



Institute of High Energy Physics
Chinese Academy of Sciences

Discovering Hybrid Mesons

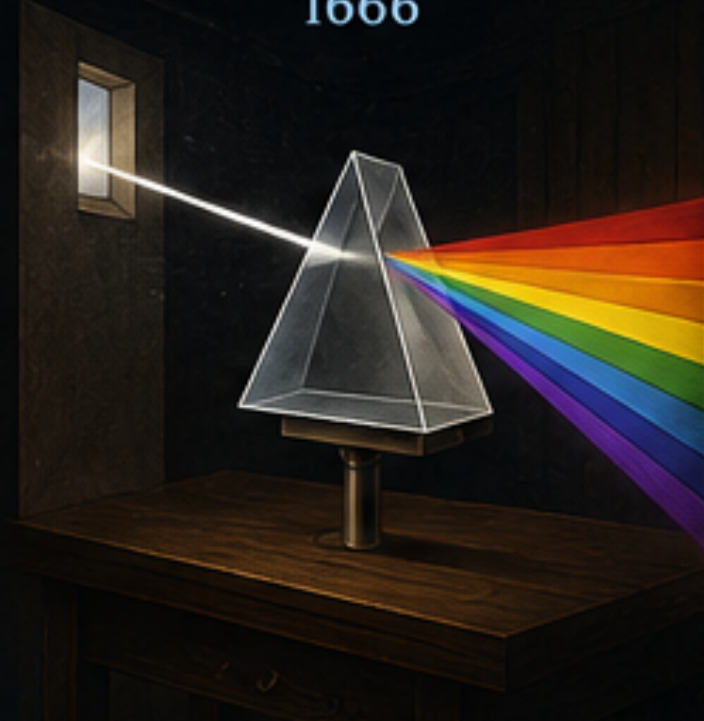


SPECTROSCOPY: LIGHT REVEALING THE UNIVERSE

From light to knowledge: spectroscopy has unlocked the structure of atoms, tested the foundations of quantum theory, and revealed a new world of particles.

1. NEWTON'S PRISM EXPERIMENT

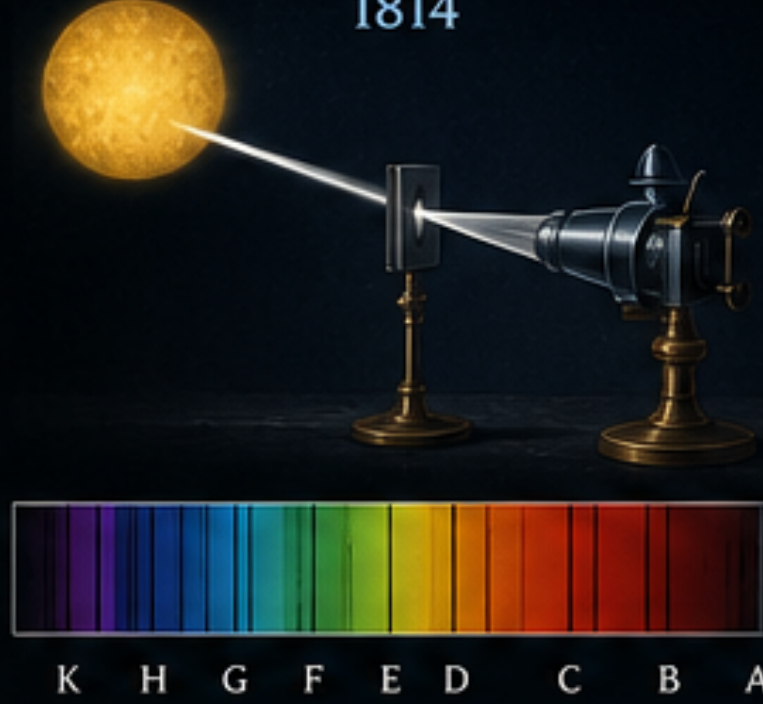
1666



Newton showed that white light is composed of colors. This was the birth of optical spectroscopy.

2. FRAUNHOFER'S SOLAR ANALYSIS

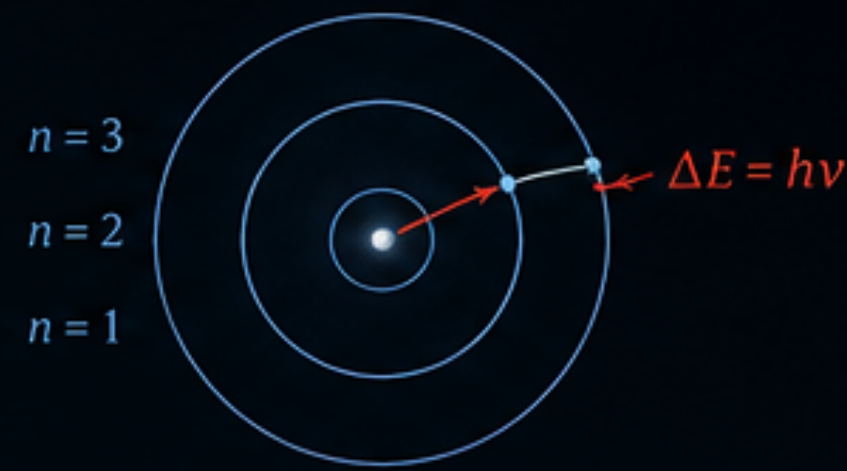
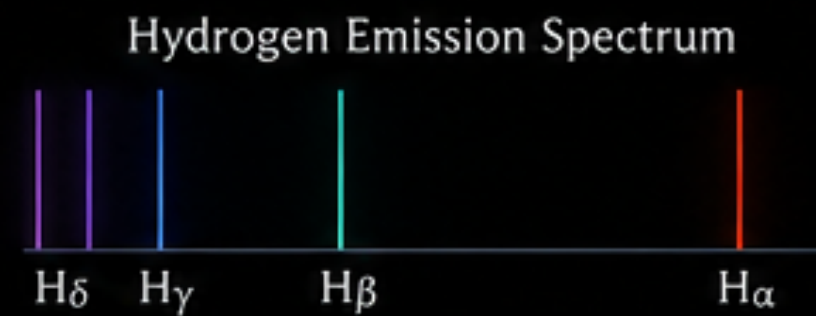
1814



Fraunhofer discovered dark absorption lines in the solar spectrum. These lines revealed that atoms leave unique fingerprints in light.

3. HYDROGEN SPECTRUM & BOHR MODEL

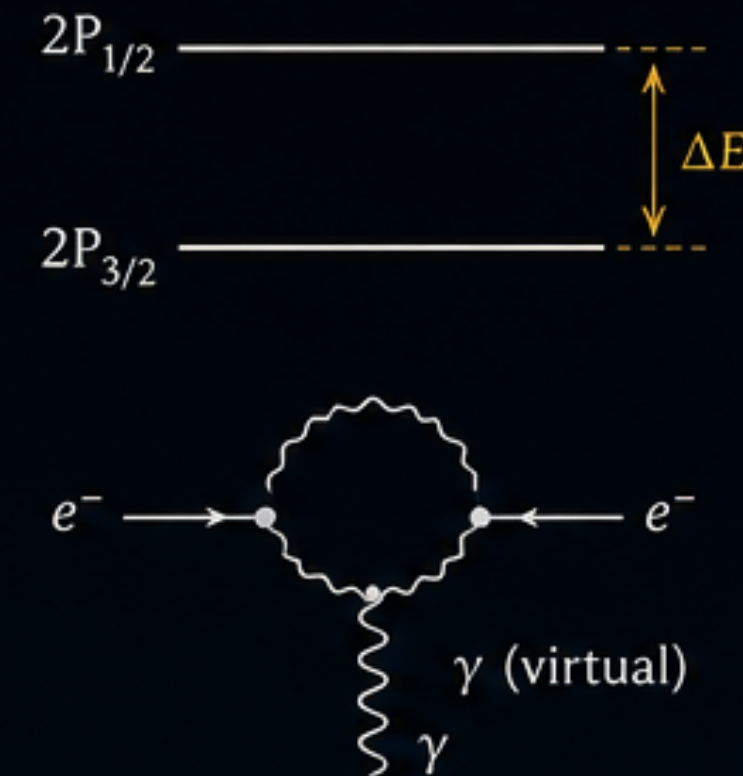
1885–1913



The discrete hydrogen spectrum led Bohr to propose quantized energy levels in atoms—launching modern atomic physics.

