



Systematics Uncertainties in Small-scale Experiments

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Disclaimer

- Small-scale experiments -> Single-purpose experiments
 - e.g. Muon g-2 experiment, ~ 200 authors
- Technical aspects already covered in previous lectures, I will concentrate on practical applications
 - A very biased selection of experimental examples, mostly from muon and QED precision physics
- Not intended as a review of physics results
 - you could find incomplete references, out-of-date results, etc.

Outline

- General aspects of systematic uncertainties in singlepurpose experiments
 - precision measurements
 - rare event searches
- Inclusion of systematics in confidence interval computations
- (Very biased) collection of relevant examples

Generalities

Single-purpose experiments

- Most single-purpose experiments can be classified into two categories:
 - 1) highly accurate measurements of particles properties
 - 2) searches for rare processes (rare decays, interactions of elusive particles)
- Besides accumulating statistics, high sensitivity is achieved through high precision (extremely good resolutions, extremely high background rejection) and/or high accuracy (no bias in measurements and background estimates)

PHYSICAL REVIEW LETTERS 131, 161802 (2023)

Editors' Suggestion

Measurement of the Positive Muon Anomalous Magnetic Moment to 0.20 ppm

Measurement based on the rate of e⁺ from μ⁺ decays, above a given energy threshold (2.5% e⁺ energy resolution)

PHYSICAL REVIEW LETTERS 131, 161802 (2023)	
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Quantity	Correction (ppb)	Uncertainty (ppb)
ω_a^m (statistical) ω_a^m (systematic)		201 25
$C_e C_p$	451 170	32 10
C_{pa}	-27	13
C_{dd} C_{ml}	0	3
$f_{\text{calib}} \cdot \langle \omega'_p(\vec{r}) \times M(\vec{r}) \rangle$		46
$B_k B_q$	-21 -21	20
$\mu'_p(34.7^{\circ})/\mu_e$		11
m_{μ}/m_e $g_e/2$		22 0
Total systematic for \mathcal{R}'_{μ}		70
Total external parameters Total for a_{μ}	622	25 215

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$C_{ml} C_{ml} f_{\text{calib}} \cdot \langle \omega'_p(\vec{r}) \times M(\vec{r}) \rangle \\B_k \\B \\B \\B \\B \\B \\C $	$ \begin{array}{c} 10 \\ 0 \\ \\ -21 \\ -21 \end{array} $	3 46 13 20	
$\mu'_{p}(34.7^{\circ})/\mu_{e}$ m_{μ}/m_{e} $g_{e}/2$	···· ····	11 22 0	
Total systematic for \mathcal{R}'_{μ} Total external parameters Total for a_{μ}	···· ··· 622	70 25 215	



Upper Limit based on the discrimination of 2-body vs. 3-body kinematics with extremely precise measurements (e.g. **0.2%** e⁺ energy resolution)

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Parameter	Impact on limit
$\phi_{e\gamma}$ uncertainty	1.1%
E_{γ} uncertainty	0.9%
$\theta_{e\gamma}$ uncertainty	0.7%
Normalization uncertainty	0.6~%
$t_{e\gamma}$ uncertainty	0.1%
E_e uncertainty	0.1%
RDC uncertainty	< 0.1 %

Eur. Phys. J. C (2024) 84:216 https://doi.org/10.1140/epjc/s10052-024-12416-2	THE EUROPEAN PHYSICAL JOURNAL C
Regular Article - Experimental Physics	
A search for $\mu^+ \rightarrow e^+ \gamma$ with the fire experiment MEG II Collaboration	st dataset of the MEG II

Upper Limit based on the discrimination of 2-body vs. 3-body kinematics with extremely precise measurements (e.g. **0.2%** e⁺ energy resolution)

- Let's suppose to search for a rare process, looking for a peak over the background in a known position of a distribution
 - high sensitivity through high precision



- Let's suppose to search for a rare process, looking for a peak over the background in a known position of a distribution
 - high sensitivity through high precision
 - the measurement has to be also highly accurate (no relevant bias)



