Extended Modified Observable Technique for a Multi-Parametric Trilinear Gauge Coupling Estimation at LEP II

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Abstract

This paper describes the extension of the Modified Observables technique in estimating simultaneously more than one Trilinear Gauge Couplings. The optimal properties, unbiasedness and consistent error estimation of this method are demonstrated by Monte Carlo experimentation using $t\nu jj$ four-fermion final state topologies. Emphasis is given in the determination of the expected sensitivities in estimating the $\lambda_{\gamma} - \Delta g_1^\gamma$ and $\Delta k_{\gamma} - \Delta g_1^T$ pair of couplings with data from the 183 GeV LEP II run.
1 Introduction

It has been shown [1], that by expanding the probability distribution function (p.d.f.) and keeping only linear terms with respect to the Trilinear Gauge Couplings (TGC’s), one can build estimators (the Optimal Observables) which are linear functions of the couplings around the expansion point. Furthermore, this linear dependence can be easily evaluated by theory. An efficient estimation of the couplings can be performed by inverting these linear relations. Such an estimation has the same accuracy as the unbinned maximum likelihood technique.

The method of the Optimal Observables has been extended [2] to incorporate the influence of the detector effects to the measurement of the kinematical vectors. An iterative procedure has been also introduced to ensure the consistency and optimality of the technique, independent of the choice of the parametric expansion point. In the same paper, the optimal properties of this (Modified Observables) method have been demonstrated for one coupling fits to the 172 GeV LEPII data. In the meanwhile larger data samples are available from the 183 GeV LEPII run and the application of the Modified Observable technique to a simultaneous estimation of two couplings is very relevant.

This paper concentrates on the simultaneous estimation of two TGC’s by employing phenomenological models [3] where two couplings could deviate freely from their Standard Model (S.M.) values whilst certain constraints are imposed on the other couplings. This paper is dealing with WW events produced in $e^+e^-$ annihilation, where one of the $W$’s decays leptonically whilst the other decays in two jets. A large sample of 60000 $WW \rightarrow ℓνq\bar{q}$ Monte Carlo (M.C.) events was used to evaluate cross sections and other statistics, as well as their dependence on the coupling values by the M.C. reweighting procedure [4]. These events have been produced either by PYTHIA [5] (employing only the CC03 production diagrams) or by EXCALIBUR [6] (full 4-fermion production) at different coupling values and they have undergone full detector simulation by the DELSIM [7] simulation programme. Moreover these events have been reconstructed and selected by the same analysis algorithms as the real data [8] [9] accumulated with the DELPHI [10] detector. The background contamination has been simulated by the production of the physics channels [8] [9] [11] which produce final state topologies indistinguishable from the signal WW events.

This paper is organised as follows: the statistical technique and its asymptotic properties are described in Section 2, whilst numerical results obtained by M.C. experimentation are presented in Section 3. Finally, Section 4 contains the comparison with other techniques and the conclusions.

2 Modified Observables in Multi-Parametric Fits

The present study is focusing on two parameter (TGC’s) estimations but this analysis can be extended to any number of parameters in a straightforward way.

The probability distribution function, with respect to the observed kinematical vector $\vec{Ω}$, is expressed [2] [3] as a function of the two couplings $\alpha_1$ and $\alpha_2$ as
\[ P(\vec{\Omega}; \alpha_1, \alpha_2) = \int \frac{d\sigma(\vec{V}; \alpha_1, \alpha_2)}{\sigma_{\text{tot}}(\alpha_1, \alpha_2)} \cdot \epsilon(\vec{V}) \cdot R(\vec{V}; \vec{\Omega}) \cdot d\vec{V} \]

\[ = \int \frac{\sum_{i=0}^{2} \sum_{j=0}^{2} c_{ij}(\vec{V}) \alpha_i^j \alpha_2^j}{\sum_{i=0}^{2} \sum_{j=0}^{2} S_{ij} \alpha_i^j \alpha_2^j} \cdot \epsilon(\vec{V}) \cdot R(\vec{V}; \vec{\Omega}) \cdot d\vec{V} \]  

