

New test of Lorentz invariance using the MICROSCOPE space mission

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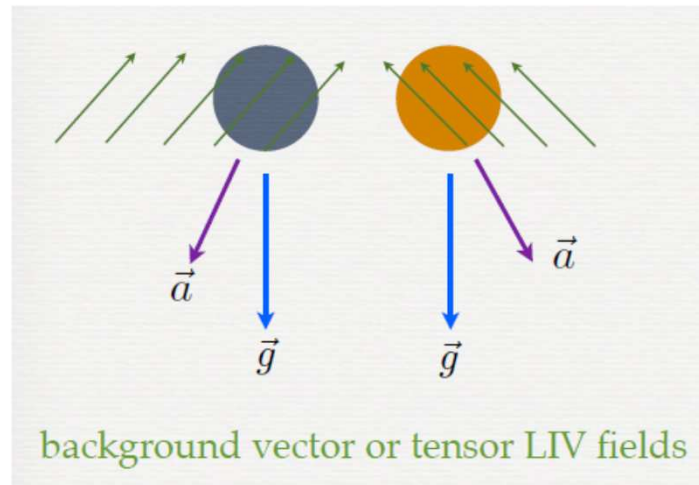
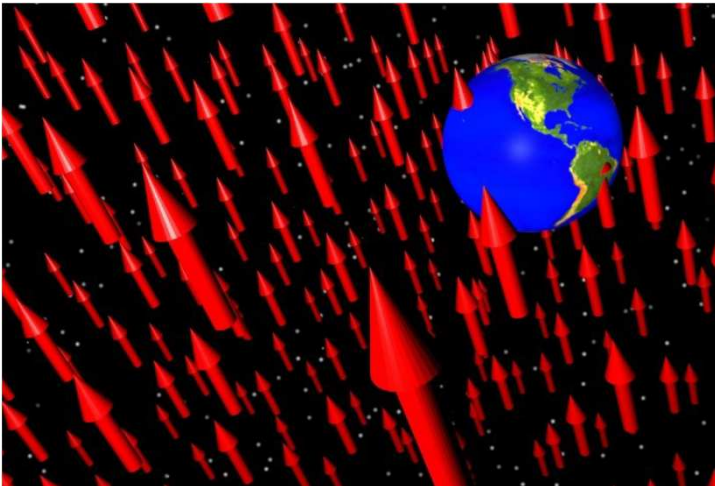
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Menu

- Introduction
- The Lorentz Violating SME
- The SME signal in the Microscope data
- Analysis Method
- Data
- Comparison to independent analysis
- Final Results
- Conclusion

Introduction

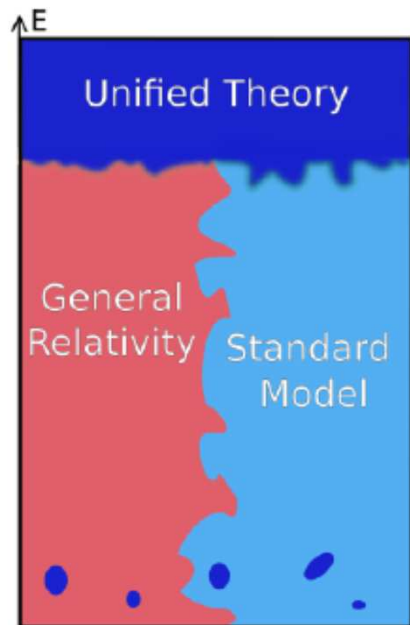
- Analysis of the MICROSCOPE data have so far been restricted to “standard WEP” models, i.e. a possible differential acceleration of test masses along the local gradient of the Newtonian potential \vec{g} .
- The Lorentz violating Standard Model Extension (SME) developed since the late 1990s by Kostelecky and co-workers leads to a different, richer, phenomenology.
- It allows for additional modulations of the signal, related to background fields in an inertial frame (SCF)



The Lorentz Violating SME

What is SME?