4. LAMB SHIFT & QED

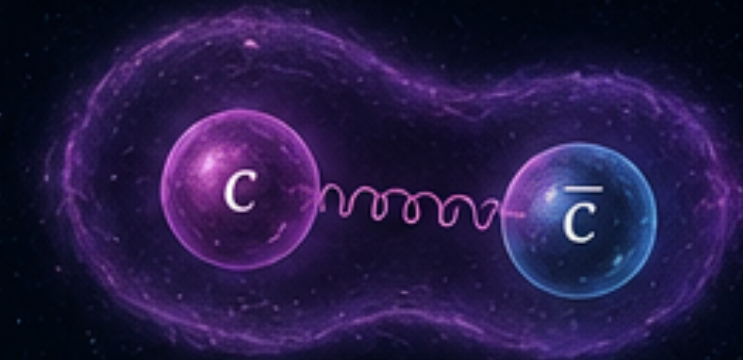
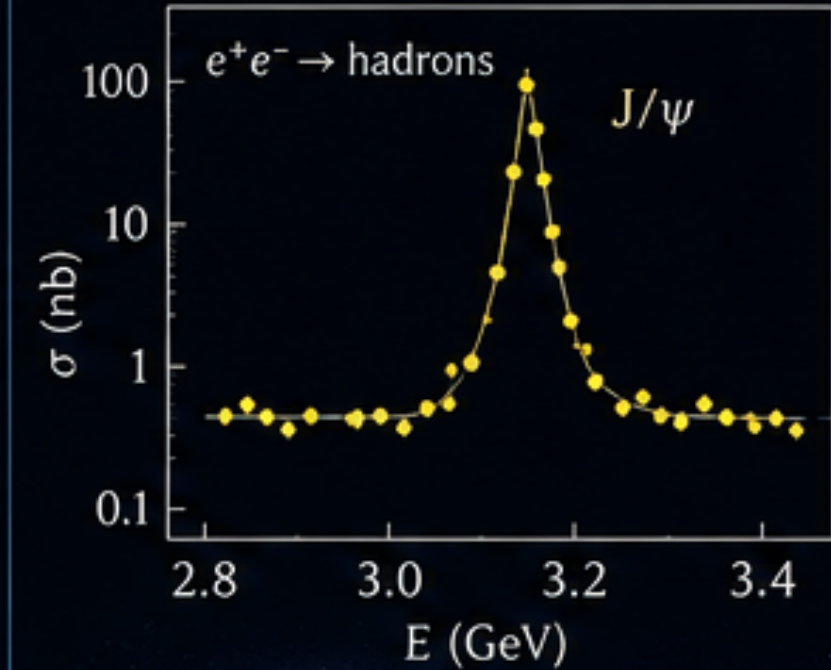
1947–195 Lamb, 1947



A tiny splitting (Lamb shift) in hydrogen revealed the effects of quantum vacuum fluctuations—confirmed by Quantum Electrodynamics, the most precise theory in physics.

5. DISCOVERY OF J/ψ

1974

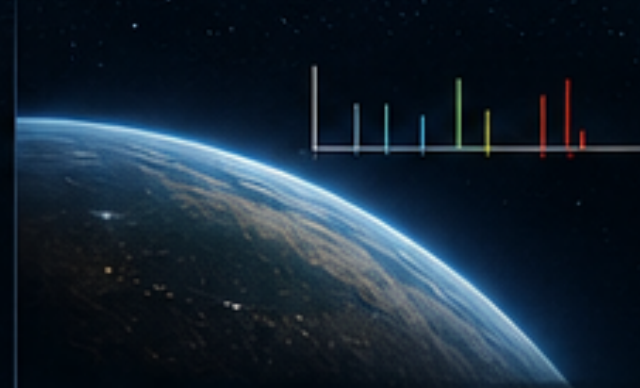


Spectroscopy of e^+e^- collisions revealed a new particle, the J/ψ , a bound state of charm quarks. This discovery opened the door to the rich field of the quark model and particle physics.

WHY SPECTROSCOPY MATTERS



Reveals the composition, temperature, density, and motion of stars, galaxies, and the cosmos.



Detects elements and molecules in planetary atmospheres, interstellar clouds, and exoplanets.



Essential tool in chemistry, materials science, and biology for identifying and quantifying matter.



Enables technologies from lasers and semiconductors to medical imaging and environmental monitoring.