(1)

where

\[ \vec{V} \] is the true kinematical vector which describes the events

\[ \epsilon(\vec{V}) \] is the efficiency of observing an event produced at \( \vec{V} \)

\[ R(\vec{V}; \vec{\Omega}) \] is the resolution function, i.e. the probability the true kinematical vector \( \vec{V} \) to be measured as \( \vec{\Omega} \)

\[ \frac{d\sigma(\vec{V}; \alpha_1, \alpha_2)}{d\vec{V}} \] is the differential cross-section

\[ \sigma_{\text{tot}}(\alpha_1, \alpha_2) \] is the total cross-section and

\[ S_{ij} = \int c_{ij}(\vec{V})d\vec{V}. \]

In [2], it has been shown that the Optimal Observables including detector effects, in the neighbourhood of the parametric point \( \{\alpha_1^0, \alpha_2^0\} \), are defined as the mean values of the following quantities:

\[ z_1(\vec{\Omega}; \alpha_1^0, \alpha_2^0) = \frac{\int y_{00}(\vec{V}; \alpha_1^0, \alpha_2^0)\epsilon(\vec{V})R(\vec{V}; \vec{\Omega})d\vec{V}}{\int y_{00}(\vec{V}; \alpha_1^0, \alpha_2^0)\epsilon(\vec{V})R(\vec{V}; \vec{\Omega})d\vec{V}} \]

\[ z_2(\vec{\Omega}; \alpha_1^0, \alpha_2^0) = \frac{\int y_{10}(\vec{V}; \alpha_1^0, \alpha_2^0)\epsilon(\vec{V})R(\vec{V}; \vec{\Omega})d\vec{V}}{\int y_{00}(\vec{V}; \alpha_1^0, \alpha_2^0)\epsilon(\vec{V})R(\vec{V}; \vec{\Omega})d\vec{V}} \]

(2)

where the functions \( y_{k,\lambda}(\vec{V}; \alpha_1^0, \alpha_2^0) \) are expressed in terms of the differential cross section coefficients as:

\[ y_{00}(\vec{V}; \alpha_1^0, \alpha_2^0) = c_{00}(\vec{V}) + c_{01}(\vec{V})\alpha_1^0 + c_{10}(\vec{V})\alpha_2^0 + c_{20}(\vec{V})\alpha_1^0\alpha_2^0 + c_{02}(\vec{V})\alpha_2^0 + c_{11}(\vec{V})\alpha_1^0\alpha_2^0 \]

\[ y_{10}(\vec{V}; \alpha_1^0, \alpha_2^0) = c_{10}(\vec{V}) + 2c_{20}(\vec{V})\alpha_1^0 + c_{11}(\vec{V})\alpha_2^0 \]

\[ y_{01}(\vec{V}; \alpha_1^0, \alpha_2^0) = c_{01}(\vec{V}) + 2c_{02}(\vec{V})\alpha_2^0 + c_{11}(\vec{V})\alpha_1^0 \]

(3)

It has also been shown that the Optimal Observables are linear functions of the couplings \( \alpha_1 \) and \( \alpha_2 \) in the neighbourhood of \( \{\alpha_1^0, \alpha_2^0\} \) i.e.:

\[ \int z_k(\vec{\Omega}; \alpha_1^0, \alpha_2^0)P(\vec{\Omega}; \alpha_1, \alpha_2)d\vec{\Omega} = \int z_k(\vec{\Omega}; \alpha_1^0, \alpha_2^0)P(\vec{\Omega}; \alpha_1^0, \alpha_2^0)d\vec{\Omega} + \]

\[ \sum_{i=1}^{2} \int z_i(\vec{\Omega}; \alpha_1^0, \alpha_2^0)z_k(\vec{\Omega}; \alpha_1^0, \alpha_2^0)P(\vec{\Omega}; \alpha_1^0, \alpha_2^0)d\vec{\Omega} - \]

\[ \left( \int z_i(\vec{\Omega}; \alpha_1^0, \alpha_2^0)P(\vec{\Omega}; \alpha_1^0, \alpha_2^0)d\vec{\Omega} \right) \cdot (\alpha_i - \alpha_i^0) \]

(4)

1This is easily proven by expanding \( \vec{\Omega} \) in a Taylor series around \( \{\alpha_1^0, \alpha_2^0\} \) and evaluating the mean values of \( \vec{\Omega} \) ignoring higher than first order in \( \alpha_1^0 \) and \( \alpha_2^0 \).
where \( k = 1, 2 \).