- Effective field theory built from SM fields & in curved spacetime
- General framework describing low-energy effects of a spontaneous LV occurring at Planck scale



Main features

- Includes all possible observer-independent LV built from SM fields and background coefficients in the Lagrangian
- These coefficients vanish if the symmetry is preserved
- LV terms are expected to be strongly suppressed compared with non violating terms
- Phenomenological, not quantitatively predictive
- Enables the derivation of experimental observables

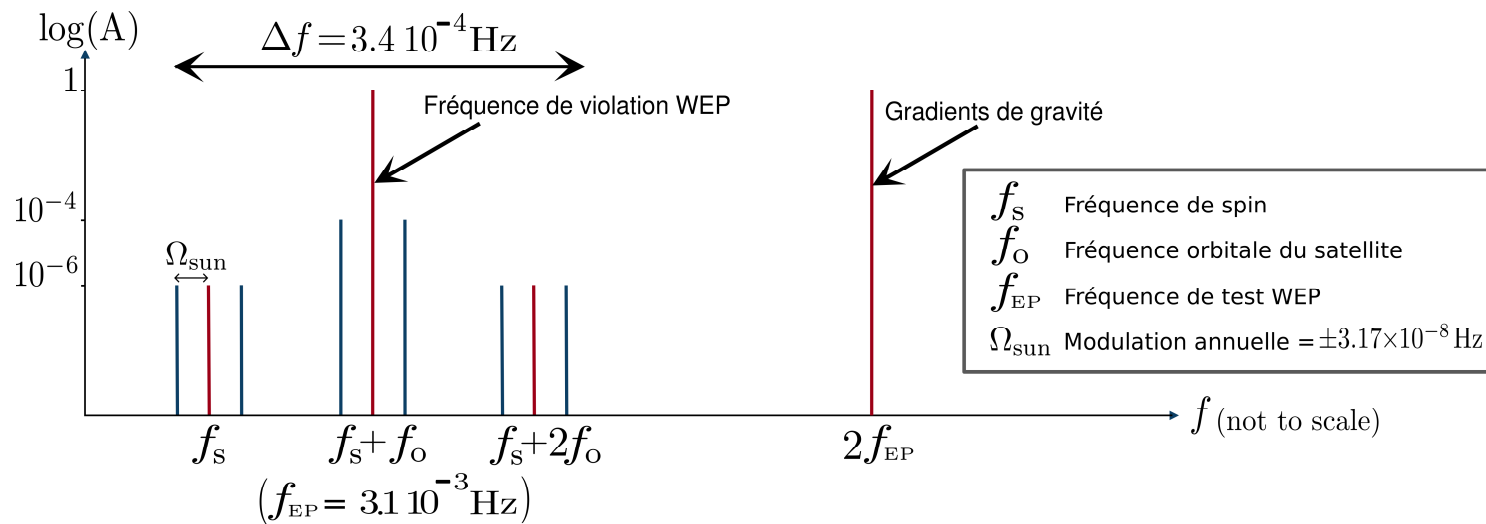
[Kostelecky *et al.*, PRD 51, 1995], [Kostelecky *et al.*, PRD 58, 1998]



The SME signal in the MICROSCOPE data

$$S^B \simeq \int d\lambda (-m^B c \sqrt{-g_{\mu\nu} u^\mu u^\nu} - (a_{\text{eff}}^B)_\mu u^\mu / c)$$

SME-coefficient: 4-vector



The SME signal in the MICROSCOPE data

$$\begin{aligned} \gamma_{\hat{x}} = & b + S_{\hat{x}\hat{x}}\Delta_{\hat{x}} + (S_{\hat{x}\hat{y}} + \dot{\Omega}_z)\Delta_{\hat{y}} + (S_{\hat{x}\hat{z}} - \dot{\Omega}_y)\Delta_{\hat{z}} + 2g_{\hat{x}} [\alpha(\bar{a}_{\text{eff}}^{(d)})_T + \beta_X\alpha(\bar{a}_{\text{eff}}^{(d)})_X + \beta_Y\alpha(\bar{a}_{\text{eff}}^{(d)})_Y + \beta_Z\alpha(\bar{a}_{\text{eff}}^{(d)})_Z] \\ & - \frac{6GM_{\oplus}R_{\oplus}^2}{5cr^5} (R_{\hat{x}\hat{x}}\tilde{x}^{\text{orb}} + R_{\hat{x}\hat{y}}\tilde{y}^{\text{orb}} + R_{\hat{x}\hat{z}}\tilde{z}^{\text{orb}}) [\tilde{x}^{\text{orb}}\alpha(\bar{a}_{\text{eff}}^{(d)})_Y - \tilde{y}^{\text{orb}}\alpha(\bar{a}_{\text{eff}}^{(d)})_X]\omega_{\tilde{z}} \\ & + \frac{2GM_{\oplus}R_{\oplus}^2}{5cr^3} [\alpha(\bar{a}_{\text{eff}}^{(d)})_Y R_{\hat{x}\hat{x}} - \alpha(\bar{a}_{\text{eff}}^{(d)})_X R_{\hat{x}\hat{y}}]\omega_{\tilde{z}}, \end{aligned}$$

