Quantifying uncertainties in primordial nucleosynthesis without Monte Carlo simulations

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Abstract

We present a simple method for determining the (correlated) uncertainties of the light element abundances expected from big bang nucleosynthesis, which avoids the need for lengthy Monte Carlo simulations. Our approach helps to clarify the role of the different nuclear reactions contributing to a particular elemental abundance and makes it easy to implement energy-independent changes in the measured reaction rates. As an application, we demonstrate how this method simplifies the statistical estimation of the nucleon-to-photon ratio through comparison of the standard BBN predictions with the observationally inferred abundances.

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I. INTRODUCTION

Big bang nucleosynthesis is entering the precision era [1]. On the one hand, there has been major progress in the observational determination of the abundances of the light elements D [2], 3\(^{\text{He}}\) [3], 4\(^{\text{He}}\) [4], and 7\(^{\text{Li}}\) [5], although the increasing precision has highlighted discrepancies between different measurements (see Refs. [10–12] for recent assessments). Secondly, we have a sound analytical understanding of the physical processes involved [13,14] and the standard BBN computer code [15,16] which incorporates this physics is robust and can be easily altered to accommodate changes in the input parameters, e.g. nuclear reaction rates [17]. The comparison of increasingly accurate observationally inferred and theoretical abundances will further constrain the values of fundamental parameters, such as the nucleon density parameter (see, e.g., Ref. [18]) or extra degrees of freedom related to possible new physics beyond the Standard Model (see, e.g., Ref. [19]). It goes without saying that error evaluation represents an essential part of such comparisons.

Because of the complex interplay between different nuclear reactions, it is not straightforward to assess the effect on a particular elemental yield of the uncertainties in the experimentally determined reaction rates. The authors of Ref. [20] first employed Monte Carlo methods to sample the error distributions of the relevant reaction cross-sections which were then used as inputs to the standard BBN computer code. This enables well-defined confidence levels to be attached to the theoretically predicted abundances, e.g. the abundance range within which say 95\% of the computed values fall correspond to 95\% C.L. limits on the expected abundance. It was later realized that error correlations are also relevant, and can be estimated with the same technique [21,22]. The Monte Carlo (MC) approach has since become the standard tool for comparing theory and data [17,23–26]. However, although it can include refinements such as asymmetric or temperature-dependent reaction rate uncertainties [17], it requires lengthy calculations which need to be repeated each time (any of) the input parameters are changed or updated. Since we may expect continued improvement in the determination of the relevant parameters, it is desirable to have a faster method for error evaluation and comparison with observations.

In this work we propose a simple method for estimation of the BBN abundance uncertainties and their correlations which requires little computational effort. The method, based on linear error propagation, is described in Sec. II. A concrete application is given in Sec. III, where theory and observations are compared using simple \(\chi^2\) statistics to obtain the best-fit value of the nucleon-to-photon ratio. In Sec. IV we study with this method the relative importance of different nuclear reactions in determining the synthesized abundances. Conclusions and perspectives for further work are presented in Sec. V.

II. PROPAGATING INPUT CROSS SECTION UNCERTAINTIES TO OUTPUT ELEMENTAL ABUNDANCES

A. Notation and input

The four relevant element abundances \(Y_i\) considered in this work are defined in Table I. (Note that the abundance of 4\(^{\text{He}}\) is conventionally quoted as a mass fraction, while the
abundances of D, $^3$He and $^7$Li are ratios by number.) In BBN calculations, the $Y_i$'s depend both on model parameters (the nucleon-to-photon ratio $\eta$, the number of neutrino families $N_\nu$, etc.) and on a network of nuclear reactions $R_k$:

$$Y_i = Y_i(\eta, N_\nu, \ldots; \{R_k\}) .$$

(1)

The most important $R_k$'s are listed, numbered as in Ref. [16], in the first two columns of Table II, while our default inputs for the rates $R_k$ and their $1\sigma$ uncertainties $\pm \Delta R_k$ are given in the third and fourth columns. The numerical values are given in ratio to the reference reaction rates compiled in Table 1 of Ref. [17]; we have chosen identical values (i.e., $R_k = 1$) except for $R_1$, the neutron decay rate, where we adopt the most recent world average for the neutron lifetime of $\tau_n = 886.7 \pm 1.9$ s [27,28], as compared to the value of $\tau_n = 888.54 \pm 3.73$ s used in Ref. [17]. The fractional uncertainties $\pm \Delta R_k/R_k (k \neq 1)$ have also been taken from Ref. [17] (see their Table 2), assuming conservatively the largest value for the temperature-dependent errors $\Delta R_7$ and $\Delta R_{10}$.

**B. The method**

For simplicity we consider only the standard BBN case (i.e., $N_\nu = 3$, etc.), so that $\eta$ is the only model parameter being varied in the calculation of the abundances $Y_i$ and of their uncertainties $\sigma_i$:

$$Y_i = Y_i(\eta) \pm \sigma_i(\eta) .$$

(2)

Our method can however be easily generalized to nonstandard cases.