Thus, given a set of \( N \) experimentally measured vectors \( \vec{\Omega}_n \) (\( n = 1, \ldots, N \)) the left hand side of (4) can be approximated as:

\[
\int z_k(\vec{\Omega}; \alpha_1^0, \alpha_2^0) \vec{P}(\vec{\Omega}; \alpha_1, \alpha_2) d\vec{\Omega} \approx \frac{1}{N} \sum_{n=1}^{N} z_k(\vec{\Omega}_n; \alpha_1^0, \alpha_2^0)
\]

(5)

The right hand side of (4) can be calculated using the theoretical expression of the cross section as a function of the couplings, provided that the resolution and efficiency functions can be parametrized analytically. Then, a simple inversion of the linear system of equations (4) results in an estimation of the coupling values with the same sensitivity as with the maximum likelihood technique.

In practice, neither the efficiency nor the resolution function can be parametrized analytically. Then, a simple inversion of the linear system of equations (4) results in an estimation of the coupling values with the same sensitivity as with the maximum likelihood technique.

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The couplings are estimated by comparing the calibration surfaces to the experimental measurements, that is to the products of the measured values of the Modified Observables with the number of observed events, which are simply expressed as:

\[ d_1(\alpha_1^0, \alpha_2^0) = \sum_{i=1}^{N} z_1(\tilde{\Omega}_i; \alpha_1^0, \alpha_2^0) \]
\[ d_2(\alpha_1^0, \alpha_2^0) = \sum_{i=1}^{N} z_2(\tilde{\Omega}_i; \alpha_1^0, \alpha_2^0) \]

Such comparisons are shown in figures (3) and (4) between a large independent set of M.C. events used as a data sample and three pairs of calibration surfaces \( (f_1(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \text{ and } f_2(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0)) \) evaluated at three different expansion points \( \{\alpha_1^0, \alpha_2^0\} \). In these figures the intersections of the calibration surfaces with the planes defined by the experimental measurements, \( d_1(\alpha_1^0, \alpha_2^0) \) and \( d_2(\alpha_1^0, \alpha_2^0) \), are also shown. It is worth noticing that the estimation, which is the common point of the pair of lines in figures (3), (4) and (6), is independent from the expansion point. This fact reflects one of the basic properties of the technique to be globally unbiased.

However, the evaluation of the estimation confidence intervals is more complicated, due to the statistical correlations between the calibration surfaces as well as between the measured quantities \( d_1(\alpha_1^0, \alpha_2^0) \) and \( d_2(\alpha_1^0, \alpha_2^0) \).

The covariant matrices \( M(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \) (expressing the statistical accuracy of the calibration surface evaluation at the expansion point \( \{\alpha_1^0, \alpha_2^0\} \) ) and \( V(\alpha_1^0, \alpha_2^0) \) (which is the covariant matrix corresponding to the measured quantities \( d_1(\alpha_1^0, \alpha_2^0) \) and \( d_2(\alpha_1^0, \alpha_2^0) \)) are calculated from the kinematical vectors of the reweighted M.C. and real events respectively.

Then, assuming gaussian errors, the probability that the selected event sample supports coupling values equal to \( \alpha_1 \) and \( \alpha_2 \), is given by the Likelihood function:

\[
L = \frac{1}{2\pi |W|} \cdot \exp[-\frac{1}{2} \left( \bar{D}(\alpha_1^0, \alpha_2^0) - \bar{F}(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \right)^T \cdot W^{-1} \cdot \left( \bar{D}(\alpha_1^0, \alpha_2^0) - \bar{F}(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \right)]
\]

where the vector \( \bar{D}(\alpha_1^0, \alpha_2^0) \), the vector calibration function \( \bar{F}(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \) and the covariant matrix \( W(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \) are defined as follows:

\[
\bar{D}(\alpha_1^0, \alpha_2^0) = \begin{pmatrix} d_1(\alpha_1^0, \alpha_2^0) \\ d_2(\alpha_1^0, \alpha_2^0) \end{pmatrix}
\]
\[
\bar{F}(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) = \begin{pmatrix} f_1(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \\ f_2(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \end{pmatrix}
\]
\[
W(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) = M(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) + V(\alpha_1^0, \alpha_2^0)
\]

Maximization of (8), with respect to \( \alpha_1 \) and \( \alpha_2 \), provides the estimation of the coupling values, whilst the confidence intervals are evaluated by the asymptotic gaussian properties

\[ -(\log L_{max} - 1.205) \text{ for 70% confidence intervals} \]
of the estimation distribution \[13\].