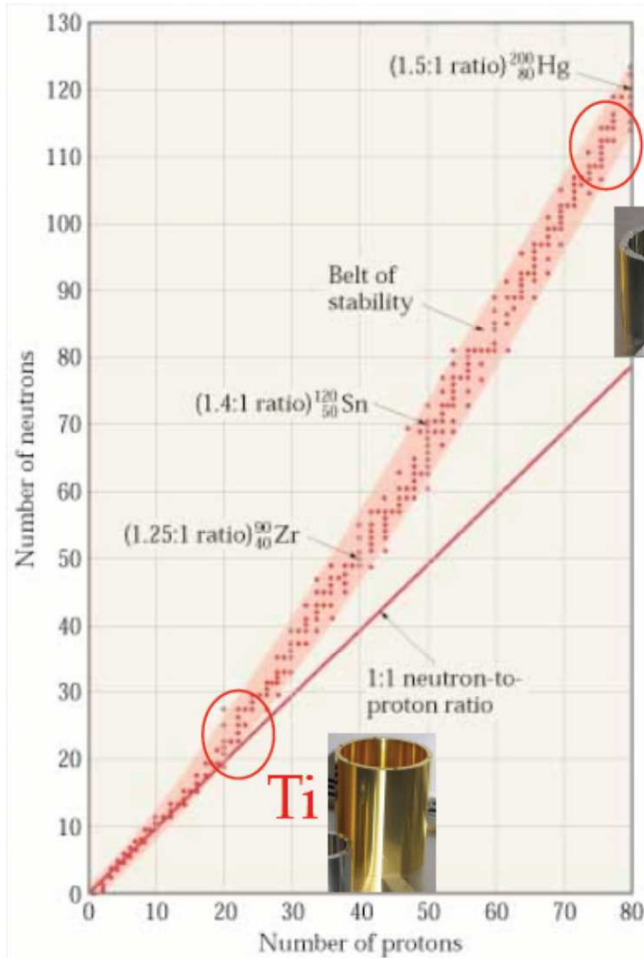
Mission data

SME-coeff. + technical parameters

- Similar expressions for acceleration along y and z in the instrument frame.
- Derived from [Kostelecky & Tasson, 2011]. Cross checked independently by J. Tasson and Q. Bailey.
- Need to determine the 4 components of the differential SME vector coefficient $\alpha(\bar{a}_{\text{eff}}^{(d)})$.

$$(\bar{a}_{\text{eff}}^{(d)})_{\mu} = \frac{(\bar{a}_{\text{eff}}^{\text{Pt}})_{\mu}}{m^{\text{Pt}}c^2} - \frac{(\bar{a}_{\text{eff}}^{\text{Ti}})_{\mu}}{m^{\text{Ti}}c^2} = \sum_{w=n,p,e} \left(\frac{N_w^{\text{Pt}}}{m^{\text{Pt}}c^2} - \frac{N_w^{\text{Ti}}}{m^{\text{Ti}}c^2} \right) (\bar{a}_{\text{eff}}^w)_{\mu}$$

The SME signal in the MICROSCOPE data



$$\alpha(\bar{a}_{\text{eff}}^{(d)})_{\mu} = A \cdot \alpha(\bar{a}_{\text{eff}})_{\mu}^n + C \cdot \alpha(\bar{a}_{\text{eff}})_{\mu}^{(e+p)}$$

$$A \approx \Delta \left(\frac{N_n}{N_n + N_p} \right) \text{ GeV}^{-1} \quad C \approx \Delta \left(\frac{N_p}{N_n + N_p} \right) \text{ GeV}^{-1} \approx -A$$