For a relatively small change $\delta R_k$ of the input rate $R_k$ ($R_k \rightarrow R_k + \delta R_k$), the corresponding deviation $\delta Y_i$ of the $i$-th elemental abundance ($Y_i \rightarrow Y_i + \delta Y_i$), as given by linear propagation, reads

$$\delta Y_i(\eta) = Y_i(\eta) \sum_k \lambda_{ik}(\eta) \frac{\delta R_k}{R_k} ,$$

(3)

where the functions $\lambda_{ik}(\eta)$ represent the logarithmic derivatives of $Y_i$ with respect to $R_k$:

$$\lambda_{ik}(\eta) = \frac{\partial \ln Y_i(\eta)}{\partial \ln R_k(\eta)} .$$

(4)

In general, the deviations $\delta Y_i$ in Eq. (3) are correlated, since they all originate from the same set of reaction rate shifts $\{\delta R_k\}$. The global information is contained in the error matrix (also called covariance matrix) [29], which is a generalization of the “error vector” $\delta Y_i$ in Eq. (3). In particular, the abundance error matrix $\sigma^2_{ij}(\eta)$ obtained by linearly propagating the input $\pm 1\sigma$ reaction rate uncertainties $\pm \Delta R_k$ to the output abundances $Y_i$ reads:

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1Our method for error propagation requires that the $\Delta R_k/R_k$'s be constant (i.e., temperature independent). We will comment on this point at the end of this Section.
\[ \sigma_{ij}^2(\eta) = Y_i(\eta)Y_j(\eta) \sum_k \lambda_{ik}(\eta)\lambda_{jk}(\eta) \left( \frac{\Delta R_k}{R_k} \right)^2. \] (5)

This matrix completely defines the abundance uncertainties. In particular, the 1σ abundance errors \( \sigma_i \) of Eq. (2) are given by the square roots of the diagonal elements,

\[ \sigma_i(\eta) = \sqrt{\sigma_{ii}^2(\eta)}, \] (6)

while the error correlations \( \rho_{ij} \) can be derived from Eqs. (5,6) through the standard definition

\[ \rho_{ij}(\eta) = \frac{\sigma_{ij}^2(\eta)}{\sigma_i(\eta)\sigma_j(\eta)}. \] (7)

Thus Eqs.(2–7) represent all that is required to calculate the errors in the predicted abundances and their correlations.

Note that the relevant physics is contained entirely in the central values \( Y_i \) and in their logarithmic derivatives \( \lambda_{ik} \), which have to be evaluated just once with a BBN numerical code, thus dramatically reducing the required computing time.\(^2\) We have made a further check of the linearity of the error propagation by calculating the logarithmic derivatives with increments equal to \( \Delta R_k \) (default) and \( 2 \Delta R_k \), obtaining practically the same functions \( \lambda_{ik} \) in either case. This means that doubling the error on \( R_k \) also doubles the corresponding error component of \( Y_i \), i.e. the error propagation is indeed linear.

We think it useful to present the results of this exercise in the form of tables so all calculations that will now follow can be done on a pocket calculator. Tables III and IV show the coefficients of polynomial fits to \( Y_i \) and \( \lambda_{ik} \), respectively, for \( \eta \) in the usually considered range \( 10^{-10} - 10^{-9} \).\(^3\) The abundances \( Y_i(\eta) \) with their associated \( \pm 2 \) standard deviation error bands calculated through Eq. (3) are shown in Fig. 1. The functions \( \lambda_{ik}(\eta) \) are shown in Fig. 2; note that some of these vary strongly (and even change sign) with \( \eta \), indicating that the physical dependence of the \( Y_i \)'s on the \( R_k \)'s may be quite subtle. Although some general features of this dependence have been addressed in Ref. [14], further work is needed to interpret the functional form of the \( \lambda_{ik} \)'s in Fig. 2. We intend to address this issue elsewhere. Finally, Fig. 3 displays the fractional uncertainties \( \sigma_i/Y_i \) and their correlations \( \rho_{ij} \) as derived from Eqs. (5–7). Notice that, in general, the error correlations are non-negligible and should be properly taken into account in statistical analyses, as first emphasized in Ref. [21].

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\(^2\)The logarithmic derivatives are numerically defined as \( \lambda_{ik}(\eta) = \frac{[Y_i(\eta, R_k + \Delta R_k) - Y_i(\eta, R_k - \Delta R_k)]R_k/2Y_i(\eta, R_k)\Delta R_k}{Y_i(\eta, R_k)} \) at given \( \eta \). We find negligible difference between left and right derivatives.

\(^3\)Small corrections to the helium abundance \( Y_4 \) due to Coulomb, radiative and finite temperature effects, finite nucleon mass effects and differential neutrino heating, have been incorporated according to the prescription given in Ref. [19]. We have used the BBN code [14] with the lowest possible settings of the time steps in the (2nd order) Runge-Kutta routine, which allows rapid convergence to within 0.01% of the true value [30]. We understand that our results are in good agreement with a recent independent computation of \( Y_4 \) using a new BBN computer code [31].
In summary, the recipe for evaluating the BBN uncertainties ±\sigma_i affecting the Y_i’s for a given value of \eta is:

(i) Determine the abundances Y_i(\eta) and their logarithmic derivatives \lambda_{ik}(\eta) using Tables III and IV, respectively;

(ii) If the central values of the reaction rates R_k are updated (R_k \rightarrow R_k + \delta R_k) with respect to those reported in Table II, then update also the central values of the abundances (Y_i \rightarrow Y_i + \delta Y_i) through Eq. (3);

(iii) For given reaction rate uncertainties \Delta R_k (e.g., from Table III), compute the abundance errors \sigma_i and their correlations \rho_{ij} using Eqs. (5–7).

C. Comparison with MC estimates and remarks

Our approach is based on the linear propagation of errors originating from many independent sources (i.e. the R_k’s). One can expect that this method will work reasonably well, both because the input fractional uncertainties \Delta R_k/R_k are relatively small, and because the final output uncertainties \sigma_i affecting the abundances Y_i are “regularized” by the central limit theorem. Indeed, our ±2\sigma bands in Fig. 1 compare well with the MC-estimated bands of Refs. [17,21,23–26], with small relative differences which depend, in part on different input \( R_k \pm \Delta R_k \)'s, and that are not larger than the spread among the various MC estimates themselves.