A set of M.C. events produced with Standard Model coupling values (6000 events at \(\{\lambda_\gamma = 0, \Delta g_Z^1 = 0\}\)), was used as data sample to demonstrate the asymptotic properties of such estimations. The \(\lambda_\gamma, \Delta g_Z^1\) couplings were simultaneously estimated by maximizing the likelihood function of (8) and the estimated coupling values are shown as functions of the expansion point in figure [3]. The fact that the estimations are close (within the statistical errors) to the true coupling values, for the whole region of the expansion points, emphasizes the optimal properties of the method. However, the optimal estimated error is achieved [1] at expansion points close to the estimated values, where the linear dependance of the Optimal Variables holds. This is shown in figure [3c] where three 70% confidence limit contours corresponding to different expansion points are presented for comparison. Obviously the optimum estimated sensitivity is achieved in the case where \(\alpha_0^1 = \hat{\alpha}_1\) and \(\alpha_0^2 = \hat{\alpha}_2\), where \(\hat{\alpha}_1\) and \(\hat{\alpha}_2\) are the estimated values. The fact that the above condition also guarantees a correct error estimation, is demonstrated in the next section by Monte Carlo experimentation.

### 3 Numerical results

A series of M.C. experiments were used to demonstrate the optimal properties of the Modified Observable technique when two TGC’s are simultaneously estimated by fitting finite statistical samples.

Fully reconstructed four fermion EXCALIBUR events, produced with S.M. coupling values, were mixed with background events to form data sets corresponding to the luminosity of 50.23 \(pb^{-1}\) accumulated by the DELPHI detector at \(\sqrt{s} \approx 183\) GeV. Each set consisted of 82, 101 and 39 events in average with an electron, muon and tau lepton in the final state, respectively. The average background contribution to each of the above subsets were 8.0, 1.4 and 8.3 events. The specific event multiplicity of each data set was chosen to follow poissonian distributions. Another set of fully four fermion and background reconstructed events, produced and selected as described in Section 1, was used to calculate cross sections and probabilties as well as their dependence on the TGC’s by reweighted Monte Carlo integration. In fitting the data sets, the \((\lambda_\gamma, \Delta g_Z^1)\) and the \((\Delta k_\gamma, \Delta g_Z^1)\) TGC schemes were used [3] and a simultaneous estimation of the free couplings was performed.

In order to take into account the differences in the production dynamics, the selection efficiencies and the background contamination between the final states \((\ell\nu jj, \ell = \mu, e, \tau)\) the measured vector \(\vec{D}(\alpha_0^1, \alpha_0^2)\) was defined as follows:

\[
\vec{D}(\alpha_0^1, \alpha_0^2) = \left( \frac{\sum_{\ell=1}^{3} (d_1^\ell(\alpha_0^1, \alpha_0^2) - B_1^\ell)}{\sum_{\ell=1}^{3} f_2^\ell(\alpha_0^1, \alpha_0^2)} \right) \tag{12}
\]

Where \(\ell = 1, 2, 3\) stands for the three lepton tags whilst \(B_1^\ell, B_2^\ell\) denotes the expected contribution of the background events to the measurement.