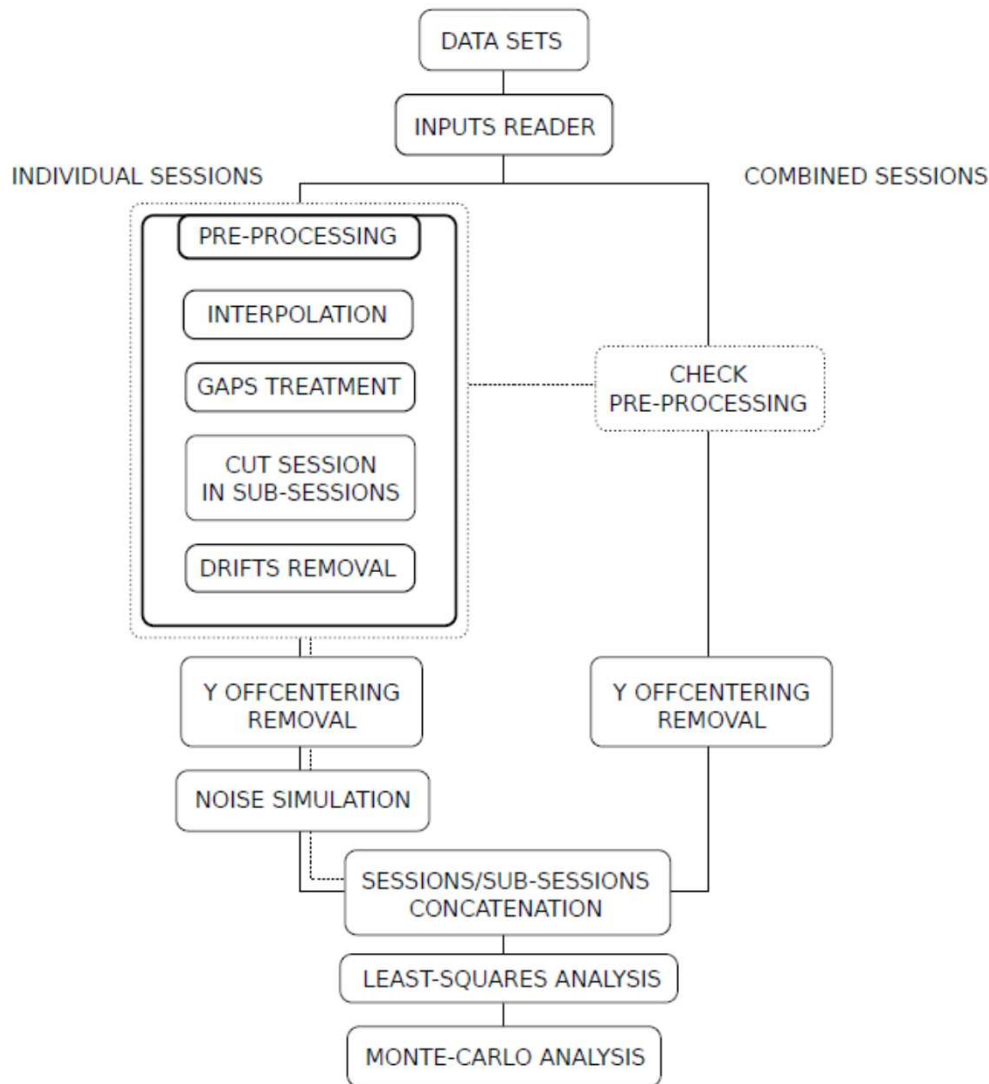
neutron ratio

Platinum	0.6
Titanium	0.54
Difference	0.06

Need to take isotopic composition and alloys into account: 9Pt:1Rh and 9Ti:0.6Al:0.4Va).