 In general-purpose experiments, many non-rare physics processes can be used to calibrate the measurements and remove biases



Source		Impact [MeV]
Photon energy scale		83
$Z \rightarrow e^+e^-$ calibration		59
$E_{\rm T}$ -dependent electron energy scale		44
$e^{\pm} \rightarrow \gamma$ extrapolation		30
Conversion modelling		24
Signal-background interference		26
Resolution	1	15
Background model		14
Selection of the diphoton production ve	rtex	5
Signal model		1
Total		90



- Single-purpose experiments typically have extreme resolution (i.e. need for extreme accuracy) and a scarcity of physics processes to be used for calibrations
 - dedicated tools need to be developed

Rare events with high-precision - the MEG case

- Single-purpose experiments typically have extreme resolution (i.e. need for extreme accuracy) and a scarcity of physics processes to be used for calibrations
 - dedicated tools need to be developed

In the MEG experiment, where a muon beam is used to search for $\mu \rightarrow e\gamma$, a profusion of calibration tools has been developed, including a **dedicated Cockroft-Walton accelerator** and a $\pi^{-}(p, n) \pi^{0}$ **experiment** with the only purpose of calibrating the photon energy reconstruction.

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Systematic Uncertainties for High Accuracy

High Accuracy for zero tests

 $\Delta \stackrel{?}{=} 0$

e.g. particles EDMs

$$\begin{split} \Delta_{meas} &= k\Delta + \delta \quad \text{with } k = 1 \pm \sigma_k, \, \delta = 0 \pm \sigma_\delta \\ \Rightarrow & \sigma_{\Delta}^{syst} = \frac{\sigma_k}{k} \Delta + \sigma_\delta \sim \sigma_\delta \end{split}$$

- Additive uncertainties dominate over multiplicative uncertainties

Systematic Uncertainties for High Accuracy

High Accuracy for non-zero measurements

e.g. particles MDMs, coupling constants

- multiplicative uncertainties are also important
- comparison with SI units is critical —> a metrology problem



Measured dimensional quantities need to be calibrated against SI standards with << ppm accuracy

Systematic Uncertainties in Rare Event Searches

Rare event searches through event patterns

- rare event searches (rare decays, dark matter, etc.) where background rejection is mostly achieved through particle identification, vetos, event topology, etc.
- dominant systematic uncertainties typically from the control of the background rejection efficiency



Systematic Uncertainties in Rare Event Searches

Rare event searches through precision

- rare event searches (e.g. rare decays) where the precise measurement of some observable (e.g. kinematics) is required to discriminate signal and background
- high precision requires also high accuracy, which can be only achieved with dedicated tools



A ognuno il suo (Each to their own)

MEG vs. TWIST

Measurement of parameters of the e+ energy spectrum in μ^+ decays



Measurement of e⁺ energy spectrum for the search of $\mu^+ \rightarrow e^+ \gamma$



Very accurate knowledge of the magnetic field is necessary to measure the spectrum parameters Magnetic field adjusted from data, exploiting the theoretical knowledge of the Michel spectrum

The role of Monte Carlo simulations

- The extremely high resolutions and accuracies of single-purpose experiments pose strong challenges to Monte Carlo simulations
- Additionally, for rare event searches, MC productions resulting in sufficient statistics of reconstructed background events are computationally unachievable or unreliable
 - e.g. MEG II 2021 data

Potential background events	~ 10 ¹²
After the trigger	2 x 10 ⁷
In the analysis region	66
Within 1σ to the signal	< 1

The role of Monte Carlo simulations

 The extremely high resolutions and accuracies of single-purpose experir

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e.g

The use of Monte Carlo simulations is typically very limited and, when MC inputs are unavoidable, the related systematic uncertainties are large

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Potential background events	~ 1012
After the trigger	2 x 10 ⁷
In the analysis region	66
Within 1σ to the signal	< 1

Confidence intervals and systematics

 In FC, the confidence belt is built using the likelihood ratio test statistics, defined as:

$$\mathscr{R}(\mathbf{p}) = \frac{\mathscr{L}(\mathbf{p})}{\mathscr{L}(\hat{\mathbf{p}})}$$

p : set of parameters

 $\hat{\mathbf{p}}$: set of parameters maximizing the likelihood (i.e. fitted value)