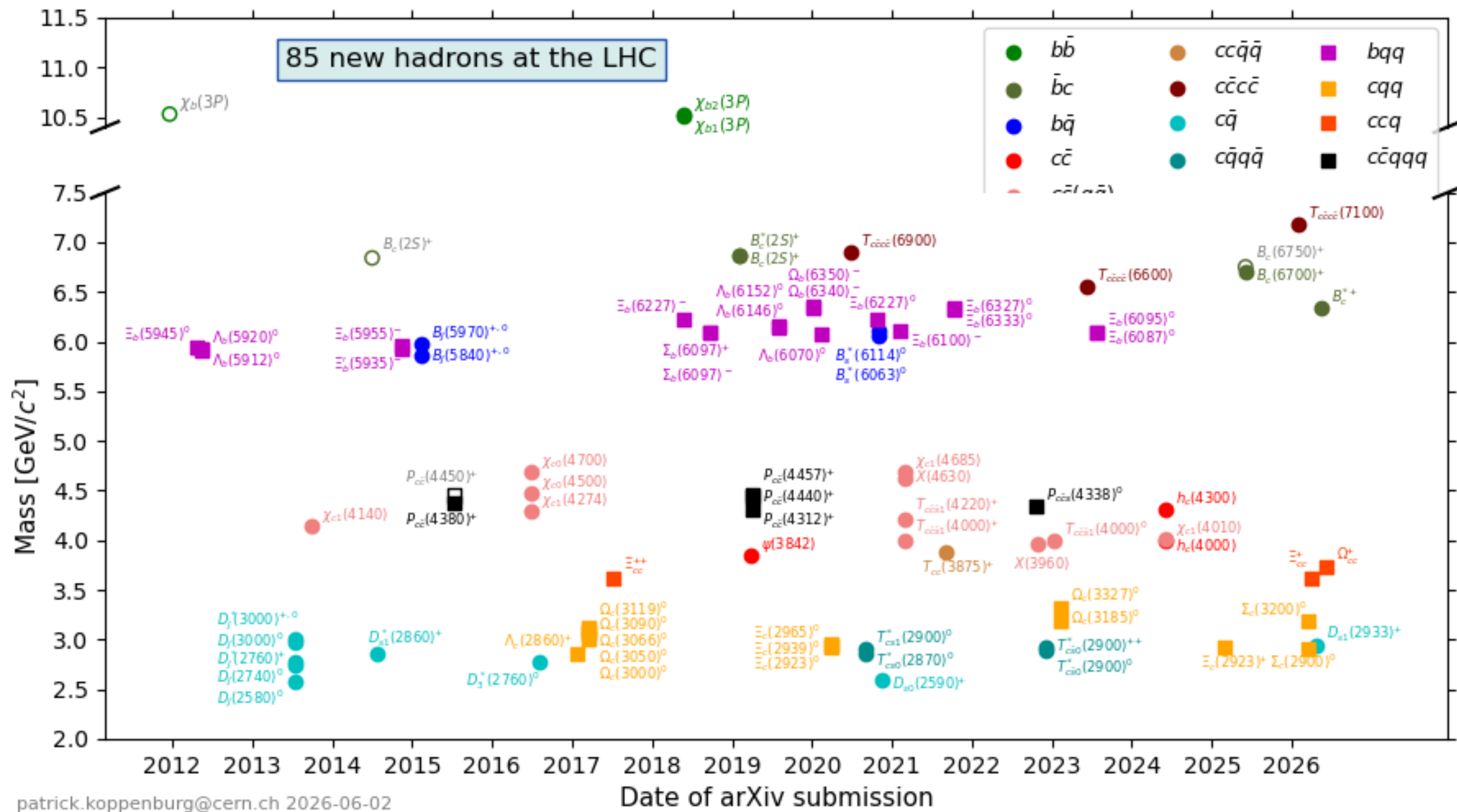


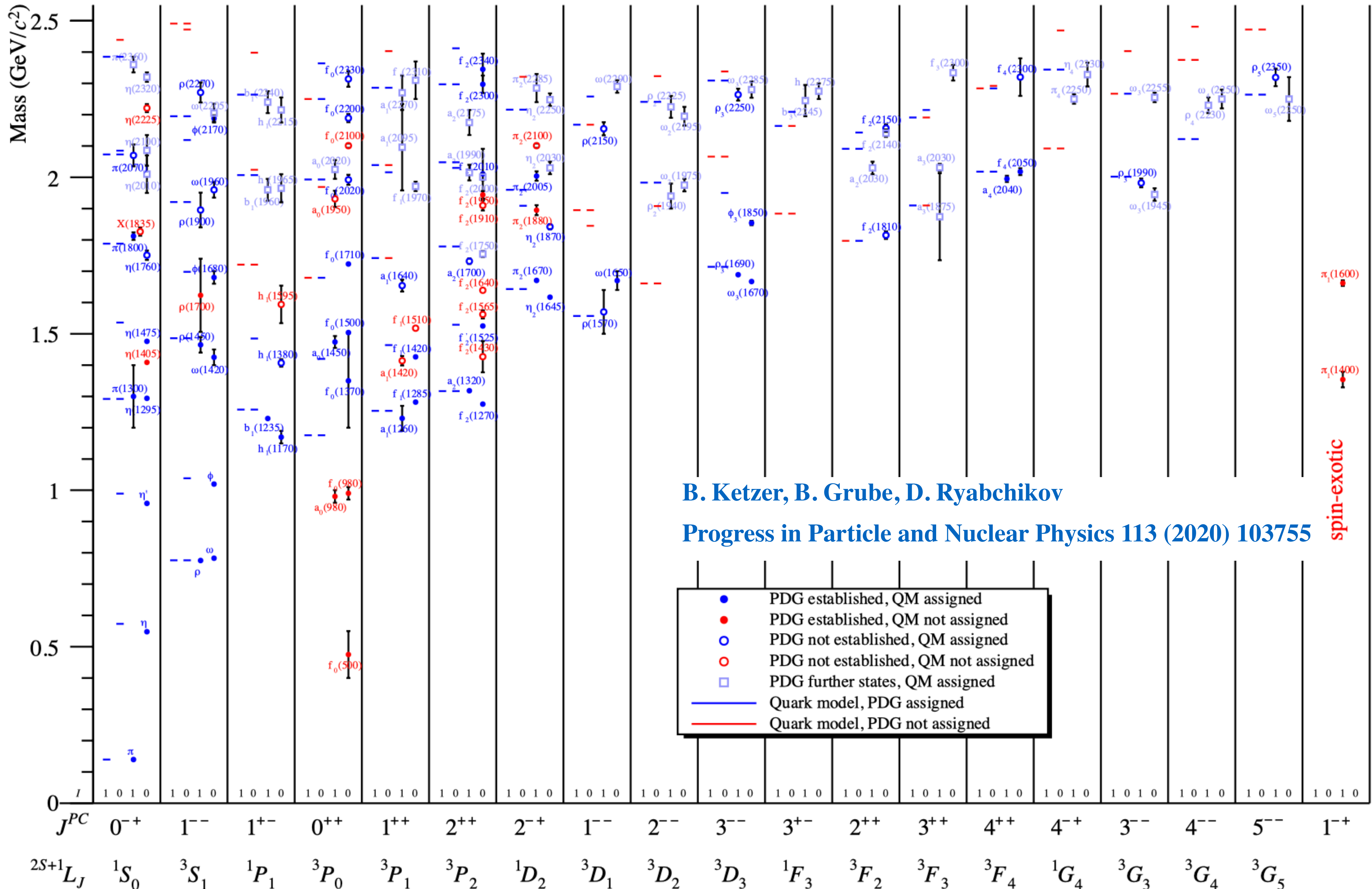
Tests fundamental physics and drives the development of new theories and technologies.

From prisms to particles, spectroscopy turns light into understanding—across all scales of nature.



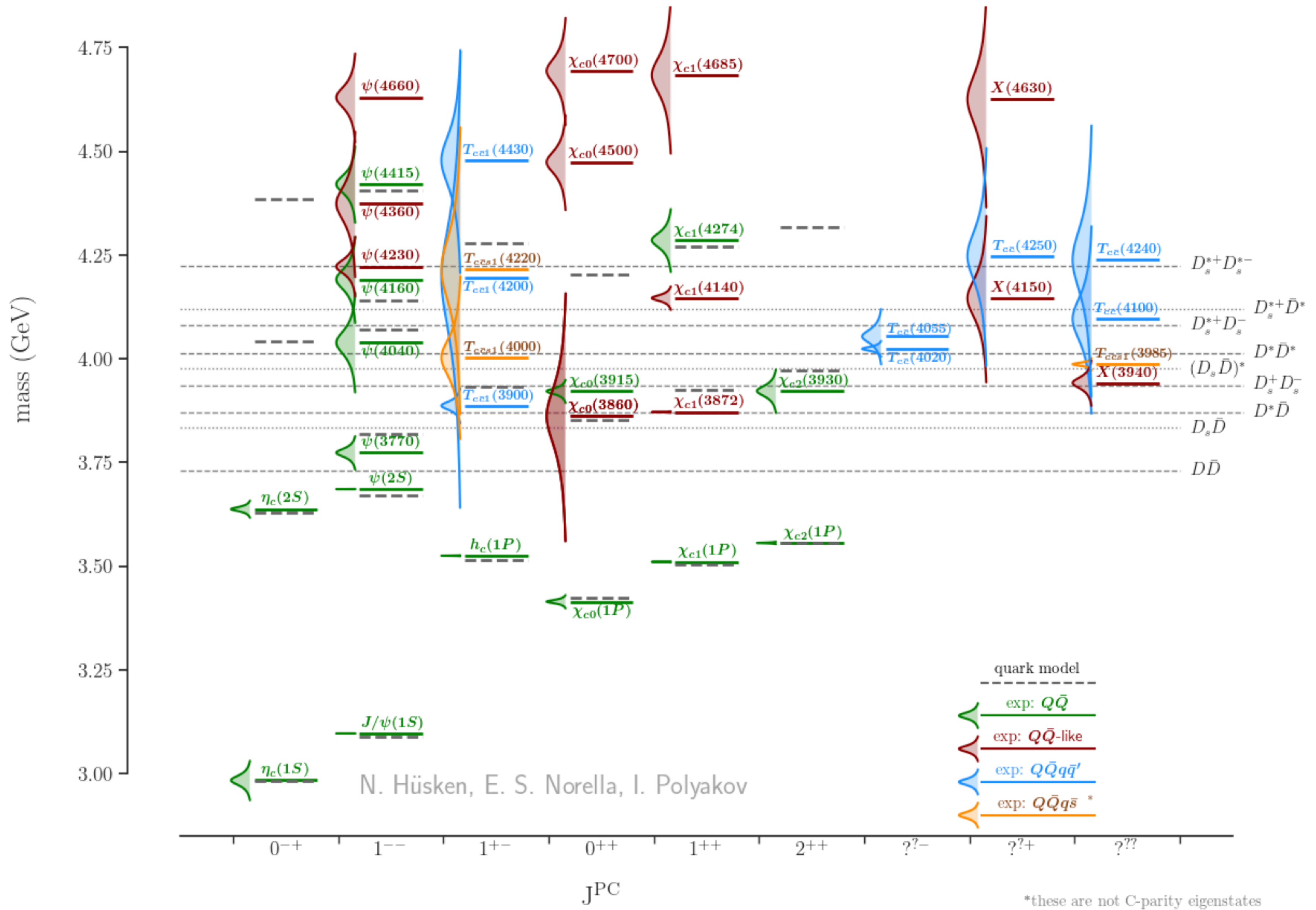
SPECTROSCOPY CONNECTS THE VERY SMALL TO THE VERY LARGE—AND CONTINUES TO REVEAL THE UNKNOWN.





B. Ketzer, B. Grube, D. Ryabchikov

Progress in Particle and Nuclear Physics 113 (2020) 103755



mass (GeV)

J^{PC}

*these are not C-parity eigenstates

N. Hüsken, E. S. Norella, I. Polyakov

hybrid hadrons and gluodynamics

C. Meyer & E.S. Swanson, arXiv:1502.07276

Past ideas for hybrid mesons

Volume 60B, number 2

PHYSICS LETTERS

5 January 1976

UNCONVENTIONAL STATES OF CONFINED QUARKS AND GLUONS[☆]

R.L. JAFFE^{*} and K. JOHNSON

*Laboratory for Nuclear Science and Department of Physics,
Massachusetts Institute of Technology, Cambridge, Mass. 02139, USA*

Received 27 October 1975

COLOURED QUARK AND GLUON CONSTITUENTS IN THE MIT BAG MODEL: A MODEL OF MESONS

Ted BARNES

Department of Physics, University of Southampton, Southampton SO9 5NH, England

Received 24 October 1977
(Revised 7 May 1979)

Volume 124B, number 3,4

PHYSICS LETTERS

28 April 1983

A FLUX TUBE MODEL FOR HADRONS

Nathan ISGUR^{1,2} and Jack PATON

Department of Theoretical Physics, University of Oxford, 1 Keble Road, Oxford, OX1, 3NP, England

Received 20 December 1982

VOLUME 37, NUMBER 18

PHYSICAL REVIEW LETTERS

1 NOVEMBER 1976

ψ Spectroscopy of a Charm String*

R. C. Giles and S.-H. H. Tye

Stanford Linear Accelerator Center, Stanford University, Stanford, California 94305

(Received 13 August 1976)

PHYSICAL REVIEW D

VOLUME 17, NUMBER 3

1 FEBRUARY 1978

Model of mesons with constituent gluons*

D. Horn[†]

California Institute of Technology, Pasadena, California 91125

J. Mandula[‡]

Massachusetts Institute of Technology, Cambridge, Massachusetts 02139

(Received 28 January 1977)