In order to be more quantitative, we compare in Fig. 4 (upper panel) the MC evaluation of the fractional uncertainties \sigma_i/Y_i as derived from Ref. [17] with our analytic estimate (using, for this exercise, the same input parameters). There is good agreement between these two totally independent estimates. In Fig. 4 (lower panel) we also show a comparison with the only MC evaluation of \rho_{ij}, we are aware of (viz., Ref. [22]), obtaining again good agreement with our calculation when the same input \( R_k \pm \Delta R_k \) are used. We conclude the discussion of Fig. 4 by noting that the uncertainty \sigma_{(2+3)} and the correlations \rho_{(2+3)j} related to the often-used combination of abundances \( Y_{(2+3)} = Y_2 + Y_3 = (D + ^3\text{He})/\text{H} \) are given, within our approach, by:

\[
\sigma^2_{(2+3)} = \sigma_2^2 + \sigma_3^2 + 2\rho_{23}\sigma_2\sigma_3 ,
\]

\[
\rho_{(2+3)j}\sigma_{(2+3)}\sigma_j = \rho_{2j}\sigma_2\sigma_j + \rho_{3j}\sigma_3\sigma_j .
\]

There are, of course, some refined features of the MC approach that cannot be addressed with our method, such as asymmetric or temperature-dependent uncertainties \Delta R_k [17]. However we consider these refinements not essential for practical applications. In a sense, the possible asymmetry between “upper” and “lower” errors is where one wants it to be. For

\(^4\)The MC values of \sigma_i/Y_i have been read off the (small) panels of Fig. 27 in Ref. [17] and, therefore, may be subject to small transcription errors.
instance, if one assumes \textit{a priori} symmetric errors in the astrophysical $S$-factors, then asymmetric errors are induced in the thermally averaged reaction rates $R \sim \langle \sigma v \rangle$; conversely, the requirement of \textit{a priori} symmetric $\Delta R_k$ errors requires that the input $S$-factor uncertainties are readjusted, as discussed in Ref. \cite{17}. Although the authors of Ref. \cite{17} have adopted the latter option ($| + \Delta R_k | = | - \Delta R_k |$), the former option or others are equally acceptable, and would clearly produce different outputs for the MC estimate of the abundance errors. For instance, the upper and lower errors of $Y_7$ appear to be rather symmetrical in the MC calculation of Ref. \cite{17}, while they are noticeably asymmetrical in Ref. \cite{22}.

Concerning the temperature-dependent \cite{17} uncertainties $\Delta R_7$ and $\Delta R_{10}$, which affect mainly the estimate of $Y_7$, our conservative choice in Table II proves to be successful for the estimate of $\sigma_7 / Y_7$ (see Fig. 4). This seems to indicate that the uncertainties at low temperatures (which are larger than those at high temperatures \cite{17}) dominate in the estimate of these errors. In any case, since practically all reaction rate uncertainties $\Delta R_k$ contribute to the final value of $\sigma_7$, temperature-dependent refinements in the propagation of just two of these ($\Delta R_7$ and $\Delta R_{10}$) do not appear to be decisive for the estimate of the global error $\sigma_7$.

In conclusion, we have shown that our simple analytic method for error evaluation represents an useful alternative to lengthy and computationally expensive MC simulations. Both the magnitude and the correlations of the total errors affecting the theoretical abundances are reproduced with good accuracy. We therefore advocate the use of this method for BBN analyses as an alternative to MC simulations.\footnote{5}

III. DETERMINING THE LIKELY NUCLEON-TO-PHOTON RATIO

The comparison of the predicted primordial abundances $Y_i(\eta) \pm \sigma_i(\eta)$ with their observationally inferred values $Y_i \pm \sigma_i$ through a statistical test allows extraction of the likelihood range for the fundamental parameter $\eta$. So far, this has been done either through fit-by-eye (see, e.g., Ref. \cite{17}) or by Monte Carlo-based maximum likelihood methods (see, e.g., Refs. \cite{25,26}). In this Section we show how limits on $\eta$ can be simply extracted using $\chi^2$ statistics based on the method described in the previous Section.

Assuming that the errors $\sigma_i$ in the determinations of different abundances $Y_i$ are uncorrelated, the experimental squared error matrix $\sigma^2_{ij}$ is simply

\[ \sigma^2_{ij} = \delta_{ij} \sigma_i \sigma_j, \tag{10} \]

where $\delta_{ij}$ is Kronecker’s delta. The total (experimental + theoretical) error matrix $S^2_{ij}$ is then obtained by summing the matrices in Eqs. (3,14):

\[ S^2_{ij}(\eta) = \sigma^2_{ij}(\eta) + \sigma^2_{ij}. \tag{11} \]

\footnote{5}{A parallel situation holds in the field of solar neutrino physics, where the correlated uncertainties of the neutrino fluxes predicted by solar models have been estimated through both Monte Carlo simulations \cite{32} and linear propagation of input errors \cite{33}. The latter technique has proved more popular because of its ease of use.}
Its inverse defines the weight matrix $W_{ij}(\eta)$:

$$W_{ij}(\eta) = [S^2_{ij}(\eta)]^{-1}.$$  \hspace{1cm} (12)

The $\chi^2$ statistic associated with the difference between theoretical ($Y_i$) and observational ($Y_i$) light element abundance determinations is then [29]:

$$\chi^2(\eta) = \sum_{ij} [Y_i(\eta) - Y_i] \cdot W_{ij}(\eta) \cdot [Y_j(\eta) - Y_j].$$  \hspace{1cm} (13)

Minimization of the $\chi^2$ gives the most probable value of $\eta$, while the intervals defined by $\chi^2 = \chi^2_{\text{min}} + \Delta \chi^2$ give the likely ranges of $\eta$ at the confidence level set by $\Delta \chi^2$ (for one degree of freedom, $\eta$).