- Given a set of values of the parameters \mathbf{p} , the expected distribution of \mathscr{R} is computed
- When the experiment is performed, the value of \mathscr{R} is computed for different hypothetical sets of **p**:
 - a set of **p** is included in a confidence interval at C.L. = 1 α if more than a fraction α of the experiments is expected to give a larger *R*(**p**) than data

Dealing with physical limits on a parameter

• For rare event searches ($p = N_{sig}$), the so-called "conditioning" is also included to properly treat the physical constraint $N_{sig} > 0$:

$$\mathcal{R} = \begin{cases} \frac{\mathcal{L}(N_{\text{sig}})}{\mathcal{L}(N_{\text{sig,best}})} & \text{if } N_{\text{sib,best}} \ge 0\\ \frac{\mathcal{L}(N_{\text{sig}})}{\mathcal{L}(0)} & \text{if } N_{\text{sib,best}} < 0 \end{cases}$$

- The FC approach requires to evaluate the expected distribution of $\mathscr{R}(\mathbf{p})$
- Typically obtained by generating pseudo-experiments (toy Monte Carlo exp.) according to the expected PDFs

- Let's consider the case of:
 - a single variable x
 - a gaussian signal ($\mu = 0, \sigma = 3$)
 - a flat background of 1000 events in x ∈ [-20,20]



Distribution of R from toy MCs generated with different <Nsig>





90% C.L. threshold







Multi-dimensional case

- The original FC problem (neutrino oscillation parameters)
- Counting experiment:
 - the observable is the number of countings nobs (Poisson PDF)
 - there is an expected number of background events, n_{bkg}
 - for each point in the 2D space, there is an expected number of signal events, n_{sig}
 - for each point, the R distribution is derived from toy MCs with Poisson distribution
 - R from data is compared to the R distribution in toy MCs

 $\mathcal{R}(\Delta m^2, \sin^2(2\theta)) = \frac{\text{Poisson}(n_{\text{obs}}; n_{\text{sig}} + n_{\text{bkg}})}{\text{Poisson}(n_{\text{obs}}; n_{\text{sig,best}} + n_{\text{bkg}})}$

with $n_{\text{sig,best}} > 0$



Multi-dimensional case



- Coverage guaranteed for each pair of true (N_{sig}, M)
 - no need of look-elsewhere corrections

Multi-dimensional case



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- The FC approach requires to evaluate the expected distribution of $\mathscr{R}(\mathbf{p})$
- Typically obtained by generating pseudo-experiments (toy Monte Carlo exp.) according to the expected PDFs
- CAVEAT: can be computationally heavy for multidimensional parameter space, complex likelihoods and very small p-values:
 - for a 5σ test, need to generate ~ 10^9 toy MC experiments

Inclusion of systematics

- For the inclusion of systematics, the most popular approaches are:
 - semi-bayesian approach (Highland-Cousins): the likelihood is integrated over the nuisance parameters before applying the desired statistical approach

e.g. likelihood for poisson-distributed yields, integrated over a gaussian uncertainty on the expected background *b*

$$q(n)_{s+b} = \frac{1}{\sqrt{2\pi\sigma_b}} \int_0^\infty p(n)_{s+b'} e^{-(b-b')^2/2\sigma_b^2} db'$$
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e.g. likelihood for poisson-distributed yields, integrated over gaussian uncertainties on the expected background b and signal efficiency ϵ

$$q(n)_{s+b} = \frac{1}{2\pi\sigma_b\sigma_\epsilon} \int_0^\infty \int_0^\infty p(n)_{b'+\epsilon's}$$
$$\times e^{-(b-b')^2/2\sigma_b^2} e^{-(1-\epsilon')^2/2\sigma_\epsilon^2} db' d\epsilon'$$

- For the inclusion of systematics, the most popular approaches are:
 - profile likelihood ratio: the likelihood is maximized with respect to the nuisance parameters when building the likelihood ratio:

 $\mathscr{L}(\mathbf{p}, \mathbf{q}) = P(\text{data} | \mathbf{p}, \mathbf{q})P(\mathbf{q})$

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 External constraint: PDF (e.g. gaussian)

$$\mathscr{L}(\mathbf{p},\mathbf{q}) = P(\text{data} | \mathbf{p},\mathbf{q}) P(\mathbf{q})$$

External constraint: PDF (e.g. gaussian) representing the uncertainty on the nuisance parameters

- For the inclusion of systematics, the most popular approaches are:
 - profile likelihood ratio: the likelihood is maximized with respect to the nuisance parameters when building the likelihood ratio:

 $\mathcal{L}(\mathbf{p}, \mathbf{q}) = P(\text{data} | \mathbf{p}, \mathbf{q})P(\mathbf{q})$

$$\mathscr{R}(\mathbf{p}) = \frac{\mathscr{L}(\mathbf{p}, \hat{\mathbf{q}})}{\mathscr{L}(\hat{\mathbf{p}}, \hat{\mathbf{q}})}$$

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Likelihood maximized over **q** for fixed **p**

$$\mathscr{R}(\mathbf{p}) = \frac{\mathscr{L}(\mathbf{p}, \hat{\mathbf{q}})}{\mathscr{L}(\hat{\mathbf{p}}, \hat{\mathbf{q}})}$$

Likelihood maximized over **p** and **q**

- For the inclusion of systematics, the most popular approaches are:
 - profile likelihood ratio: the likelihood is maximized with respect to the nuisance parameters when building the likelihood ratio:

 $\mathscr{L}(N_{\text{sig}}, \mathbf{p}) = P(\text{data} | N_{\text{sig}}, \mathbf{p})P(\mathbf{p})$

$$\mathcal{R} = \frac{\text{Poisson}(n_{\text{obs}}; n_{\text{sig}} + n_{\text{bkg}})P(n_{\text{bkg}})}{\text{Poisson}(n_{\text{obs}}; n_{\text{sig,best}} + n_{\text{bkg,best}})P(n_{\text{bkg,best}})}$$

- For the inclusion of systematics, the most popular approaches are:
 - profile likelihood ratio: the likelihood is maximized with respect to the nuisance parameters when building the likelihood ratio:

$$\begin{aligned} \mathscr{L}(N_{\text{sig}}, \mathbf{p}) &= P(\text{data} \mid N_{\text{sig}}, \mathbf{p}) P(\mathbf{p}) \\ \mathcal{R} &= \begin{cases} \frac{\mathcal{L}(N_{\text{sig}}, \hat{\mathbf{p}}(N_{sig}))}{\mathcal{L}(\hat{N}_{\text{sig}}, \hat{\mathbf{p}})} & \text{if } N_{\text{sig,best}} \ge 0 \\ \frac{\mathcal{L}(N_{\text{sig}}, \hat{\mathbf{p}}(N_{sig}))}{\mathcal{L}(0, \hat{\mathbf{p}}(0))} & \text{if } N_{\text{sig,best}} < 0 \end{cases} \end{aligned}$$

- For the inclusion of systematics, the most popular approaches are:
 - profile likelihood ratio: the likelihood is maximized with respect to the nuisance parameters when building the likelihood ratio:

Set of nuisance parameters

$$\begin{aligned} \mathscr{L}(N_{\mathrm{sig}},\mathbf{p}) &= P(\mathrm{data} \,|\, N_{\mathrm{sig}},\mathbf{p}) P(\mathbf{p}) \\ \mathcal{R} &= \begin{cases} \frac{\mathcal{L}(N_{\mathrm{sig}},\hat{\mathbf{p}}(N_{sig}))}{\mathcal{L}(N_{\mathrm{sig}},\hat{\mathbf{p}})} & \text{which maximizes the}\\ \mathrm{ikelihood for a given N_{\mathrm{sig}}} \\ \mathrm{if} N_{\mathrm{sig},\mathrm{best}} \geq 0 \\ \frac{\mathcal{L}(N_{\mathrm{sig}},\hat{\mathbf{p}}(N_{sig}))}{\mathcal{L}(0,\hat{\mathbf{p}}(0))} & \mathrm{if} N_{\mathrm{sig},\mathrm{best}} < 0 \end{cases} \end{aligned}$$