Lattice Hybrid Computations

Volume 129B, number 5

PHYSICS LETTERS

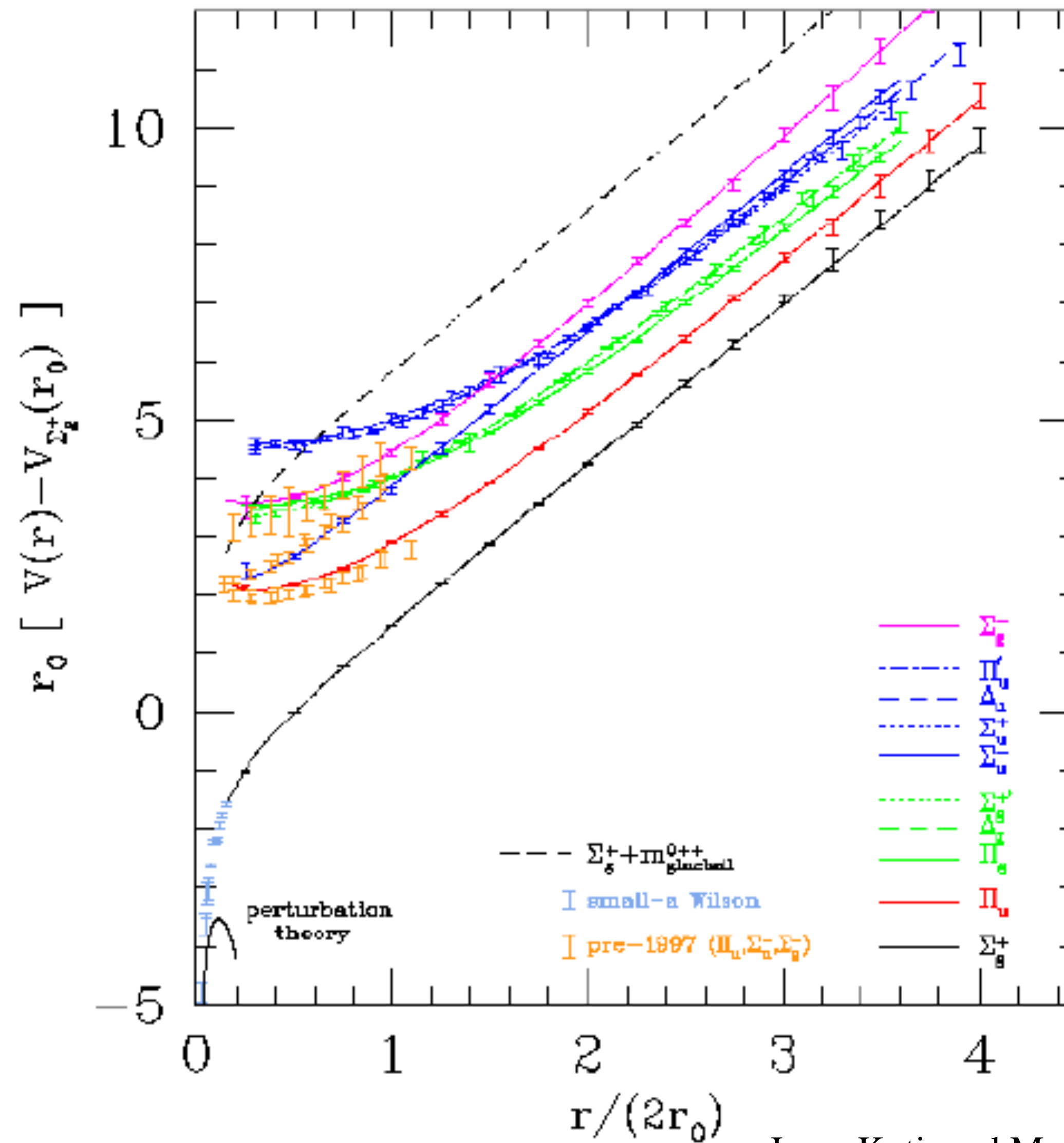
29 September 1983

MESONS WITH EXCITED GLUE

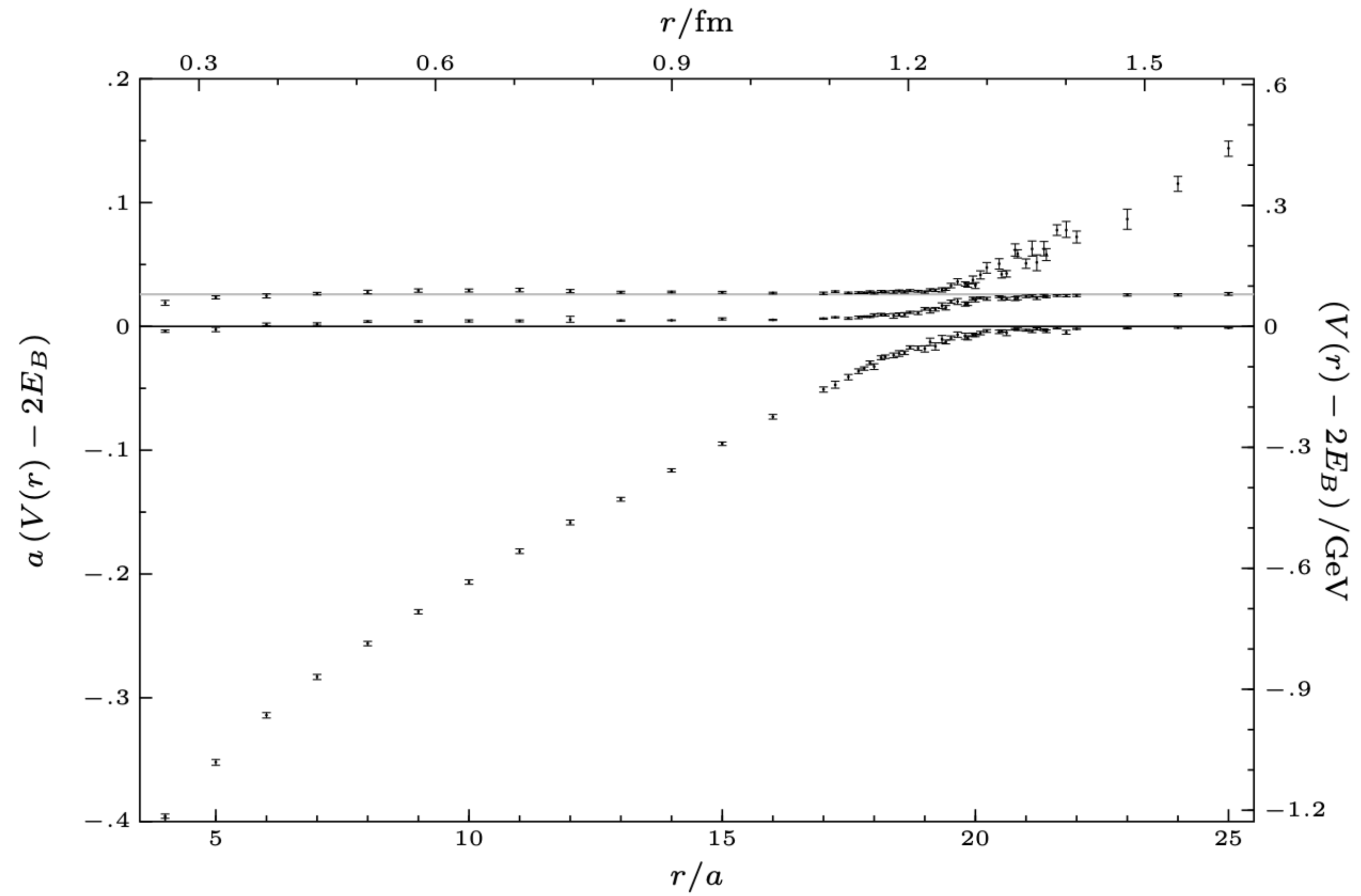
L.A. GRIFFITHS, C. MICHAEL and P.E.L. RAKOW

Department of Applied Mathematics and Theoretical Physics, University of Liverpool, P.O. Box 147, Liverpool L69 3BX, UK

Lattice Hybrid Computations



Lattice Hybrid Computations



Lattice Hybrid Computations

The 'gluelump' spectrum (static octet source + glue)

J^{PC}	mass (GeV)
1^{+-}	0.87(15)
1^{--}	1.25(16)
2^{--}	1.45(17)
2^{+-}	1.86(19)
3^{+-}	1.86(18)
0^{++}	1.98(18)
4^{--}	2.13(18)
1^{-+}	2.15(20)