$$\rightarrow A \simeq 0.06 \text{ GeV}^{-1}, C \simeq -0.06 \text{ GeV}^{-1}$$

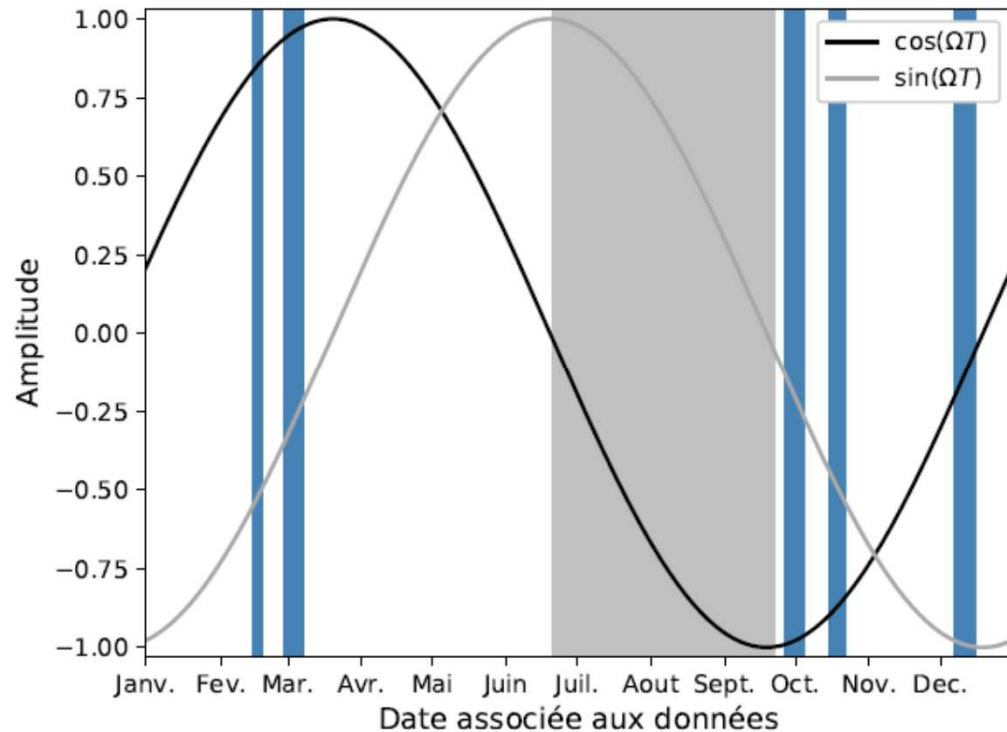
Analysis Method



- We use a MCLS method, already successfully applied to Galileo data, and simulated ACES data.
- Determine parameters from an ordinary LS fit to data.
- Determine their uncertainties and covariances by generating N synthetic data sets with same noise properties, gaps, non-stationarity, etc... and do statistics on the N sets of fitted parameters.
- Apply exactly the same analysis to temperature data, in order to obtain systematic effect on each parameter (using sensitivity coefficient given in Touboul et al. 2017).