In order to illustrate this, we estimate $\eta$ using recent observational data for the three light element abundances ($Y_2, Y_4, Y_7$) whose primordial origin is most secure. It is well known that the observationally inferred values $Y_2$ and $Y_4$ are still controversial, and the conflict between different determinations has driven a lively debate on the status of BBN (see Refs. [18,34] and references therein). In this paper we do not enter into this debate but rather apply our method to two possible (although mutually incompatible) selections of measurements which we name data set “A” and data set “B”:

Data set A: \begin{align*}
Y_2 &= 1.9 \pm 0.4 \times 10^{-4}, \\
Y_4 &= 0.234 \pm 0.0054, \\
Y_7 &= 1.6 \pm 0.36 \times 10^{-10}.
\end{align*}

Data set B: \begin{align*}
Y_2 &= 3.40 \pm 0.25 \times 10^{-5}, \\
Y_4 &= 0.243 \pm 0.003, \\
Y_7 &= 1.73 \pm 0.12 \times 10^{-10}.
\end{align*}

The data set “A” is used in Ref. [26], the authors of which make a detailed MC-based fit to $\eta$, thus enabling comparison with our method. Note that their adopted value of the primordial deuterium abundance from observations of high redshift quasar absorption systems is consistent with another recent observation, $Y_2 = (2.15 \pm 0.35) \times 10^{-4}$ [3], but in conflict with the significantly smaller value reported in Ref. [2], which we adopt for data set “B”. Similarly the primordial helium mass fraction inferred from observations of metal-poor blue compact galaxies which we adopt for data set “B” is from Ref. [7] in which it is argued that previous analyses leading to the smaller value of $Y_2$ used in data set “A” underestimate the true abundance (although this is disputed in Ref. [6].) Finally the estimates for the primordial lithium abundance in both data sets are based on observations of pop II stars, with the slightly higher value [9] of $Y_7$ in data set “B” taken from an updated analysis. Readers who prefer to adopt different combinations of these, or indeed other, estimates for

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6We remind the reader that $\Delta \chi^2 = 1$, 2.71, 3.84, and 6.64 correspond to confidence levels (C.L.’s) of 68%, 90%, 95%, and 99%, respectively.
the primordial abundances are invited to perform their own fit to \( \eta \) by following the simple prescription given here.

Before performing the \( \chi^2 \) fit, it is useful to get an idea of what one should expect by comparing the data with the theoretical predictions at various values of \( \eta \). Fig. 5 shows the theoretical predictions for the abundances \( Y_2, Y_4 \), and \( Y_7 \) in the three possible planes \((Y_i, Y_j)\), for representative values of \( x \equiv \log_{10}(\eta/10^{-10}) \). The corresponding 1\( \sigma \) error ellipses show clearly the size and the correlation of the “theoretical” errors. The observational data sets “A” and “B” are also indicated on the figure, as crosses with 1\( \sigma \) error bars. Clearly the former prefers \( \eta \sim 2 \times 10^{-10} \) while the latter favours \( \eta \sim (4–5) \times 10^{-10} \).

A more precise estimate of the likely range of \( \eta \) is obtained, as anticipated, through a \( \chi^2 \) fit. The results are shown in Fig. 6. The value of \( \chi^2_{\text{min}} \) is almost zero for the fit to data set “A”, indicating very good agreement between theory and observations, while it is somewhat larger for data set “B”. Note that the characteristic minimum in the \( ^7\text{Li} \) curve at \( \eta \approx 2.6 \times 10^{-10} \) (see bottom panel of Fig. 4) allows its measured abundance to be compatible with both high D/low \( ^4\text{He} \) (data set “A”) or low D/high \( ^4\text{He} \) (data set “B”). The 95\% C.L. ranges allowed by each of the two data sets, as obtained by cutting the curves at \( \Delta \chi^2 = \chi^2 - \chi^2_{\text{min}} = 3.84 \), are:

\[
\begin{align*}
\text{Data set A} : \quad \eta &= 1.78^{+0.54}_{-0.34} \times 10^{-10} , \\
\text{Data set B} : \quad \eta &= 5.13^{+0.72}_{-0.66} \times 10^{-10} .
\end{align*}
\]

The range for case “A” agrees very well with the 95\% C.L. range estimated in Ref. \[26\] with the same inputs but with a different method (Monte Carlo + maximum likelihood). Of course, the incompatibility between the above two ranges of \( \eta \) reflects the incompatibility between the input abundance data within their stated errors.

\section*{IV. ROLE OF DIFFERENT REACTIONS IN LIGHT ELEMENT NUCLEOSYNTHESIS}

The role of the different nuclear reactions rates listed in Table II in the synthesis of the light elements can be studied by “perturbing” the values of the input reaction rates and observing their effect on the predicted abundances. More precisely, one can study the contribution to the total uncertainty \( \sigma_i \) of \( Y_i \) induced by a +1\( \sigma \) shift of \( R_k \):

\[
R_k \rightarrow R_k + \Delta R_k \quad \Rightarrow \quad Y_i \rightarrow Y_i + \delta Y_i .
\]

Within our approach, this can be done very easily using Eq. (3), with the \( \Delta R_k \)’s from Table II. Of course, the results depend on the value chosen for \( \eta \). To illustrate various trends, we choose the best-fit values \( \eta = 1.78 \times 10^{-10} \) and \( \eta = 5.13 \times 10^{-10} \), corresponding to data sets “A” and “B” respectively.