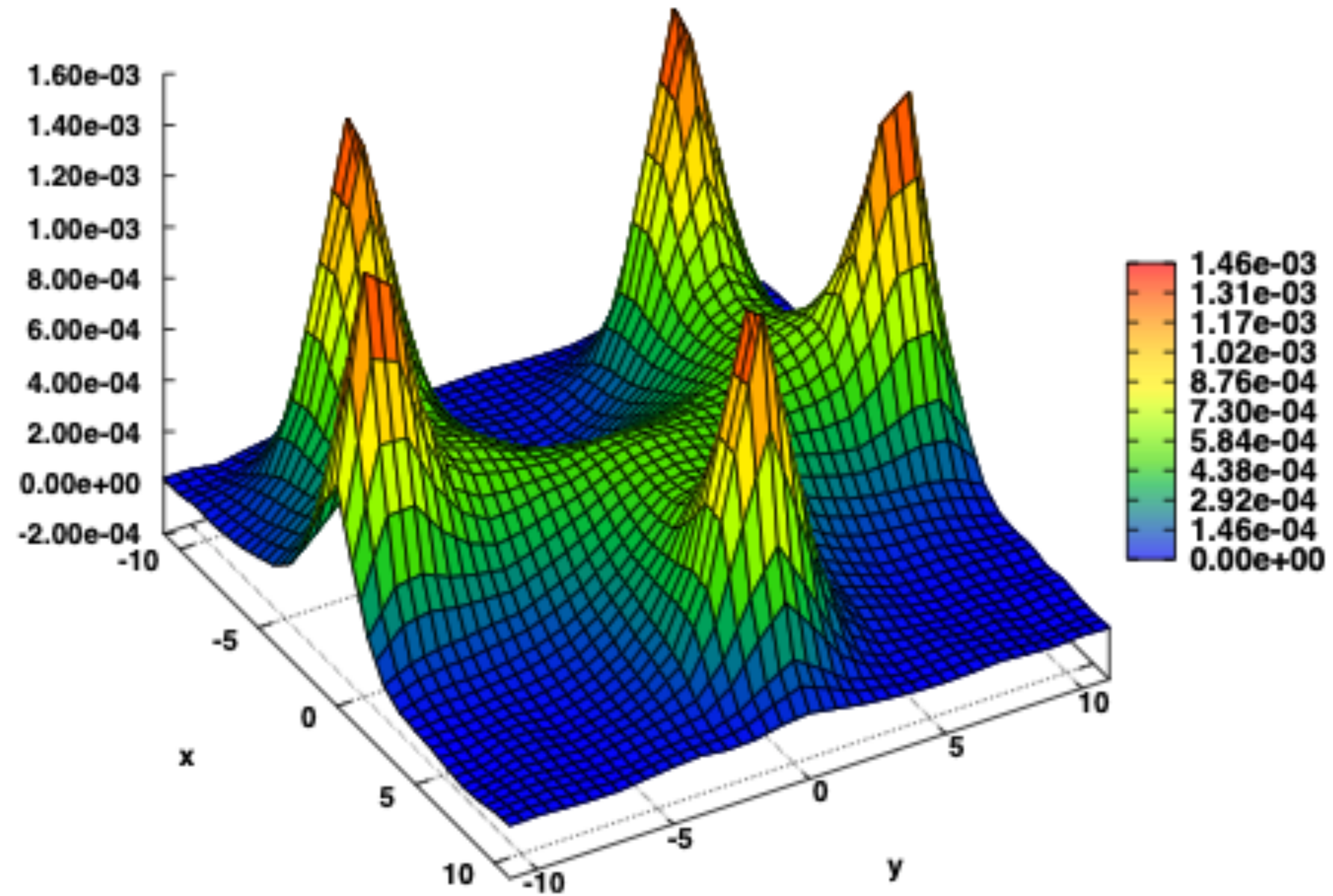
[only mass differences are well-defined]

M. Foster and C. Michael [UKQCD Collaboration], Phys. Rev. D 59, 094509 (1999).

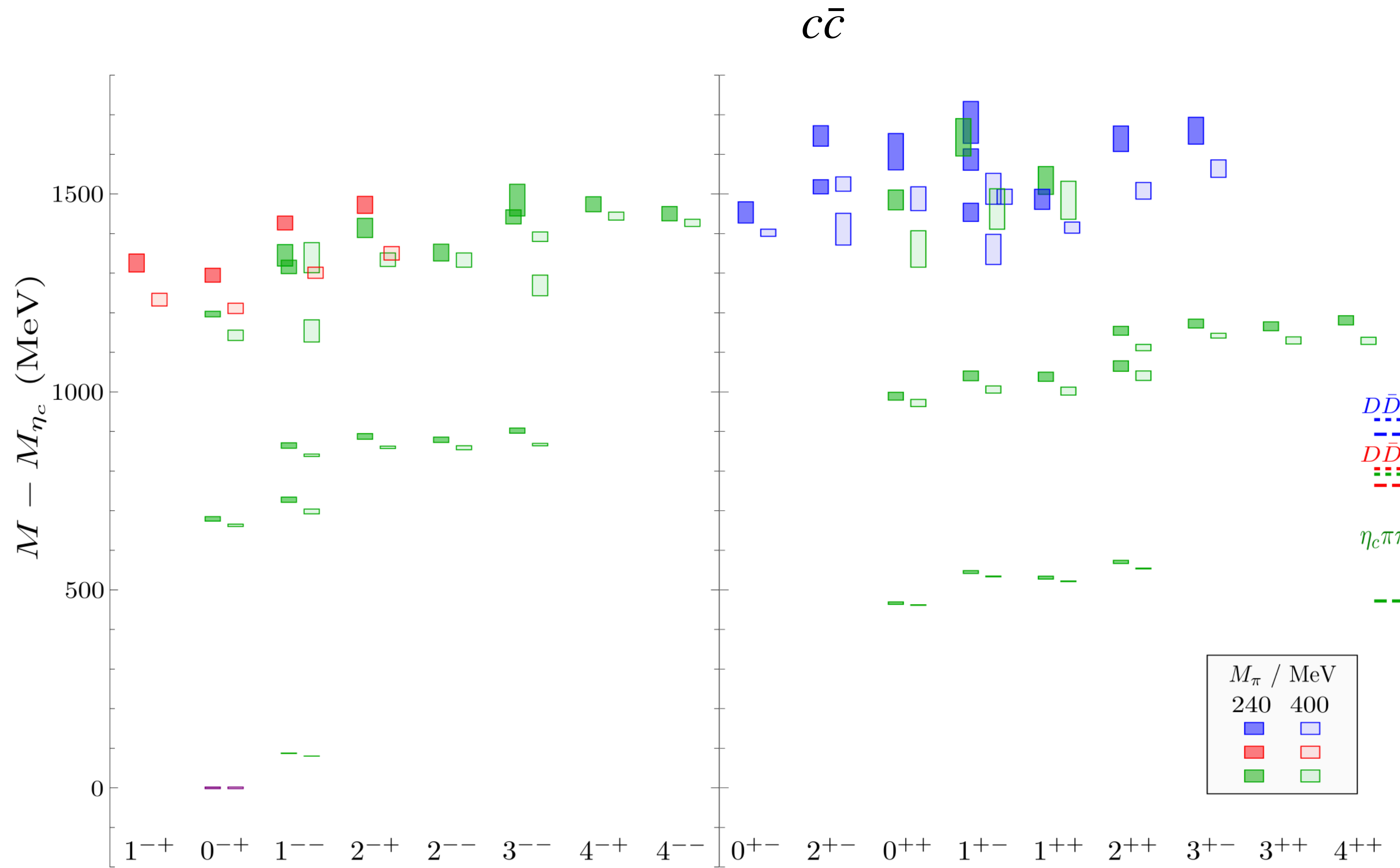
G. S. Bali and A. Pineda, Phys. Rev. D 69, 094001 (2004)

K. Marsh and R. Lewis, Phys. Rev. D 89, 014502 (2014)

Lattice Hybrid Computations



Lattice Hybrid Computations



Effective Field Theory

Obtain Schrödinger-type equations for heavy quark hybrids with pNRQCD.

$$\mathcal{L} = \text{tr} \left(H^{i\dagger} (\delta_{ij} i\partial_0 - h_{Hij}) H_j \right)$$
$$h_{Hij} = \left(-\frac{\nabla^2}{m_Q} + V_{\Sigma_u^-}(r) \right) \delta_{ij} + (\delta_{ij} - \hat{r}_i \hat{r}_j) \left[V_{\Pi_u}(r) - V_{\Sigma_u^-}(r) \right]$$

$$\left[-\frac{1}{mr^2} \partial_r r^2 \partial_r + \frac{1}{mr^2} \begin{pmatrix} l(l+1) + 2 & 2\sqrt{l(l+1)} \\ 2\sqrt{l(l+1)} & l(l+1) \end{pmatrix} + \begin{pmatrix} E_{\Sigma}^{(0)} & 0 \\ 0 & E_{\Pi}^{(0)} \end{pmatrix} \right] \begin{pmatrix} \psi_{\Sigma}^{(N)} \\ \psi_{-\Pi}^{(N)} \end{pmatrix} = \mathcal{E}_N \begin{pmatrix} \psi_{\Sigma}^{(N)} \\ \psi_{-\Pi}^{(N)} \end{pmatrix}$$

