-one hard disjoint \(\text{NP}\)-pair under a tally \(\text{NP}\)-oracle, i.e., there exists a tally set \(A\) in \(\text{NP}\) and a disjoint pair \((C_1, C_2)\) such that the following holds:

1. the components \(C_1\) and \(C_2\) are computable in \(\text{NP}^A\) with only one query to the oracle \(A\), and
2. every disjoint \(\text{NP}\)-pair strongly many-one reduces to \((C_1, C_2)\).

It is known that there is a close connection between disjoint \(\text{NP}\)-pairs and functions from \(\text{NPSV}\), single-valued functions computable in nondeterministic polynomial time (cf. [Sel94, BKM09, GSSZ04] for definitions and background information). Using this correspondence we can formulate Corollary 8.2 differently as:

Corollary 8.3 There exists a tally \(\text{NP}\)-set \(A\) and a function \(f \in \text{NPSV}^A\) such that every function from \(\text{NPSV}\) is many-one reducible to \(f\).

From Theorem 8.2 we also get a sufficient condition for the existence of complete disjoint \(\text{NP}\)-pairs:

Corollary 8.4 If \(\text{NP} = \text{NP}^{\text{NP}^{\text{Tally}}}\), then there exist \(\leq_s\)-complete disjoint \(\text{NP}\)-pairs.
We can rephrase this corollary using the notion of low sets from [Sch83]. Recall that a set $A \in \text{NP}$ is low for the $n$th level $\Sigma^n_p$ of the polynomial hierarchy if $(\Sigma^n_p)^A \subseteq \Sigma^n_p$. Intuitively, if a set $A$ is low for $\Sigma^n_p$, then $A$ is useless as an oracle for the class $\Sigma^n_p$. All sets $A \in \text{NP}$ which are low for $\Sigma^n_p$ are collected in the $n$th level $L_n$ of the low hierarchy. Using this terminology, we can express Corollary 8.4 differently as:

**Corollary 8.5** If $\text{NP} \cap \text{Tally} \subseteq L_1$, then there exist $\leq^*_s$-complete disjoint $\text{NP}$-pairs.

Whether or not $\text{NP} \cap \text{Tally} \subseteq L_1$ is open, but Ko and Schöning [KS85] have shown that $\text{NP} \cap \text{Sparse} \subseteq L_2$.

### 9 Conclusion

We provided new characterizations for our central questions Q1 on the existence of optimal proof systems and Q2 on the existence of complete sets for promise classes. Our results show that, in fact, Q1 is a special case of Q2, but also that proof systems are the hardest and most intriguing instance among all promise classes.

Whether or not there exist optimal proof systems or complete sets for promise classes remains unanswered by our results above. Hence, our central questions Q1 and Q2 remain open. These problems have been open for more than twenty years by now, and it seems that many researchers tend to believe in a negative answer. However, we want to point out that the status of Q1 (and thus also of Q2) is somewhat different from that of many other open problems in computational complexity. Namely, unlike for many open questions in complexity, a positive answer to Q1 (the answer contrary to “common” belief), is not known to imply any unlikely consequences (such as collapse consequences). In fact, the only known consequences of a positive answer to Q1 seem be positive answers to Q2. Regarding the latter, we mention that Itsykson [Its09] has recently exhibited the first promise class where Q2 indeed receives a positive answer.

In summary, the question of the existence of optimal proof systems is an important problem which even becomes more interesting by the fact that it does not seem quite clear what the expected answer should be. One promising approach for future research is further investigation into those models in which we have positive answers: under advice (Theorem 7.2), under oracles (Theorem 8.1), and for randomization (cf. [HI10, Hir10]).

Trying to weaken the models in which we get completeness results on the one hand, and exploring the limitations of these techniques on the other hand, might yield further insights into Q1 and Q2.

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### References