w/o systematics

w/ \pm 3 uncertainty on μ













- For the inclusion of systematics, the most popular approaches are:
 - simplified approach: in the toy MCs used to build the likelihood ratio distributions, nuisance parameters are randomly fluctuated in the generation or fit of the toy MC experiments (conceptually similar to the semi-bayesian approach)

- What values of the nuisance parameters should be used when generating the toy MC samples?
- Several options:
 - a priori estimate: fixed values decided a priori
 - can have significant under- or over-coverage

- What values of the nuisance parameters should be used when generating the toy MC samples?
- Several options:
 - conservative: generate with the values giving the worst upper limit
 - can have very large over-coverage

- What values of the nuisance parameters should be used when generating the toy MC samples?
- Several options:
 - **Highland-Cousins:** extract a random value of the nuisance parameters for each toy MC experiment, according to an a-priori distribution
 - can have some over-coverage when the nuisance parameter have a true single value (not varying from experiment to experiment)

- What values of the nuisance parameters should be used when generating the toy MC samples?
- Several options:
 - **a-posteriori Highland-Cousins:** extract a random value of the nuisance parameters for each toy MC experiment, according to an a-posteriori distribution derived from data
 - can still have some over-coverage, but less than the a-priori method

- We have seen that the generation of MC samples for extremely rare events can be problematic
- Nonetheless, sometimes the use of MC to extract PDFs is unavoidable
 - if the PDF shape is not known a priori, and the MC sample is too small to infer a reliable parameterization, there is a strong risk of overestimating or underestimating the systematic uncertainties, due to the inclusion of unnecessary shape uncertainties

True distribution

Exponential estimated from the MC





— True distribution

- Exponential estimated from the MC
 - Exponential estimated on data with constraint from the MC



MC

Data

- We have seen that the generation of MC samples for extremely rare events can be problematic
- Nonetheless, sometimes the use of MC to extract PDFs is unavoidable
 - if the PDF shape is not known a priori, and the MC sample is too small to infer a reliable parameterization, there is a strong risk of overestimating or underestimating the systematic uncertainties, due to the inclusion of unnecessary shape uncertainties
 - using MC histograms to represent the PDFs, with a proper treatment of uncertainties, could be the solution

The Beeston-Barlow approach

- Binned fit with different populations (e.g. signal, backgrounds)
- The expected MC content of each bin i for each population j (A_{ji}) is treated as a nuisance parameter, constrained from the actual MC

fi: expected yield in bin i of data

 p_j : data/MC scale factor for population j (to be estimated)

$$\ln \mathscr{L} = \sum_{i=1}^{n} d_i \ln f_i - f_i + \sum_{i=1}^{n} \sum_{j=1}^{m} a_{ji} \ln A_{ji} - A_{ji}$$

Likelihood for data (Poisson) d_i: observed yield in bin i of data

 $f_i = \sum_{j=0}^{n} p_j A_{ji}$

Likelihood for MC (Poisson)

a_{ij}: observed yield in bin i of MC for population j A_{ji}: expected yield in bin i of MC for population j (nuisance parameters)

The Beeston-Barlow approach



Practical Examples

Phys. Rev. Lett. 131, 161802 (2023)

Muon g-2 experiment at FNAL

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: measurement of the muon g-2
- Approach: measurement of the spin precession frequency of muons in orbit in a magnetic field



 Experimental observable: rate of positrons with E > E_{thr} emitted in forward direction w.r.t. the muon momentum

SYSTEMATICS

 Main criticality: accuracy of the magnetic field

$$a_{\mu} = \frac{\omega_a}{\tilde{\omega}'_p(T_r)} \frac{\mu'_p(T_r)}{\mu_e(H)} \frac{\mu_e(H)}{\mu_e} \frac{m_{\mu}}{m_e} \frac{g_e}{2}$$

