Doppler shift lifetime measurements using the TIGRESS Integrated Plunger at ISAC-II/TRIUMF

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Selected goals of nuclear science research



- Understand the mechanisms of shell evolution in medium-mass and heavy nuclei as a function of isospin.
- Develop a theoretical framework that is able to make accurate predictions of nuclear properties.

TRIUMF 5 Year Plan 2015-2020.

Studying nuclear structure using the electromagnetic force

- The electromagnetic force provides a convenient non-intrusive probe of nuclear systems bound by the strong force.
- Lifetime measurements using gamma-ray spectroscopy provide:
 - 1. An observable sensitive to nuclear structure.
 - 2. A useful benchmark for nuclear model calculations.



The Recoil Distance Method with TIP and TIGRESS



The TIGRESS Integrated Plunger (TIP) device



P. Voss et al. Nucl. Inst. and Meth. A 746 (2014) 87, P. Voss et al. Phys. Proc. 66 (2015) 524.

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TIP with the CsI(TI) wall



TIP commissioning experiment details

- Measurement of the $2^+_1 \rightarrow 0^+_1$ lifetime in $^{84}{\rm Kr.}$
- Previous Coulex experiment reports $au=5.84\pm0.18$ ps.
- TIP and 11 TIGRESS detectors in a 3/5/3 configuration with 24-element CsI(TI) wall for particle identification (PID).
- Excited state populated via partially unsafe Coulex reaction (beam energy of 250 MeV or 2.979 MeV/u).
- A total of 13 target/degrader separation distances from 20–400 μm were analyzed.
- Data analysis via a comparison to Geant4-simulated lineshapes developed for low-statistics experiment analysis.

T. J. Mertzimekis et al. Phys. Rev. C 64 (2001) 024314.

Geant4 simulation framework



- Coulomb excitation followed by gamma-ray decay.
- Analytic solutions for single step *E*2 process (Coulex kinematics, angular distributions, etc.) with track weighting to handle thick target integration.
- Gamma-ray sensitive detectors ported from GRIFFIN/TIGRESS code originating from Guelph.

Adrich et al. Nucl. Inst. and Meth. A 598 (2009) 454, Alder et al. Rev. Mod. Phys. 28 (1956) 432.

Geant4-facilitated data analysis: Doppler-shift factors



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RDM using TIP at ISACII

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Histogram analysis using the likelihood ratio χ^2

• For a Poisson likelihood function, the likelihood ratio χ^2 is given by

$$\chi^{2} = 2 \sum_{i=1}^{k} y_{i} - n_{i} + n_{i} \ln(n_{i}/y_{i})$$

where y_i is the model and n_i the observed data.

- Pros:
 - Versatile (goodness of fit, point estimation, error analysis).
 - Control over parent distribution.
 - Minimizing likelihood ratio $\chi^2 \equiv \max$ maximizing likelihood function.
 - No variance estimation!
- Cons:
 - Non-linear.
 - . . .

Baker and Cousins. Nucl. Inst. and Meth. A 221 (1984) 437.

Data analysis procedure

- For a given input lifetime *τ*, simulate gamma-ray spectra grouped by the Doppler-shift factor at all distances.
- Model data using

$$y_i = \alpha_0 s_i + \alpha_1 + \alpha_2 \operatorname{erfc}\left(\frac{i-c}{w\sqrt{2}}\right),$$

where s_i is the simulated data and the α 's are free parameters.

- Minimize $\chi^2_{d,g}$ for each distance and group.
- Minimum in total $\chi^2 = \sum_{\text{dist. gr.}} \chi^2_{d,g}$ corresponds to best fit lifetime $\tau_{\text{min.}}$

Best fit lifetime from χ^2 analysis



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Comparison of fitting methods

1. Likelihood ratio χ^2

$$\chi_{\Lambda}^2 = 2 \sum_{i=1}^k y_i - n_i + n_i \ln(n_i/y_i)$$

2. Neyman χ^2

$$\chi_N^2 = \sum_{i=1}^k \frac{(n_i - y_i)^2}{n_i}.$$

- Probably the most common fitting statistic (ROOT default for histogram fits, for example).
- Uses data to estimate variance.
- Assumes bin error can be estimated by $\sqrt{n_i}$: what if $n_i = 0$?