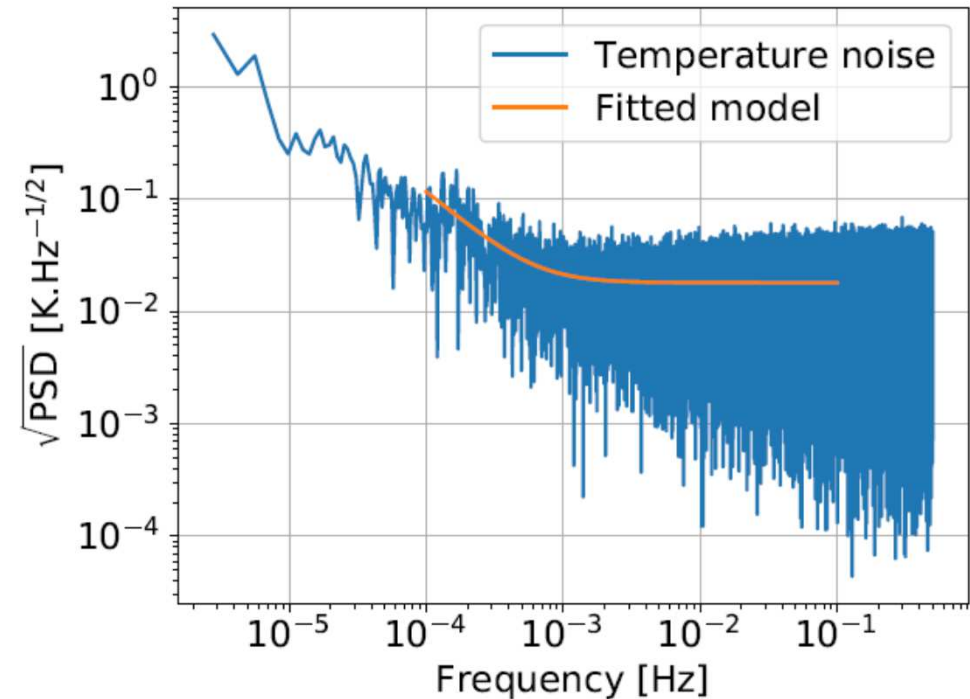
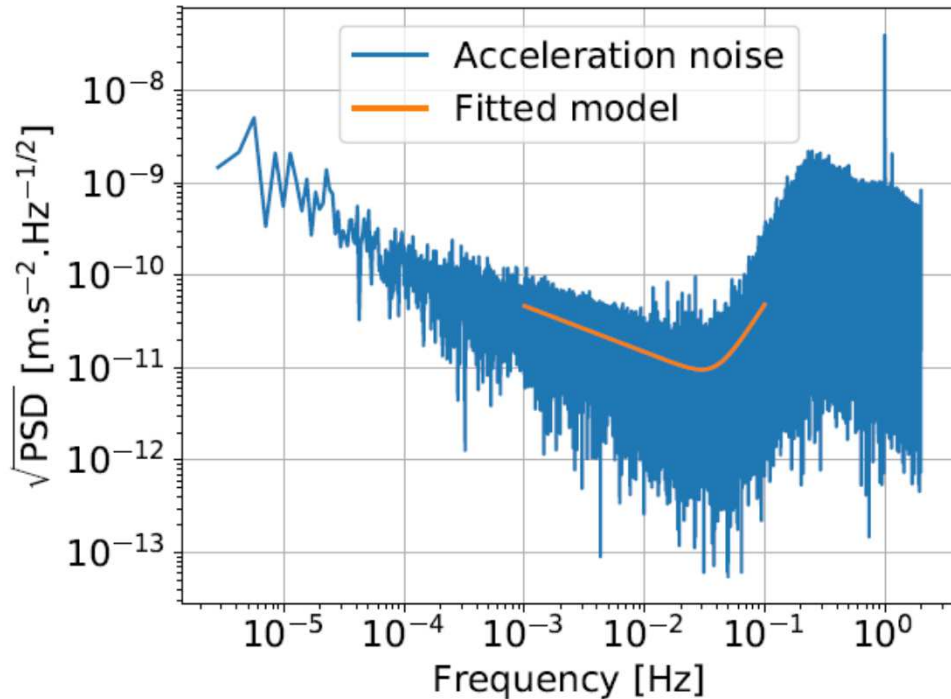
Data

n° session	Durée en nombre d'orbites	Date de début et de fin	Pourcentage de données manquantes
210	50	2017/02/14 - 2017/02/18	7.6×10^{-4}
218	119	2017/02/28 - 2017/03/08	2.8×10^{-4}
326	101	2017/09/27 - 2017/10/04	3.7×10^{-4}
358	91	2017/10/14 - 2017/10/21	5.2×10^{-4}
404	119	2017/12/07 - 2017/12/15	4.9×10^{-4}



- The data sets cover most of 2017. Allows resolution of annual modulations
- Includes differential acceleration, temperature and all auxiliary data (orbit, attitude, gravity).
- In the following session 404 will be used for illustrating the procedure.

Data



- Data gaps are extremely rare ($<0.0008\%$) and all = 0.25 s (one missing point), identified by a (0,1) mask. Removing them can only affect PSD noise estimation, not the fit of the parameters. They are highly unlikely to have a significant effect on the MCLS result.
- We first remove long term drifts by fitting a polynomial of order 5. We checked that fitting it at the same time with all other parameters makes no difference.