Solution: instead of taking absolute field and frequency measurements, a **standard metrologic process**

(precession frequency of shielded protons in a spherical sample, ω_p) is measured in the same field, and the ratio is used

 Residual uncertainties from beam dynamics, temperature stability, external inputs

2021 JINST 16 P12041

OBJECTIVE AND E	Quantity	Correction (ppb)	Uncertainty (ppb)	
	ω_a^m (statistical)		201	
Objective: r	ω_a^m (systematic)		25	accuracy of the
Approach	C _e	451	32	
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10 ² Fermilab	$g_e/2$		0	quency of
$\frac{\mu}{q-2}$	Total systematic for \mathcal{R}'_{μ}		70	s in a spherical
$10^{-1} \overset{\text{L}}{\underset{0}{\overset{0}{\overset{0}{}}}} 10^{-2} \overset{\text{L}}{\overset{0}{}} 30 \overset{\text{L}}{} 40 \overset{\text{L}}{} 5$	Total external parameters		25	neasured in the
	Total for a_{μ}	622	215	the ratio is used
Experiment				ainties from
positrons with	$E > E_{thr}$ emitted in	forward	beam dynam	ics, temperature

direction w.r.t. the muon momentum

stability, external inputs

Electron g-2

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: measurement of the electron g-2
- Approach: measurement of energy transitions for a single electron in a magnetic field inside a Penning trap

$$n=1 \xrightarrow{r} \frac{1}{\bar{f_c}} = \frac{1}{\bar{v_c}} - \frac{3}{2}\delta$$

$$n=1 \xrightarrow{r} \frac{1}{\bar{v_a}} = \frac{1}{\bar{v_c}} - \frac{3}{2}\delta$$

$$n=0 \xrightarrow{r} \frac{1}{\bar{v_a}} = \frac{1}{\bar{v_c}} - \frac{1}{\bar{v_c}} = \frac{1}{\bar{v_c}} + \frac{1}{\bar{v_c}}$$

$$n=0 \xrightarrow{m_s=-1/2} m_s=+1/2$$

Experimental observable: quantum jumps hv, excited with electrodes,
induce currents in the electrodes
themselves, with a resonance if the frequency of the excitation matches v

SYSTEMATICS

•

- The necessary quantities ν_a, ν_c are measured in situ with the same approach—> no metrology issue
- Dominant uncertainty is expected from the correction between frequencies for free and trapped electrons
- Indeed, B field fluctuations are observed and induce additional systematics



Electron g-2



induce currents in the electrodes themselves, with a resonance if the frequency of the excitation matches ν



nEDM

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: search for a neutron EDM
- Approach: measurement of spin precession frequency in E + B field



$$f_n = \frac{1}{\pi\hbar} |\mu_n \vec{B_0} + d_n \vec{E}|$$

- Experimental technique: Ramsey spectroscopy of polarized ultra-cold neutrons within a shielded chamber:
 - π/2 spin flip by oscillating megnetic field
 - find the frequency fn that maximizes the asymmetry btw. spin up and down

SYSTEMATICS

- Main criticality: this is not a genuine measurement of zero, because µnB₀ has to be subtracted from a non-zero measurement:
 - absolute value of B₀ need to be known
 - comagnetometry against a metrologic standard (¹⁹⁹Hg)
- Several residual systematics:
 - systematics in the ¹⁹⁹Hg measurement
 - magnetic non-uniformities
 - asymmetries in the distribution of neutrons w.r.t. the magnetic field in the chamber

....