Similarly the calibration surface vector was defined as:

\[
\vec{F}(\alpha_1, \alpha_2; \alpha_0^1, \alpha_0^2) = \left( \frac{\sum_{\ell=1}^{3} f_1^\ell(\alpha_1, \alpha_2; \alpha_0^1, \alpha_0^2)}{\sum_{\ell=1}^{3} f_2^\ell(\alpha_1, \alpha_2; \alpha_0^1, \alpha_0^2)} \right) \tag{13}
\]
The asymptotic property of the log likelihood ratio \[13\] was used to demonstrate the unbiasedness of the proposed techniques. That is, the \(\chi^2\) (n.d.f.=2) probability of obtaining the specific value of \(\lambda\), where

\[
\lambda = -2 \cdot \log \frac{L(\alpha_1^{\text{true}}, \alpha_2^{\text{true}})}{L(\hat{\alpha}_1, \hat{\alpha}_2)}
\]

(14)
in fits of the data sets should follow an equiprobable distribution. Furthermore, the consistency in evaluating the error matrix of the estimated couplings \((\hat{\mathcal{E}})\) in every fit, is checked by using the other asymptotic property \[13\] of the likelihood estimations to be gaussian distributed around the true parameter values. Thus for an unbiased estimation of the central values and for a correct error matrix evaluation the quantity \(\delta\):

\[
\delta = \left( \frac{\hat{a}_1 - a_1^{\text{true}}}{\hat{a}_2 - a_2^{\text{true}}} \right) \cdot \hat{\mathcal{E}} \cdot \left( \frac{\hat{a}_1 - a_1^{\text{true}}}{\hat{a}_2 - a_2^{\text{true}}} \right)
\]

(15)
should follow a \(\chi^2\) (n.d.f.=2) distribution. This property is demonstrated by presenting the \(\chi^2\) (n.d.f.=2) probabilities to obtain specific \(\delta\) values in fitting the data sets.

The above tests of \(\lambda\) and \(\delta\) \(\chi^2\)-probability distributions can be considered as extensions of the sampling and pull distribution tests respectively, commonly used in one parameter fits.

Due to the limited number of the available M.C. events, only sixty independent data sets could be constructed. Although the number of the data sets is enough to show the optimal properties of the proposed technique, the bootstrap procedure \[14\] has been also used to construct a large number of semicorrelated data sets.

Results of estimating the \((\lambda_\gamma, \Delta g_1^Z)\) and \((\Delta k_\gamma, \Delta g_1^Z)\) couplings with the Modified Observable technique are shown in figure 6. In both TGC schemes the optimal properties of the technique in estimating central values and error matrices are obvious. Specifically the sixty completely uncorrelated samples produce \(\chi^2\) (n.d.f. = 2) probabilities distributed with mean values close to 0.5 and root mean squares close to 1/\(\sqrt{12}\), whilst the equiprobable behaviour of the \(\chi^2\) (n.d.f. = 2) probability values obtained by fitting the bootstrapped samples is striking.

The \(\chi^2\) behaviour of the \(\lambda\) and \(\delta\) quantities is further used to quantify the sensitivity of this technique. Indeed such property \[13\] ensures that the estimated values \(\{\hat{\alpha}_1, \hat{\alpha}_2\}\) follow a two dimensional gaussian distribution with a covariant matrix which characterises the average sensitivity in estimating the couplings. The covariant matrix elements (i.e. the variances and correlations of the couplings estimations) are found by fitting a 2-dim gaussian to the estimated coupling values from the 60 independent sets. These average sensitivities are summarized in Tables 1 and 2 for \((\lambda_\gamma, \Delta g_1^Z)\) and \((\Delta k_\gamma, \Delta g_1^Z)\) estimations. The same uncorrelated M.C. sets of events were treated as if they have been collected by a “perfect” detector and the two pairs of couplings were estimated by an unbinned extended likelihood fit as well as by the Modified Observable technique \[14\]. The average sensitivities

\[4\text{The bootstrap procedure advocates that one can select randomly } N \text{ events to form a set from a pool of } K \text{ available events, and repeat the random selection to construct many bootstrapped sets. The distribution of statistics evaluated from each of the bootstrapped set approximates well the true distribution, as long as } K \text{ is big enough compared to } N.\]

\[5\text{The true kinematical vector } \vec{V} \text{ of each event was used to calculate the matrix element and the calibration surfaces. In the following, when } \vec{V} \text{ is used, the methods and their results will be named as “perfect”.}\]
obtained from these estimations (“perfect” detector extended unbinned likelihood and “perfect” detector Modified Observables) are also shown for comparison in Tables 1 and 2 where the equivalence of the Modified Observables to the likelihood fit is obvious. The loss of sensitivity in the case of a realistic detector is a natural consequence of the loss of information due to the imperfect experimental resolution. However, the consistent inclusion of the detector effects in the realistic case guarantees consistent central value and confidence interval estimation. It is also worth noticing that in the realistic case, the evaluated errors and correlations in every individual Modified Observable estimation are gaussian distributed with means very close to the average sensitivities, as it is shown in figure 7.