Lineshape fits for $^{84}\text{Kr},\,60~\mu\text{m}:$ all data



Lineshape fits for 84 Kr, 60 μ m: 1% of data



Simulated lineshapes: groups 1 and 5



Best fit lifetime: 5.880 ± 0.013 (stat.) ± 0.070 (sys.) ps Literature value: 5.84 ± 0.18 ps [Mertzimekis 2001]

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Commissioning experiment summary

- Systematic uncertainties from the following 3 sources were identified:
 - 1. Transitions from higher-lying (feeding) states,
 - 2. Misalignment of the target and degrader foils,
 - 3. Choice of fit range for the χ^2 analysis.
- No deorientation effect observed in the data.
- Final reported lifetime: $au = 5.880 \pm 0.008$ (stat.) ± 0.070 (sys.) ps.
- + Excellent agreement with literature value of 5.84 ± 0.18 ps with factor of ${\sim}2$ reduction in uncertainty.
- A robust and flexible framework has been developed for the planning and analysis of RDM experiments using TIP.
- Paper for submission to Nucl. Inst. and Meth. A in preparation.
- T. J. Mertzimekis et al. Phys. Rev. C 64 (2001) 024314.

TIP RIB experiment motivation



TIP RIB experiment details

- Performed December 9-14, 2015.
- Measurement of the $\mathbf{2}_1^+ \rightarrow \mathbf{0}_1^+$ lifetime in $^{94}Sr.$
- Previous fast timing experiment reports 10 \pm 4 ps.
- TIP and 16 TIGRESS detectors with 24-element CsI(TI) wall.
- Running conditions same as commissioning experiment.
- Except the beam rate: factor of $\sim 10^4$ lower for RIB.
- Data was recorded at three target/degrader separation distances: 50, 100, 150 $\mu m.$

Mach et al. Nuc. Phys. A 523 (1991) 197.

Particle identification with RIBs



(a) Electrons from β^- decay (b) Al recoils from Coulex reaction

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Best fit lifetime from χ^2 analysis



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Simulated lineshapes: groups 1 and 5



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$B(E2:2_1^+ \rightarrow 0_1^+)$ measurement



Impact of high-precision $B(E2:2_1^+ \rightarrow 0_1^+)$ measurement



Summary of results

Current work

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 \begin{array}{lll} \mbox{Lifetime } \tau & 7.80^{+0.5}_{-0.4} \mbox{ (stat.) } \pm 0.07 \mbox{ (sys.) ps} \\ B(E2;2^+_1 \rightarrow 0^+_1) & 0.0254^{+0.0015}_{-0.0014} \mbox{ (stat.) } \pm 0.0002 \mbox{ (sys.) } e^2 b^2 \end{array}
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Mach (1991)

Lifetime au 10 ± 4 ps $B(E2; 2_1^+ \to 0_1^+)$ 0.020 ± 0.008 e²b²

- A robust and flexible framework has been developed for the planning and analysis of RDM experiments using TIP.
- Details of simulation framework, data analysis, and results for the ⁸⁴Kr commissioning experiment for submission to Nucl. Inst. and Meth. A in preparation.
- ⁹⁴Sr lifetime measurement paper submitted to Phys. Rev. C.
- T. J. Mertzimekis et al. Phys. Rev. C 64 (2001) 024314.

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Acknowledgments

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Theory references

- Kumar 1985: Quadrupole plus pairing Hamiltonian.
- Skalski 1997: Nilsson-Strutinsky method with the Woods-Saxon potential.
- Lalazissis 1995: Relativistic mean field.
- Buchinger 1994 (a): Finite range liquid drop.
- Buchinger 1994 (b): Strutinsky energy theorm.

A. Kumar and M. R. Gunye, Phys. Rev. C 32, 2116 (1985). J. Skalski, S. Mizutori, and W. Nazarewicz, Nuclear Physics A 617, 282 (1997). G. Lalazissis and M. Sharma. Nuclear Physics A 586, 201 (1995).