nEDM

OBJECTIVE AND E	XPERIMENTAL APPROACH		SYSTEMATICS				
 Objective: 		• Main	oritioality: th	his is not a			
 Approach precession 	Effect	Shift	Error	to be			
precession	Error on $\langle z \rangle$		7	non-zero			
Four-layer Mu-metal shield	Higher-order gradients \hat{G}	69	10				
Precession chamber	Transverse field correction $\langle B_T^2 \rangle$	0	5	\Rightarrow of B ₀ need to			
UV light source Mercury polarizing	Hg EDM [8]	-0.1	0.1				
UV light source	Local dipole fields		4	etry against a			
Spin flipper 1	$v \times E$ UCN net motion		2	Indard (¹⁹⁹ Hg)			
5T -magnet	Quadratic $v \times E$		0.1	vstematics:			
	Uncompensated G drift		7.5				
 Experiment 	Mercury light shift		0.4	i the ¹⁹⁹ Hg			
spectrosco	Inc. scattering ¹⁹⁹ Hg		7				
neutrons v	TOTAL	69	18	-uniformities			
- π/2 spi				n the distribution			
field				r.t. the magnetic			
IIEIU		fie	field in the chamber				
- find the							
the asyr							

muEDM

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: search for a muon EDM
- Approach: detection of non-zero spin precession in a magnetic field, with MDM precession canceled by a suitable combination of E and B fields



$$\vec{d} = \frac{\eta e}{2mc}\vec{s}$$
$$= \frac{\eta q}{\beta}\vec{s} \times \vec{q}$$

10 0

$$(t) \propto \frac{2P_0 E_f \alpha |d_\mu|}{a\hbar\gamma^2} t$$

Experimental observable: timedependent asymmetry of positrons emitted along and opposite to the B field

SYSTEMATICS

- Main criticality:
 - it is not necessary to know extremely well the main components of the field, but fake EDM can arise from fringe fields
- EDM is CP-violating, standard electrodynamics is CP-conserving:
 - systematics can be canceled by inverting B and injection direction
 - indeed, it moves the systematics from electrodynamics to the symmetry between the two injection modes
- Detector asymmetries to be kept under control

MEG & MEG II



MEG & MEG II



KOTO

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: search for $K_L^0 \to \pi^0 \nu \overline{\nu}$
- Approach: search for kaon decays with only 2 photons and nothing else



 Experimental technique: kaons decaying in a volume surrounded by hermetic neutral and charged particle detectors, used to veto background decays

SYSTEMATICS

- Main criticality: hermeticity and PID to be precisely controlled to get rid of the Standard Model and beam-halo backgrounds up 10⁷ rejection factor:
 - dominant systematics from the expected background rates


KOTO



KATRIN

OBJECTIVE AND EXPERIMENTAL APPROACH

- Objective: measurement of the electron anti-neutrino mass
- Approach: measurement of the beta decay end-point



 Experimental technique: electromagnetic filter to count events above a certain energy threshold

SYSTEMATICS

- Dominant systematics:
 - electric and magnetic field (accuracy & stability)
 - non-Poisson background
- Origin of non-Poisson background:
 - nuclear decays from contaminants produce keV electrons
 - they ionize the residual gas, producing secondaries
 - many secondaries from a very small number of primaries (i.e. correlated background)—> non-Poisson fluctuations

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OBJECTIVE AND E

- Objective: r anti-neutrin
- Approach: decay end-



Effect	68.2% CL uncertainty on m_{ν}^2 (eV ²)
Statistical	0.29
Non-Poissonian background	0.11
Source-potential variations	0.09
Scan-step-duration-dependent background	0.07
qU-dependent background	0.06
Magnetic fields	0.04
Molecular final-state distribution	0.02
Column density ${\bf x}$ inelastic scattering cross section	0.01
Activity fluctuations	0.01
Energy-loss function	<0.01
Detector efficiency	<0.01
Theoretical corrections	<0.01
High-voltage stability and reproducibility	<0.01
	0.24

ematics: magnetic field stability) background oisson ays from its produce ٦S the residual cing 5 ndaries from a number of

 Experimental technique: electromagnetic filter to count events above a certain energy threshold primaries (i.e. correlated background)—> non-Poisson fluctuations

Conclusions

- Compared to general-purpose experiments, systematics in single-purpose experiments pose some special challenges:
 - metrology issues
 - need of dedicated (hardware) tools for the control of systematics
- Control of systematic uncertainties is a critical aspect in the design of single-purpose experiments, often requiring special expertise from outside the HEP field