As a last point, figure 8 shows the sampling, the pull and the error distributions of a single coupling (∆kγ) estimation. Similar estimations of the same coupling [2], using 172 GeV data samples, have been found to exhibit non gaussian tails. However, it was advocated that with small event samples, where the statistical error is large compared to the linear part of the calibration curves, the evaluated error from the fits is expected to underestimate the sensitivity of the technique. Obviously such pathologies are absent when the relative statistical error is smaller, as in the case of the data sample accumulated at 183 GeV.

4 Conclusion

In this paper the Modified Observables technique [2] was generalized in order to be applied for a simultaneous estimation of two couplings by deploying the appropriate TGC scheme [3]. The technique, including the detector effects and the background contribution, was demonstrated to be asymptotically a consistent estimator. This consistency was also shown to be independent of the initial expansion values. However the optimal sensitivity is achieved for expansion points close to the estimated values of the couplings.

The properties of the technique, when fitting finite size event samples, were investigated by M.C. experimentation. Sets of M.C. events, of the same size as the data samples accumulated by each of the LEP experiments at √s ≃ 183 GeV, were fitted to estimate the {λγ, ∆g1} and {Δκγ, ∆g2} pairs of couplings. The distributions of these estimations demonstrated the optimal behaviour (unbiasedness, consistent error matrix evaluation) of the technique. Moreover a comparison with the unbinned extended likelihood results shows that the Modified Observable estimators are practically reaching the maximum sensitivity. These two methods are completely equivalent at the “perfect” detector case (tables 1 and 2). A deterioration of the sensitivity (up to 20%) when dealing with realistic detectors is due to the imperfect resolution of the measuring apparatus.

A comparison [9] between the sensitivity of several multiparametric TGC estimators, which include detector effects, shows that the Modified Observables are equivalent to the Iterative Optimal Variables and Multidimensional Clustering techniques [15] whilst outperform classical methods of one or two dimensional binned likelihood fits.
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**Table 1:** Comparison of the statistical properties of the technique proposed in this paper with the unbinned extended likelihood estimations.

| Method                                | \(\lambda_\gamma - \Delta g^z_i\) | \(\sigma_{\lambda_\gamma}\) | \(\sigma_{\Delta g^z_i}\) | \(\rho\)         |
|---------------------------------------|-----------------------------------|-----------------------------|-----------------------------|-------------------|
| “Perfect” Extended Likelihood         |                                   | 0.21 ± 0.01                 | 0.20 ± 0.01                 | -0.73 ± 0.06      |
| “Perfect” Modified Observables        |                                   | 0.22 ± 0.01                 | 0.21 ± 0.01                 | -0.74 ± 0.06      |
| Modified Observables                  |                                   | 0.25 ± 0.01                 | 0.23 ± 0.01                 | -0.74 ± 0.06      |

**Table 2:** Comparison of the statistical properties of the techniques proposed in this paper with the unbinned extended likelihood estimations.

| Method                                | \(\Delta k_\gamma - \Delta g^z_i\) | \(\sigma_{\Delta k_\gamma}\) | \(\sigma_{\Delta g^z_i}\) | \(\rho\)         |
|---------------------------------------|-----------------------------------|-----------------------------|-----------------------------|-------------------|
| “Perfect” Extended Likelihood         |                                   | 0.35 ± 0.03                 | 0.14 ± 0.01                 | -0.22 ± 0.08      |
| “Perfect” Modified Observables        |                                   | 0.38 ± 0.03                 | 0.13 ± 0.01                 | -0.25 ± 0.09      |
| Modified Observables                  |                                   | 0.44 ± 0.03                 | 0.15 ± 0.01                 | -0.28 ± 0.10      |
Figure 1: The mean values (see text) of the quantities $y_1^{01}$ and $y_1^{00}$ as functions of the $y_0^{00}$ and $y_0^{01}$ respectively. These mean values correspond to kinematical vectors produced with couplings:

- $\alpha_1 \equiv \Delta g_1 = 0$, $\alpha_2 \equiv \lambda_\gamma = 0$ in (a) and (b)
- $\alpha_1 \equiv \Delta g_1 = 0$, $\alpha_2 \equiv \lambda_\gamma = -1$ in (c) and (d)
- $\alpha_1 \equiv \Delta g_1 = 0$, $\alpha_2 \equiv \lambda_\gamma = +1$ in (e) and (f)
Figure 2: The $f_1(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0)$, $f_2(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0)$ calibration surfaces for several expansion points $\alpha_1^0(\equiv \Delta g^1) = 0$, $\alpha_2^0(\equiv \lambda_1) = 0$ in (a) and (b), $\alpha_1^0(\equiv \Delta g^1) = 0$, $\alpha_2^0(\equiv \lambda_1) = +1$ in (c) and (d), $\alpha_1^0(\equiv \Delta g^1) = 0$, $\alpha_2^0(\equiv \lambda_1) = -1$ in (e) and (f).
Figure 3: The calibration surfaces $f_1(\alpha_1, \alpha_2; \alpha_0^0, \alpha_2^0)$, $f_2(\alpha_1, \alpha_2; \alpha_1^0, \alpha_0^0)$ as functions of the couplings $\alpha_1 = \Delta g_1^2$, $\alpha_2 = \lambda$, at the expansion point $\{\alpha_1^0 = 0, \alpha_2^0 = 0\}$. The horizontal shadowed planes correspond to the experimental measurements. The two lines representing the intersection of the calibration surfaces with the measured values are shown in (c).
Figure 4: The calibration surfaces \( f_1(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \), \( f_2(\alpha_1, \alpha_2; \alpha_1^0, \alpha_2^0) \) as functions of the couplings \( \alpha_1 = \Delta g_1 \), \( \alpha_2 = \lambda \), at the expansion points \( \{\alpha_1^0 = 0, \alpha_2^0 = -1\} \) in [(a),(b)] and \( \{\alpha_1^0 = 0, \alpha_2^0 = +1\} \) in [(d),(e)]. The horizontal shadowed planes correspond to the experimental measurements. The two lines representing the intersection of the calibration surfaces with the measured values are shown in (c) and (f) for the two pairs of expansion points, respectively.
Figure 5: In [(a) and (b)] the estimated $\alpha_1 = \Delta g_i$ and $\alpha_2 = \lambda_i$ coupling values as functions of the expansion points are shown. In (c) the evaluated 70% C.L. contours for $\alpha_0^1 = 0, \alpha_0^2 = 0$ (solid points), $\alpha_0^1 = 0.2, \alpha_0^2 = -0.6$ (squares) and $\alpha_0^1 = -0.2, \alpha_0^2 = 0.6$ (stars) are compared.
Figure 6: The distributions of $\chi^2$ (n.d.f.=2) probabilities in obtaining $\lambda$ [(a),(b),(e),(f)] and $\delta$ [(c),(d),(g),(h)] values in $\{\lambda, \Delta g_1\}$ [(a),(b),(c),(d)] and $\{\Delta \kappa, \Delta g_1\}$ [(e),(f),(g),(h)] estimations by the Modified Observables technique. The lines with slopes consistent with zero in (b),(d),(f) and (h) are first degree polynomial fits to the bootstrap results.
Figure 7: Confidence interval estimations with the Modified Observables technique. The distributions of errors [(a),(b),(d),(e)] and correlations [(c),(f)] in estimating the \{\lambda, \Delta g_i^1\} [(a),(b),(c)] and \{\Delta \kappa, \Delta g_i^1\} [(d),(e),(f)] pair of couplings. The data points correspond to the 60 independent data sets whilst the histograms to the bootstrap results. The arrows indicate the average sensitivities summarized in Tables 1 and Table 2.
Figure 8: The sampling, pull and error distribution of $\Delta \kappa_\gamma$ estimation with the Modified Observable technique.