F. Buchinger et al., Phys. Rev. C 14 49, 1402 (1994).

TIP commissioning experiment details

⁸⁴ Kr properties	
E_{γ}	881.615 keV
$ au_{lit.}$	$5.84\pm0.18~\text{ps}$

Plunger setup.				
	Material	Thickness [mg/cm ²]	Thickness [µm]	
Target	Al	1.07 ± 0.04	3.96 ± 0.16	
Degrader	Cu	3.90 ± 0.16	4.35 ± 0.18	

Beam properties		
Beam energy	250 MeV	
Safe Coulex	200 MeV	
Rate	$\sim 2 imes 10^8$ pps	

TIP RIB experiment details

⁹⁴ Sr properties	
Eγ	836.9 keV
$ au_{lit.}$	$10\pm4~{ m ps}$

Plunger setup.			
	Material	Thickness [mg/cm ²]	Thickness [µm]
Target	Al	1.09 ± 0.04	4.05 ± 0.17
Degrader	Cu	3.69 ± 0.15	4.12 ± 0.17

_ .

Beam properties		
Beam energy	280 MeV	
Safe Coulex	227 MeV	
Rate	$\sim 25 imes 10^4$ pps	

Thick target integration with Geant4



Feeding analysis



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Deorientation effect



Analysis of low-statistics data sets

Challenges associated with the comparison of simulated data sets from Geant4 to low statistics data:

- Typical least squares analysis requires an estimate of the variance either from the model (Pearson χ^2) or from the data (Neyman χ^2).
- These are derived under the implicit assumption of a Gaussian error distribution.
- With few counts, errors are poorly estimated.
- Even worse, if there are 0 counts in the data, the very commonly used Neyman χ^2 statistic is undefined!
- Inconsistent normalization (which depends on choice of statistic!) without inclusion of explicit normalization parameter.

Basic quantities for histogram analysis

Consider a histogram with k bins labeled by the index i running from 1 to k and a model with J parameters labeled by index j. Define the following quantities:

- n_i = number of events in bin *i*.
- $\mathbf{n} = (n_1, n_2, \dots, n_k).$
- $\boldsymbol{\alpha} = (\alpha_1, \alpha_2, \dots, \alpha_j).$
- y_i = number of events predicted by the model (via α) in bin *i*.

•
$$\mathbf{y} = (y_1, y_2, \dots, y_k).$$

Baker and Cousins. Nucl. Inst. and Meth. A 221 (1984) 437.

Likelihood ratio χ^2 derivation

Assume the data in each bin are independently Poisson distributed. Then the likelihood function \mathcal{L} is given by

$$\mathcal{L}(\mathbf{y};\mathbf{n}) = \prod_{i=1}^{k} \frac{y_i^{n_i} \exp(-y_i)}{n_i!}.$$

Let **m** be the true (unknown) values of **n**. The likelihood ratio Λ is

$$\Lambda = \frac{\mathcal{L}(\mathbf{y};\mathbf{n})}{\mathcal{L}(\mathbf{m};\mathbf{n})},$$

and the likelihood ratio test theorem says that the "likelihood chi-square"

$$\chi^2_{\Lambda} = -2 \ln \Lambda = -2 \ln \mathcal{L}(\mathbf{y}; \mathbf{n}) + 2 \ln \mathcal{L}(\mathbf{m}; \mathbf{n})$$

asymptotically obeys a χ^2 distribution.

Baker and Cousins. Nucl. Inst. and Meth. A 221 (1984) 437, Wilks. Ann. Math. Stat. 9 (1938).

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Likelihood ratio χ^2 derivation

Replace the unknown \mathbf{m} with its bin-by-bin maximum likelihood estimator \mathbf{n} . A bit of algebra yields

$$\chi_{\Lambda}^{2} = -2 \ln \mathcal{L}(\mathbf{y}; \mathbf{n}) + 2 \ln \mathcal{L}(\mathbf{m}; \mathbf{n})$$
$$= -2 \ln \mathcal{L}(\mathbf{y}; \mathbf{n}) + 2 \ln \mathcal{L}(\mathbf{n}; \mathbf{n})$$
$$= 2 \sum_{i=1}^{k} y_{i} - n_{i} + n_{i} \ln(n_{i}/y_{i}).$$

Some attractive features:

- It's a χ^2 statistic (familiarity, versatility, error analysis).
- No variance estimation.
- Self-normalizing.
- A clear way to handle bins with $n_i = 0$.
- Minimizing χ^2_{Λ} is equivalent to maximizing $\mathcal{L}(\mathbf{y}; \mathbf{n})$.

Baker and Cousins. Nucl. Inst. and Meth. A 221 (1984) 437.

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