Comparison to independent analysis

- Session 218, same as the one analyzed by OCA/ONERA in Touboul et al., PRL 2017.
- Use a standard WEP model to compare results:

Parameter	Value and uncertainties	Units	PRL 2017
δ	$(4.0 \pm 9.6_{\text{stat}} \pm 13.0_{\text{syst}}) \times 10^{-15}$	–	$(-1 \pm 9_{\text{stat}} \pm 9_{\text{syst}}) \times 10^{-15}$
Δ_x	$(20.2 \pm 0.04) \times 10^{-6}$	<i>m</i>	$(20.1 \pm 0.1) \times 10^{-6}$
Δ_z	$(-5.77 \pm 0.04) \times 10^{-6}$	<i>m</i>	$(-5.6 \pm 0.1) \times 10^{-6}$

- Only statistical uncertainties shown for the offcenterings.
- All correlation coefficients < 0.08.
- Larger systematic uncertainty might be due to OCA/ONERA having used more temperature data than just session 218 [M. Rodrigues, personal communication].

Two very different and completely independent analysis methods give same result ! 😊

Final Results

Coefficient	Value and uncertainties [GeV]
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_T$	$(6.3 \pm 12) (10)(6.0) \times 10^{-14}$
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_X$	$(0.81 \pm 1.7) (1.4)(0.98) \times 10^{-9}$
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Y$	$(0.67 \pm 3.1) (1.4)(2.7) \times 10^{-7}$
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Z$	$(-1.55 \pm 7.1) (3.2)(6.3) \times 10^{-7}$

- Combined analysis using all five sessions, weighted according to their individual uncertainties on WEP - δ
- Uncertainties at 68% confidence
- Correlation coefficients are ≈ 0.9 between T and X components, 1 between Y and Z, and ≤ 0.02 otherwise.

As we have the full covariance matrix we can determine independent linear combinations:

SME linear combination	Value and uncertainty [GeV]
a_1	$(1.7 \pm 5.5) \times 10^{-14}$
a_2	$(0.85 \pm 1.7) \times 10^{-9}$
a_3	$(0.33 \pm 1.2) \times 10^{-9}$
a_4	$(-1.7 \pm 7.7) \times 10^{-7}$

	$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_T$	$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_X$	$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Y$	$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Z$
a_1	1.0	-6.0×10^{-5}	4.8×10^{-6}	2.0×10^{-6}
a_2	5.9×10^{-5}	0.99	0.11	0.050
a_3	-1.3×10^{-5}	-0.12	0.91	0.39
a_4	1.2×10^{-9}	-4.9×10^{-5}	-0.40	0.92

Final Results

Following SME data tables [Kostelecky and Russel] we give “maximal sensitivities”, by assuming only one non-zero coefficient at a time rounding logarithmically the 2σ uncertainty.

Coefficient	Maximal sensitivity [GeV]	Prev. best [GeV]
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_T$	10^{-13}	10^{-11}
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_X$	10^{-8}	10^{-8}
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Y$	10^{-9}	10^{-8}
$\alpha(\bar{a}_{\text{eff}}^{(n-e-p)})_Z$	10^{-8}	10^{-7}

Torsion balance – reinterpretation [1,2]

Superconducting gravimeters [3], Lunar Laser Ranging [4], Binary Pulsars [5]

Improvements by 1-2 orders of magnitude !

- [1] Schlamminger et al., PRL 2008
- [2] Kostelecky & Tasson, PRD 2011
- [3] Flowers et al., PRL 2017
- [4] Bourgoin et al., PRL 2017
- [5] Shao, Symmetry 2019

Conclusion and Outlook

- First SME analysis of Microscope data, improving on previous constraints by 1-2 orders of magnitude.
- 5 data sets spread over a year, more to come \Rightarrow expect improvement in the near future.
- Our analysis confirms independently previous results (OCA/ONERA) when using the same data and model (“standard” WEP).
- We are ready to analyze new data and compare to others.
- Our software is “versatile” i.e. any model can be easily tested
 \Rightarrow give us your favorite model, we will test it!