Figs. 7 and 8 show the deviations \( \delta Y_i \) (normalized to the total error \( \sigma_i \)) induced by +1\( \sigma \) shifts in the \( R_k \)’s, plotted in the same set of planes as used for Fig. 4. The 1\( \sigma \) error ellipses shown in these figures are obtained by combining the deviation vectors \( \delta Y_i/\sigma_i \) in an uncorrelated manner. Several interesting conclusions can be drawn from this exercise. As expected, the uncertainty in the weak interaction rate \( R_1 \) has the greatest impact on
$Y_4$ for the high value of $\eta$ (Fig. 3), since essentially all neutrons end up being bound in $^4$He. However at the lower value of $\eta$ (Fig. 7), the uncertainty in $R_2$ — the “deuterium bottleneck” — plays an equally important role as $R_1$ in determining $Y_4$ because nuclear burning is less complete here than at high $\eta$. Similarly with reference to the reaction rates $R_7$, $R_{10} - R_{12}$ which synthesize $^7$Li, at low $\eta$ it is the competition between $R_7$ and $R_{11}$ which largely determines $Y_7$, while at high $\eta$ it is the competition between $R_{10}$ and $R_{12}$. The anticorrelation between $Y_4$ and $Y_2$ is driven mainly by $R_2$ at low $\eta$ and, to a lesser extent, by $R_4$ and $R_5$, while the reverse is the case at high $\eta$. The anticorrelation between $Y_4$ and $Y_7$ at low $\eta$ is also basically driven by $R_2$, while the correlation at high $\eta$ is due to both $R_2$ and $R_4$. Thus we have a direct visual basis for assessing in what direction the output abundances $Y_i$ are pulled by possible changes in the input cross sections $R_k$.

V. CONCLUSIONS

We have shown that a simple method based on linear error propagation allows us to quantify the uncertainties associated with the elemental abundances expected from big bang nucleosynthesis, in excellent agreement with the results obtained from Monte Carlo simulations. This method makes transparent which nuclear reaction rate is mainly responsible for the uncertainty in the abundance of a given element. If determinations of the primordial abundances improve to the point where the observational errors become smaller than the theoretical uncertainties (say for $^7$Li), this will enable attention to be focussed on the particular reaction rate whose value needs to be experimentally better known.

We have also demonstrated that for standard BBN, our method enables the use of simple $\chi^2$ statistics to obtain the best-fit value of $\eta$ from the comparison of theory and observations. At present there are conflicting claims regarding the primordial abundances of, particularly, D and $^4$He, and different choices of input data sets imply values of $\eta$ differing by a factor of $\sim 3$. However this quantity can also be determined through measurements of the angular anisotropy of the cosmic microwave background (CMB) on small angular scales. Within a decade the forthcoming all-sky surveyors MAP and PLANCK are expected to pinpoint the nucleon density to within $\sim 5\%$ [35]. Such measurements probe the acoustic oscillations of the coupled photon-matter plasma at the (re)combination epoch and will thus provide an independent check of BBN, assuming $\eta$ did not change significantly between the two epochs. Nevertheless precise measurements of light element abundances, particularly $^4$He, are still crucial because they provide a unique probe of physical conditions, in particular the expansion rate at the BBN epoch. To illustrate, if $\eta$ was determined by the CMB measurements to be $\approx 2 \times 10^{-10}$ (consistent with data set “A”), but the abundance of $^4$He was established to be actually closer to its higher value of $\approx 24\%$ in data set “B”, this

\footnote{New physics beyond the Standard Model can change $\eta$, e.g. by increasing the photon number through massive particle decay [36] or, more exotically, by decreasing the photon number through photon mixing with a shadow sector [37]. However such possibilities are strongly constrained by the absence of distortions in the Planck spectrum of the CMB [38] and also, in the latter case, by the absence of Sakharov oscillations in the power spectrum of large-scale structure [39].}
would be a strong indication that the expansion rate during BBN was higher than in the standard case with $N_\nu = 3$ neutrinos. Although the number of $SU(2)$ doublet neutrinos is indeed 3, there are many light particles expected in extensions of the Standard Model, e.g. singlet neutrinos, which can speed up the expansion rate during nucleosynthesis [19]. The generalization of our method to such non-standard cases is straightforward and we intend to present these results in a future publication [20]. It is clear that BBN analyses will continue to be important in this regard for both particle physics and cosmology.

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### TABLE I. The four light elemental abundances $Y_i$ considered in this work. Alternative symbols used in the literature are indicated in parentheses.

| Symbol | ... or | Definition            |
|--------|--------|-----------------------|
| $Y_4$  | $(Y'_p)$ | $^4\text{He}$ mass fraction |
| $Y_2$  | $(y_{2p})$ | $^3\text{He}/^2\text{H}$ (by number) |
| $Y_3$  | $(y_{3p})$ | $^3\text{He}/^4\text{He}$ (by number) |
| $Y_7$  | $(y_{7p})$ | $^7\text{Li}/^4\text{He}$ (by number) |

### TABLE II. The BBN reaction rates $R_k$ and their $1\sigma$ uncertainties $\pm \Delta R_k$ adopted in this work. The numbering follows Ref. [16] while the reference “unit” values ($R_k \equiv 1$) correspond to the rates in Ref. [17].

| $k$ | Reaction | $R_k$ | $\pm \Delta R_k$ |
|-----|----------|-------|-----------------|
| 1   | $n \rightarrow p e \bar{\nu}_e$ | 0.9979 | $\pm 0.0021$ |
| 2   | $p(n,\gamma)d$ | 1 | $\pm 0.07$ |
| 3   | $d(p,\gamma)^3\text{He}$ | 1 | $\pm 0.10$ |
| 4   | $d(d,n)^3\text{He}$ | 1 | $\pm 0.10$ |
| 5   | $d(d,p)t$ | 1 | $\pm 0.10$ |
| 6   | $t(d,n)^4\text{He}$ | 1 | $\pm 0.08$ |
| 7   | $t(\alpha,\gamma)^7\text{Li}$ | 1 | $\pm 0.26$ |
| 8   | $^3\text{He}(n,p)t$ | 1 | $\pm 0.10$ |
| 9   | $^3\text{He}(d,p)^4\text{He}$ | 1 | $\pm 0.08$ |
| 10  | $^3\text{He}(\alpha,\gamma)^7\text{Be}$ | 1 | $\pm 0.16$ |
| 11  | $^7\text{Li}(p,\alpha)^4\text{He}$ | 1 | $\pm 0.08$ |
| 12  | $^7\text{Be}(n,p)^7\text{Li}$ | 1 | $\pm 0.09$ |