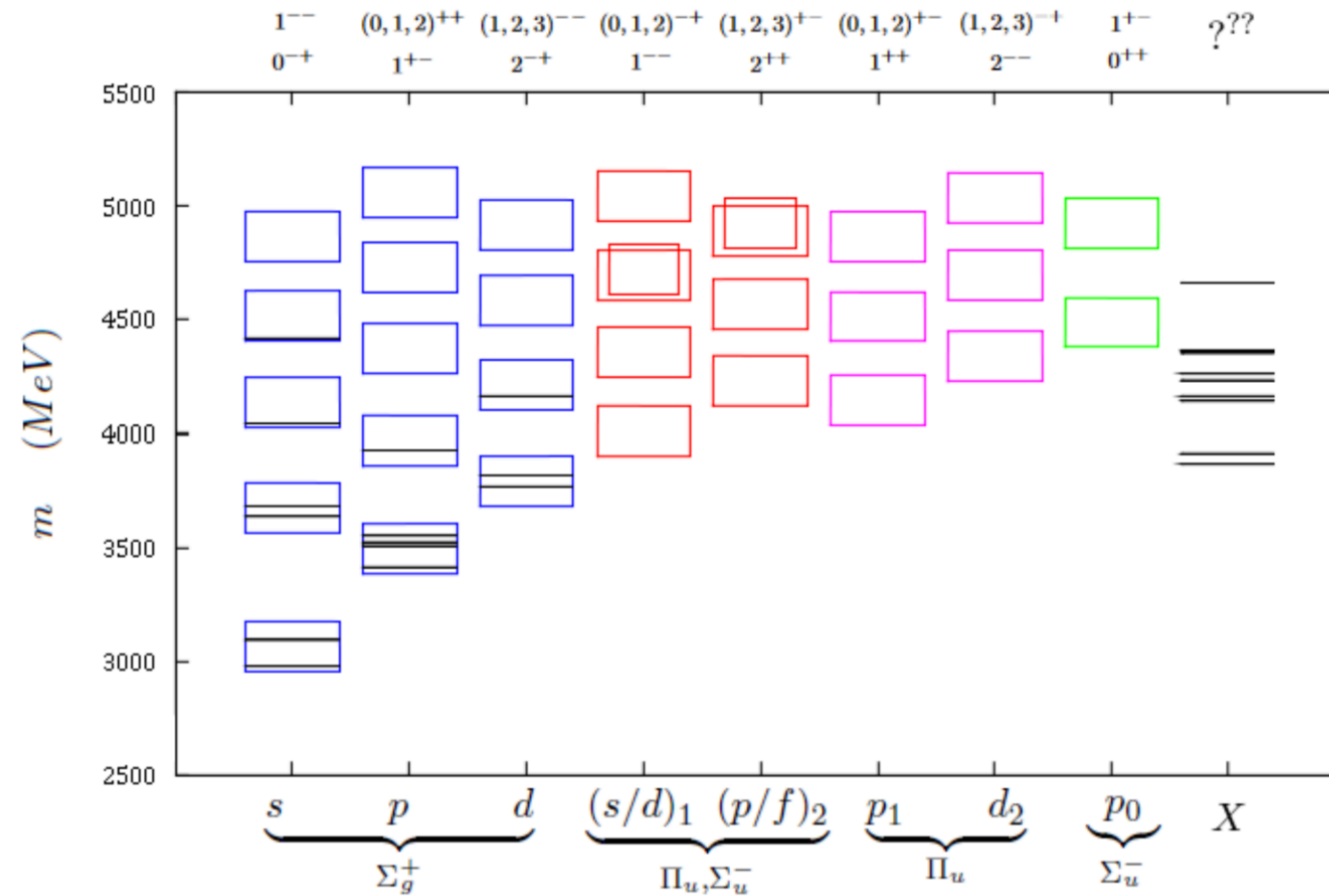
$$\left[-\frac{1}{mr^2} \partial_r r^2 \partial_r + \frac{l(l+1)}{mr^2} + E_{\Pi}^{(0)} \right] \psi_{+\Pi}^{(N)} = \mathcal{E}_N \psi_{+\Pi}^{(N)}$$

M. Berwein, N. Brambilla, J. Castella, A. Vairo, arXiv:1510.04299

R. Oncala and J. Soto, arXiv: 1702.03900

N. Brambilla, W.-K. Lai, J. Segovia, J. Castella, A. Vairo, arXiv:1805.07713

Effective Field Theory



Born-Oppenheimer Approximation

all other models fail to reproduce this pattern

multiplet $\vec{\sigma}$

H_1	$\psi^\dagger \vec{B}_\chi$	1^{--}	$(0, 1, 2)^{-+}$
H_2	$\psi^\dagger \nabla \times \vec{B}_\chi$	1^{++}	$(0, 1, 2)^{+-}$
H_3	$\psi^\dagger \nabla \cdot \vec{B}_\chi$	0^{++}	1^{+-}
H_4	$\psi^\dagger [\nabla \vec{B}]_2 \chi$	2^{++}	$(1, 2, 3)^{+-}$

constituent gluon with $(J^{PC})_g = 1^{+-}$

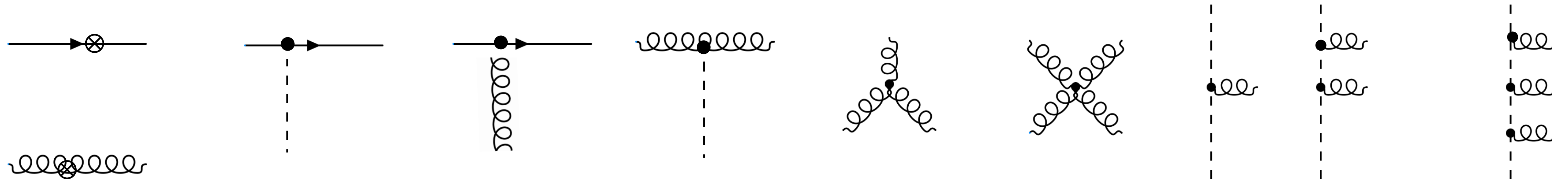
modelling hybrids

Hamiltonian QCD

Work in Coulomb gauge

$$H_{QCD} = \int d^3x \left[\psi^\dagger (-i\boldsymbol{\alpha} \cdot \nabla + \beta m) \psi + \frac{1}{2} (\mathcal{F}^{-1/2} \boldsymbol{\Pi} \mathcal{F} \cdot \boldsymbol{\Pi} \mathcal{F}^{-1/2} + \mathbf{B} \cdot \mathbf{B}) - g \psi^\dagger \boldsymbol{\alpha} \cdot \mathbf{A} \psi \right] + H_C$$

$$H_C = \frac{1}{2} \int d^3x d^3y \mathcal{F}^{-1/2} \rho^A(\mathbf{x}) \mathcal{F}^{1/2} \langle \mathbf{x}, A | \frac{g}{\nabla \cdot \mathbf{D}} (-\nabla^2) \frac{g}{\nabla \cdot \mathbf{D}} | \mathbf{y}, B \rangle \mathcal{F}^{1/2} \rho^B(\mathbf{y}) \mathcal{F}^{-1/2}$$

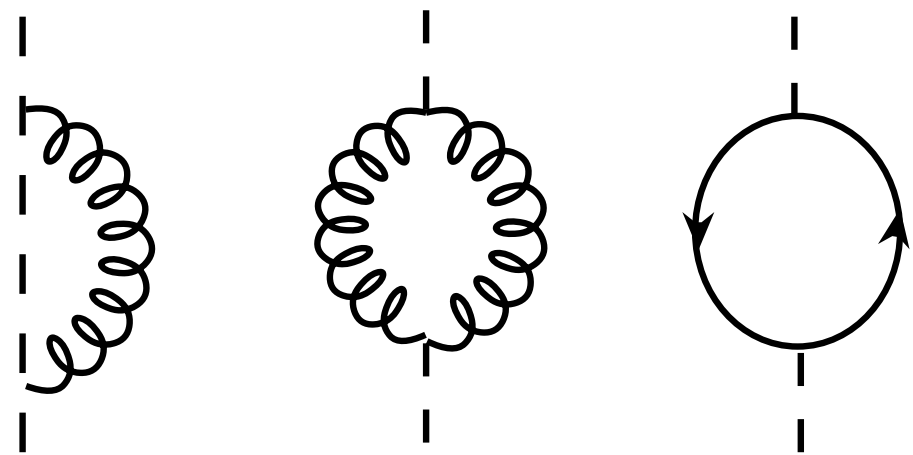


Hamiltonian QCD

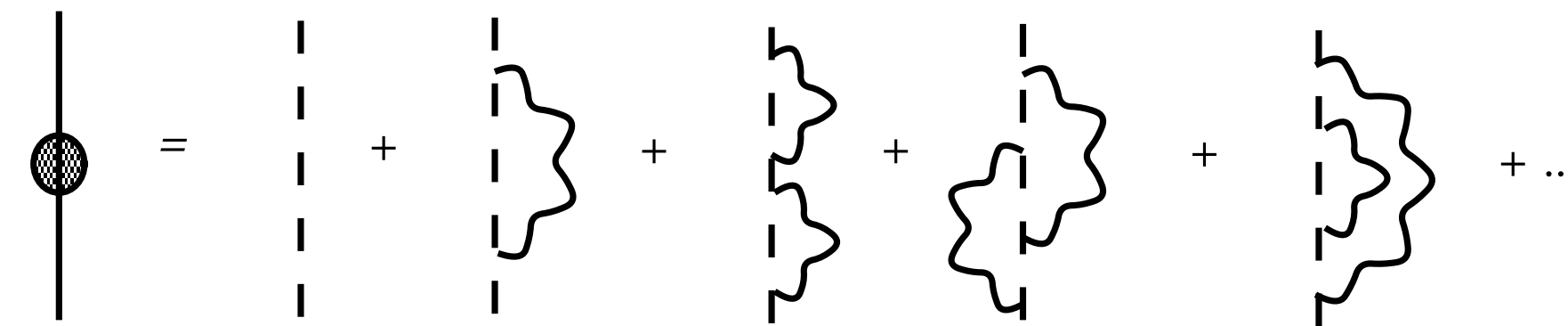
vacuum Ansatz

$$\Psi_0[\mathbf{A}] = \langle \mathbf{A} | \omega \rangle = \exp \left[-\frac{1}{2} \int \frac{d\mathbf{k}}{(2\pi)^3} \mathbf{A}^a(\mathbf{k}) \omega(k) \mathbf{A}^a(-\mathbf{k}) \right]$$