### TABLE III. Polynomial fit to the central value of the elemental abundances, $Y_i = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + a_4 x^4 + a_5 x^5$, with $x \equiv \log_{10}(\eta/10^{-10})$ in the range 0–1. The abundances were obtained using the BBN computer code [16] with the input $R_k$’s as in Table II. The value of $Y_4$ has been corrected using the prescription of Ref. [19]. The accuracy of the fit is better than 1/25 of the total theoretical uncertainty for each $Y_i$.

| $Y_i \times 10^3$ | $a_0$ | $a_1$ | $a_2$ | $a_3$ | $a_4$ | $a_5$ |
|-------------------|-------|-------|-------|-------|-------|-------|
| $Y_2$             | +0.4808| –1.8112| +3.2564| –3.3525| +1.8834| –0.4458|
| $Y_3$             | +3.4308| –6.1701| +8.1311| –9.7612| +7.7018| –2.5244|
| $Y_4$             | +2.2305| +0.5479| –0.6050| +0.6261| –0.3713| +0.0949|
| $Y_7$             | +0.5369| –2.8036| +7.6983| –12.571| +12.085| –3.8632|
TABLE IV. Polynomial fits to the logarithmic derivatives, $\lambda_{nk}(x) = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + a_4 x^4 + a_5 x^5$, as functions of $x \equiv \log_{10}(\eta/10^{-10}) \in [0,1]$ (see Fig. 2). In most cases, polynomials of degree < 5 provide sufficiently accurate fits. Only non-negligible logarithmic derivatives are tabulated.

| $k$ | $a_0$ | $a_1$ | $a_2$ | $a_3$ | $a_4$ | $a_5$ |
|-----|-------|-------|-------|-------|-------|-------|
| $\lambda_{2k}$ 1 | +0.7130 | −0.7964 | +0.4577 | +0.0914 | 0 | 0 |
| 2 | −0.7025 | +0.2611 | +1.2008 | −0.8934 | 0 | 0 |
| 3 | −0.0189 | −0.1879 | +0.2502 | −0.6806 | 0 | 0 |
| 4 | −0.4228 | −0.1698 | +0.0207 | +0.0247 | 0 | 0 |
| 5 | −0.4138 | −0.1477 | +0.1103 | +0.0010 | 0 | 0 |
| 8 | −0.0073 | −0.0003 | +0.0801 | −0.0416 | 0 | 0 |
| 9 | −0.0011 | −0.0348 | +0.0592 | −0.0511 | 0 | 0 |
| $\lambda_{3k}$ 1 | +0.0940 | +0.1892 | −0.1484 | −0.0199 | 0 | 0 |
| 2 | +0.0981 | +0.9948 | −3.1667 | +3.4108 | −1.2845 | 0 |
| 3 | +0.0610 | +0.1640 | +0.5368 | −0.2605 | 0 | 0 |
| 4 | +0.3050 | +0.0805 | −0.4208 | +0.1555 | 0 | 0 |
| 5 | −0.5118 | +0.1274 | +0.4081 | −0.2106 | 0 | 0 |
| 6 | −0.0327 | +0.0829 | −0.0939 | +0.0362 | 0 | 0 |
| 8 | −0.5580 | −0.0287 | +1.3574 | −0.8735 | 0 | 0 |
| 9 | −0.1080 | −0.5089 | −1.0157 | +0.8163 | 0 | 0 |
| $\lambda_{4k}$ 1 | +0.8138 | −0.1465 | +0.0408 | 0 | 0 | 0 |
| 2 | +0.0610 | −0.1962 | +0.2416 | −0.1049 | 0 | 0 |
| 4 | +0.0082 | −0.0058 | +0.0034 | 0 | 0 | 0 |
| 5 | +0.0075 | −0.0058 | +0.0034 | 0 | 0 | 0 |
| $\lambda_{7k}$ 1 | +1.9638 | +1.8520 | −19.721 | +27.542 | −11.041 | 0 |
| 2 | −0.9214 | −1.6472 | +5.6187 | +54.059 | −112.80 | +56.426 |
| 3 | +0.0500 | −1.0433 | +4.1384 | −2.4441 | 0 | 0 |
| 4 | +0.1734 | −2.7428 | +11.209 | −10.736 | +2.5263 | 0 |
| 5 | +0.1837 | −0.0875 | +0.0158 | −0.1350 | 0 | 0 |
| 6 | −0.9877 | −0.8168 | +6.1555 | −4.4380 | 0 | 0 |
| 7 | +0.9644 | +0.6888 | −5.6151 | +4.0284 | 0 | 0 |
| 8 | +0.0529 | +0.6318 | −3.3995 | +2.5754 | 0 | 0 |
| 9 | −0.0764 | +0.3017 | −2.9632 | +1.9311 | 0 | 0 |
| 10 | +0.0690 | −1.5360 | +9.5808 | −10.168 | +2.9807 | 0 |
| 11 | −1.4095 | −0.3543 | +6.4780 | −4.7956 | 0 | 0 |
| 12 | −0.0043 | −0.3779 | +7.7358 | −42.390 | +61.401 | −26.778 |
REFERENCES

[1] D. N. Schramm and M. S. Turner, Rev. Mod. Phys. (Colloquia) 70, 303 (1998).
[2] S. Burles and D. Tytler, astro-ph/9712263, in Primordial Nuclei and their Galactic Evolution, Proc. ISSI Workshop (Bern, 1997) edited by N. Prantzos, M. Tosi, and R. von Steiger (Kluwer, Dordrecht), to appear.
[3] J. K. Webb, R. F. Carswell, K. M. Lanzetta, R. Ferlet, M. Lemoine, and A. Vidal-Madjar, astro-ph/9710089, in Structure and Evolution of the Intergalactic Medium from QSO Absorption Line Systems, Proceedings of the 13th IAP Workshop (Paris, 1997), edited by P. Petitjean and S. Charlot (Nouvelles Frontières, Paris), to appear.
[4] D. S. Balser, T. M. Bania, R. T. Rood and T. M. Wilson, Astrophys. J. 430, 667 (1994).
[5] G. Gloeckler and J. Geiss, Nature (London) 381, 210 (1996).
[6] K. A. Olive, G. Steigman and E. D. Skillman, Astrophys. J. 483, 788 (1997).
[7] Y. I. Izotov, T. X. Thuan, and V. A. Lipovetsky, Astrophys. J. Suppl. 108, 1 (1997).
[8] S. G. Ryan, T. C. Beers, C. P. Deliyannis and J. A. Thorburn, Astrophys. J. 458, 543 (1996).
[9] P. Bonifacio and P. Molaro, Mon. Not. Roy. Astron. Soc. 285, 847 (1997).
[10] P. Molaro, astro-ph/9709245, in From Quantum Fluctuations to Cosmological Structures, Proc. 1st Moroccan International Astronomy School (Casablanca, 1996), edited by D. Valls-Gabaud et al., ASP Conf. Ser. 126, p. 103.
[11] C. Hogan, astro-ph/9712031, in Primordial Nuclei and their Galactic Evolution, Proc. ISSI Workshop (Bern, 1997) edited by N. Prantzos, M. Tosi, and R. von Steiger (Kluwer, Dordrecht), to appear.
[12] S. A. Levshakov, W. H. Kegel, and F. Takahara, astro-ph/9712139, in Toward the first light of HDS, Proc. SUBARU HDS Workshop (Tokyo, 1997), to appear.
[13] J. A. Bernstein, L. S. Brown, and G. Feinberg, Rev. Mod. Phys. 61, 25 (1989).
[14] R. Esmailzadeh, G. D. Starkman, and S. Dimopoulos, Astrophys. J. 378, 504 (1991).
[15] R. V. Wagoner, Astrophys. J. 179, 343 (1973).
[16] L. Kawano, Report No. Fermilab-Pub-88/34-A, 1988 (unpublished); Report No. Fermilab-Pub-92/04-A, 1992 (unpublished).
[17] M. S. Smith, L. H. Kawano, and R. A. Malaney, Astrophys. J. Suppl. 85, 219 (1993).
[18] K. A. Olive, astro-ph/9712160, in Proc. TAUP’97, 5th International Workshop on Topics in Astroparticle and Underground Physics (Laboratori Nazionali del Gran Sasso, Assergi, 1997) to appear.
[19] S. Sarkar, Rep. Prog. Phys. 59 (1996), 1493.
[20] L. M. Krauss and P. Romanelli, Astrophys. J. 358, 47 (1990).
[21] P. J. Kernan and L. M. Krauss, Phys. Rev. Lett. 72, 3309 (1994).
[22] P. J. Kernan and L. M. Krauss, Case Western Reserve University report No. CWRU-P2-94, astro-ph/9402010. This report, published in an abridged version as Ref. 21, contains additional numerical results relevant for comparison with our work.
[23] L. M. Krauss and P. J. Kernan, Astrophys. J. Lett. 432, 79 (1994); Phys. Lett. B 347, 347 (1995).
[24] C. J. Copi, D. N. Schramm, and M. S. Turner, Science 267, 192 (1995); Phys. Rev. Lett. 75, 3981 (1995).
[25] N. Hata, R. J. Scherrer, G. Steigman, D. Thomas, T. P. Walker, S. Bludman, and P.
Langacker, Phys. Rev. Lett. 75, 3977 (1995); N. Hata, R. J. Scherrer, G. Steigman, D. Thomas, and T. P. Walker, Astrophys. J. 458, 637 (1996); N. Hata, G. Steigman, S. Bludman, and P. Langacker, Phys. Rev. D 55, 540 (1997).

[26] K. A. Olive and D. Thomas, Astropart. Phys. 7, 27 (1997). See also B. D. Fields and K. A. Olive, Phys. Lett. B 368, 103 (1996); B. D. Fields, K. Kainulainen, K. A. Olive, and D. Thomas, New Astron. 1, 77 (1996).

[27] R. M. Barnett et al., Phys. Rev. D 54, 1 (1996), and 1997 off-year partial update for the 1998 edition available on the Particle Data Group WWW pages (URL: http//pdg.lbl.gov/). This update includes the revised Penning-trap measurement of Ref. [28].

[28] \( \tau_n = 889.2 \pm 4.8 \) s, J. Byrne, et al., Europhys. Lett. 33, 187 (1996). Here the authors revise and replace their previous estimate (\( \tau_n = 893.5 \pm 5.3 \) s) which appeared in J. Byrne et al., Phys. Rev. Lett. 65, 289 (1990).

[29] W. T. Eadie, D. Drijard, F. E. James, M. Roos, and B. Sadoulet, *Statistical Methods in Experimental Physics* (North Holland, Amsterdam and London, 1971).