running coupling

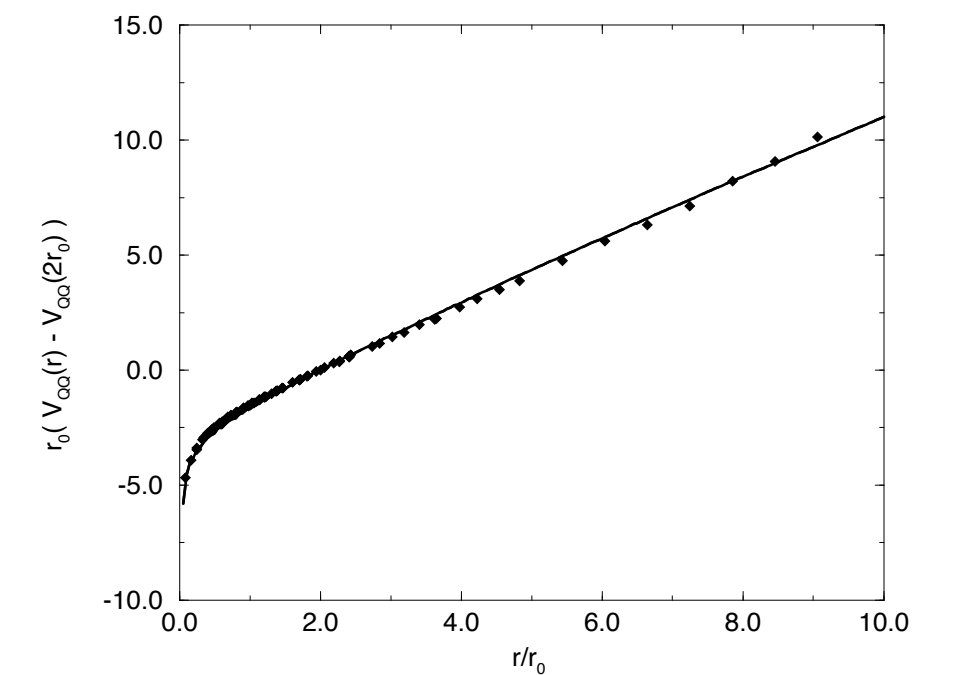


nonperturbative potential

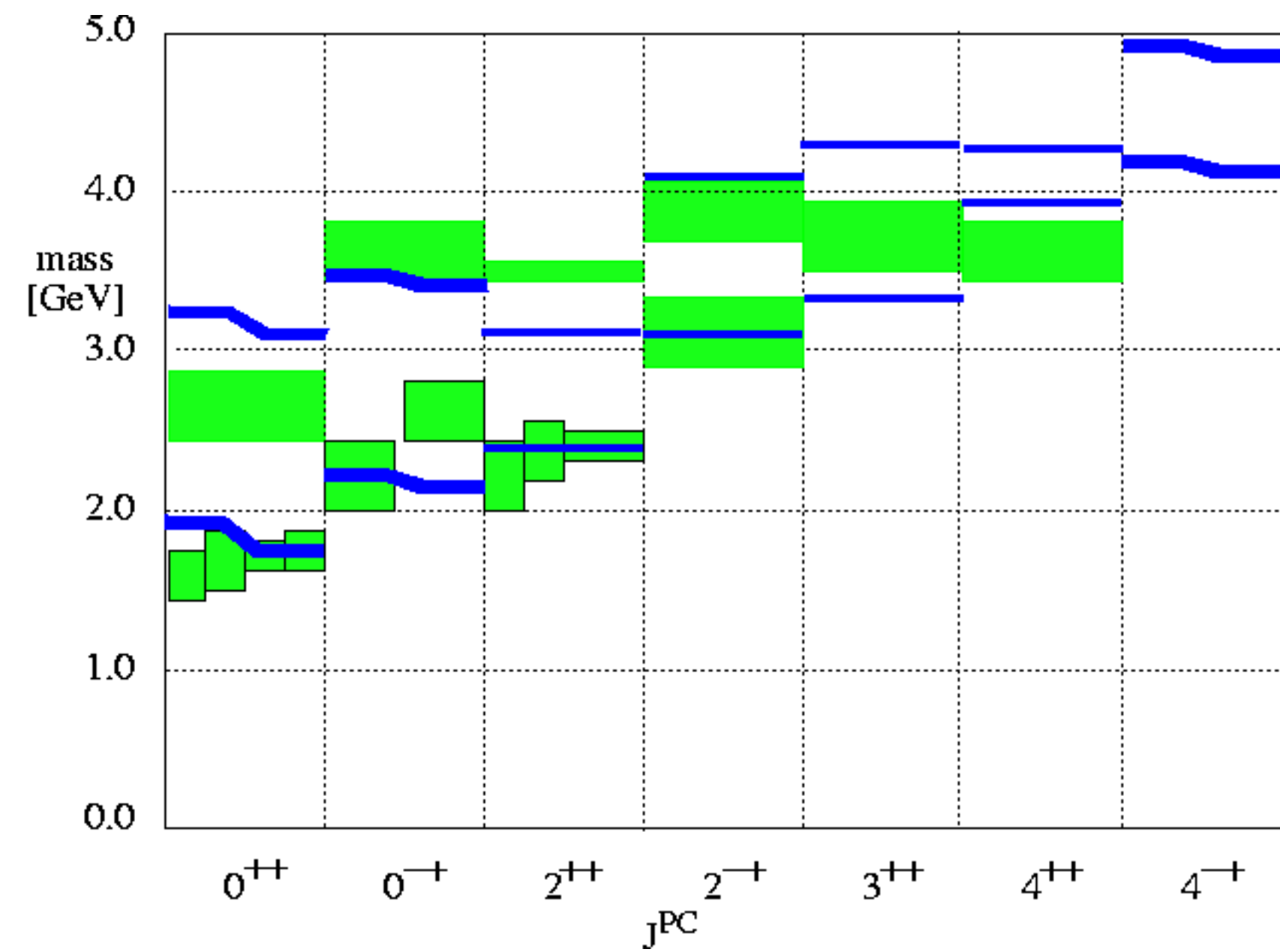
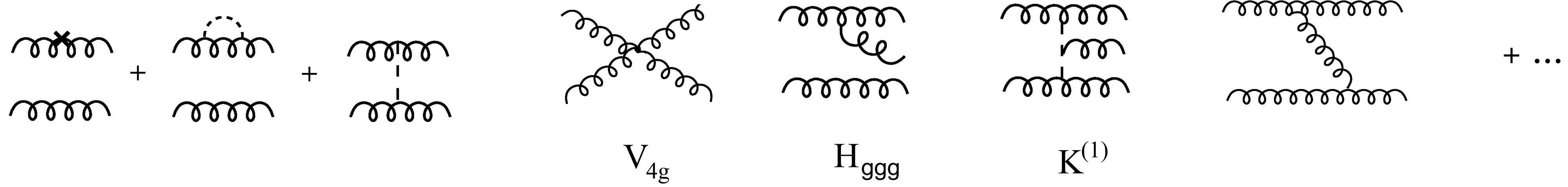


gluonic gap equation

$$\frac{\delta}{\delta\omega} \left(\text{gluon loop} + \text{gluon self-energy} + \text{ghost loop} \right) = 0$$