[30] P. J. Kernan, PhD thesis, Ohio State University (1993).

[31] R. Lopez, private communication.

[32] J. N. Bahcall and R. N. Ulrich, Rev. Mod. Phys. 60, 297 (1988); J. N. Bahcall and W. C. Haxton, Phys. Rev. D 40, 931 (1989); X. Shi and D. N. Schramm, Phys. Lett. B 283, 305 (1992); L. M. Krauss, E. Gates, and M. White, Phys. Lett. B 299, 94 (1993).

[33] G. L. Fogli and E. Lisi, Astropart. Phys. 3, 185 (1995); N. Hata and P. Langacker, Phys. Rev. D 50, 632 (1994); E. Gates, L. M. Krauss, and M. White, Phys. Rev. D 51, 2631 (1995).

[34] P. J. Kernan and S. Sarkar, Phys. Rev. D 54, R3681 (1996); S. Sarkar, astro-ph/9611232, in *Aspects of Dark Matter in Astro- and Particle Physics*, edited by H. V. Klapdor-Kleingrothaus and Y. Ramachers (World Scientific, 1997) p. 235.

[35] L. Knox, Phys. Rev. D 52 4307 (1995); G. Jungman, M. Kamionkowski, A. Kosowskii, and D. N. Spergel, Phys. Rev. D 54 1332 (1996); M. Zaldarriaga, D. N. Spergel and U. Seljak, Astrophys. J. 488 1 (1997); J. R. Bond, G. Efstathiou and M. Tegmark, Mon. Not. R. Astron. Soc. 291, L33 (1997).

[36] K. Sato and M. Kobayashi, Prog. Theor. Phys. 58, 1775 (1977); D. A. Dicus, E. W. Kolb, V. L. Teplitz and R. V. Wagoner, Phys. Rev. D 17, 1529 (1978).

[37] J. G. Bartlett and L. Hall, Phys. Rev. Lett. 66, 541 (1991).

[38] J. Ellis, G. B. Gelmini, D. V. Nanopoulos, J. Lopez and S. Sarkar, Nucl. Phys. B 373, 399 (1992); W. Hu and J. Silk, Phys. Rev. Lett. 70, 2661 (1993).

[39] M. Birkel and S. Sarkar, Phys. Lett. B 408, 59 (1997). See also D. J. Eisenstein, W. Hu, J. Silk and A. Szalay, Astrophys. J. Lett. 494, L1 (1998).

[40] G. Fiorentini, E. Lisi, S. Sarkar, and F. L. Villante, in preparation.
FIG. 1. Primordial abundances $Y_i(\eta)$ (solid lines) and their $\pm 2\sigma$ bands (dashed lines), as functions of the nucleon-to-photon ratio $\eta$. 

FIGURES

Light element abundances $Y_i(\eta) \pm 2\sigma_i(\eta)$
FIG. 2. Logarithmic derivatives $\lambda_{ik}(\eta)$ of the abundances $Y_i$ with respect to the reaction rates $R_k$, as functions of $\eta$. 

Logarithmic derivatives $\lambda_{ik}(\eta) = \frac{\partial \ln Y_i(\eta)}{\partial \ln R_k}$
FIG. 3. Fractional abundance uncertainties $\sigma_i/Y_i$ (upper panel) and their correlations $\rho_{ij}$ (lower panel), as functions of $\eta$. 
FIG. 4. Monte Carlo estimates of $\sigma_i/Y_i$ (SKM '93 [17], dots) and $\rho_{ij}$ (KK '94 [22], dots), compared with our analytic evaluation (solid lines), using the same inputs.
FIG. 5. Standard BBN predictions (dotted lines) in the 2-dimensional planes defined by the abundances $Y_{2}$, $Y_{4}$, and $Y_{7}$, as functions of $x \equiv \log_{10}(\eta/10^{-10})$. The theoretical uncertainties are depicted as $1\sigma$ error ellipses at $x = 0, 0.1, 0.2, \ldots, 1$. The crosses indicate the two observational data sets (with $1\sigma$ errors).
Fig. 6. Our $\chi^2$ fit to the data sets A and B, including observational and (correlated) theoretical errors.
FIG. 7. Individual contributions of different reaction rates $R_k$ to the uncertainties in $Y_2$, $Y_4$, and $Y_7$, normalized to the corresponding total errors $\sigma_2$, $\sigma_4$, and $\sigma_7$, for $\eta = 1.78 \times 10^{-10}$. Each arrow corresponds to the shift $\delta Y_i$ induced by a $+1\sigma$ shift of $R_k$. Some small error components have not been plotted.
FIG. 8. Same as Fig. 7, but for $\eta = 5.13 \times 10^{-10}$. 

$\delta Y_1 / \sigma_1$

$\delta Y_2 / \sigma_2$

$\delta Y_3 / \sigma_3$

$\delta Y_4 / \sigma_4$

$\delta Y_5 / \sigma_5$

$\delta Y_6 / \sigma_6$

$k = 1 \ n \ \text{decay}$

$2 \ p(\gamma) d$

$3 \ d(\gamma) ^3 \text{He}$

$4 \ d(d,n) ^3 \text{He}$

$5 \ d(d,p) t$

$6 \ t(d,n) ^4 \text{He}$

$7 \ t(\alpha,\gamma) ^7 \text{Li}$

$8 \ ^3 \text{He}(r,p) t$

$9 \ ^3 \text{He}(d,p) ^4 \text{He}$

$10 \ ^3 \text{He}(\alpha,\gamma) ^7 \text{Be}$

$11 \ ^7 \text{Li}(p,\alpha) ^4 \text{He}$

$12 \ ^7 \text{Be}(n,p) ^7 \text{Li}$