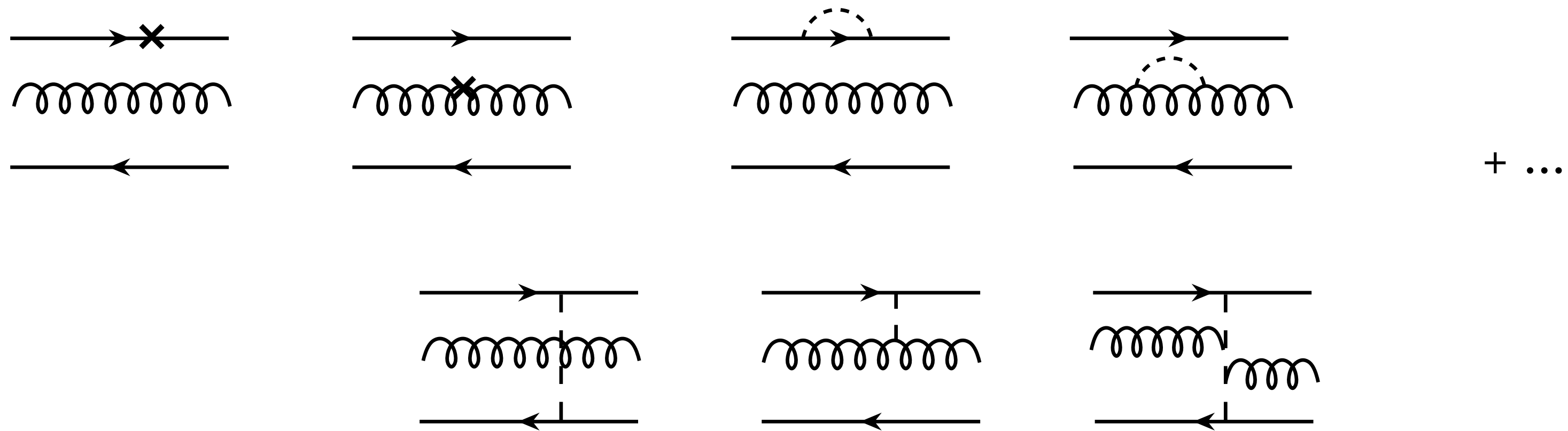


exotic states - glueballs

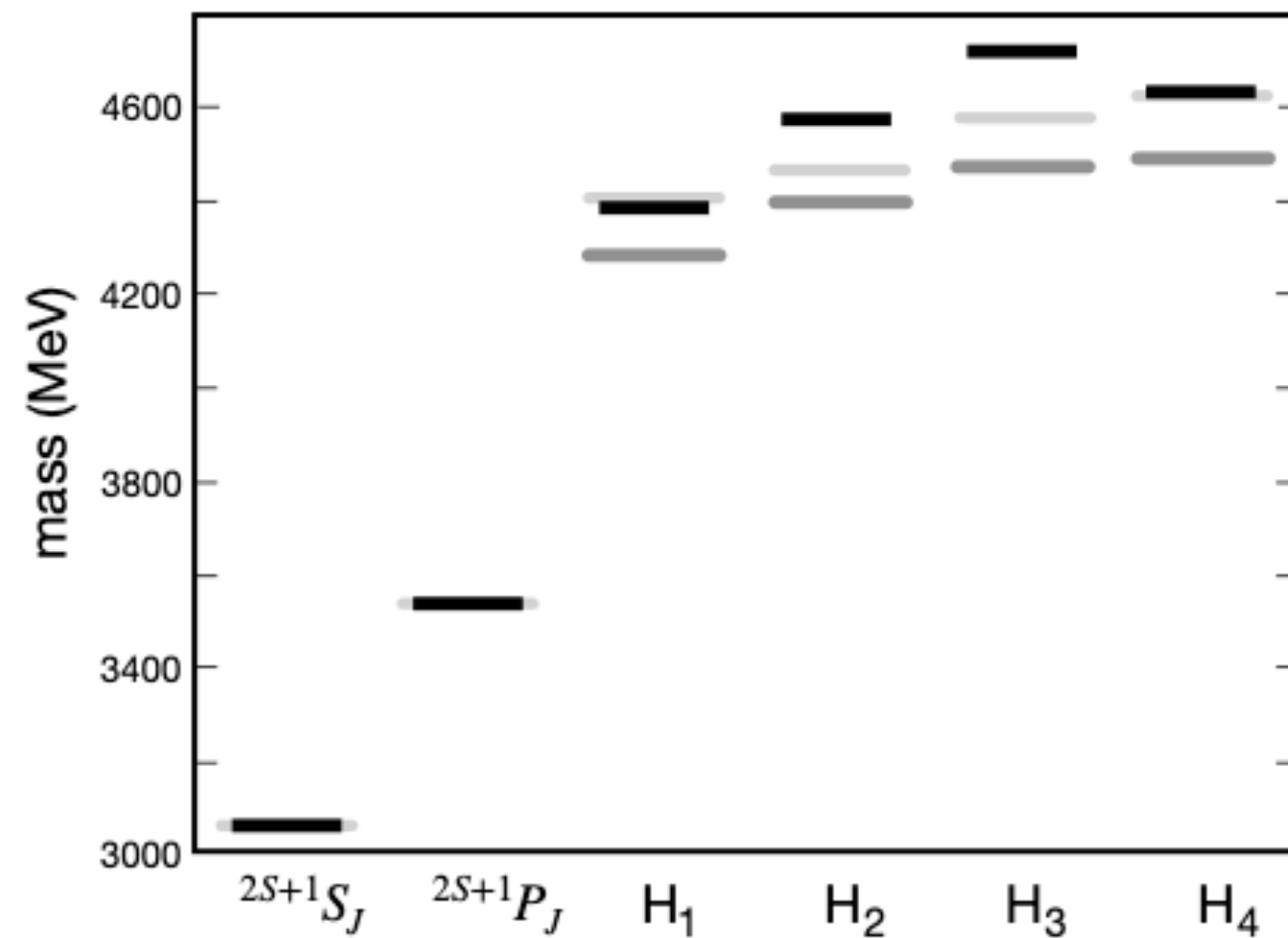


Swanson & Szczepaniak, arXiv:[hep-ph/0308268](https://arxiv.org/abs/hep-ph/0308268)

exotic states - hybrids

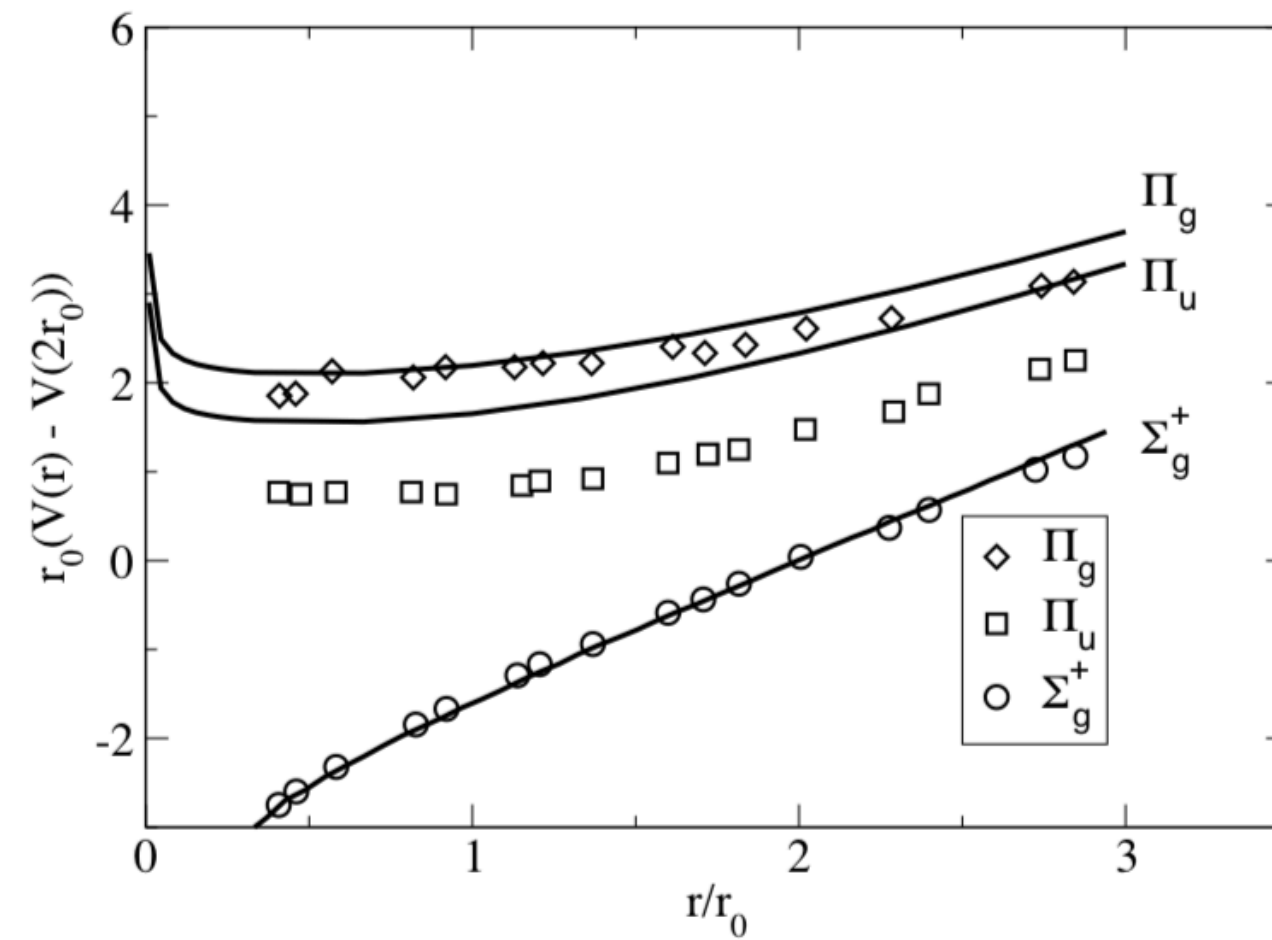


$c\bar{c}g$ spectrum



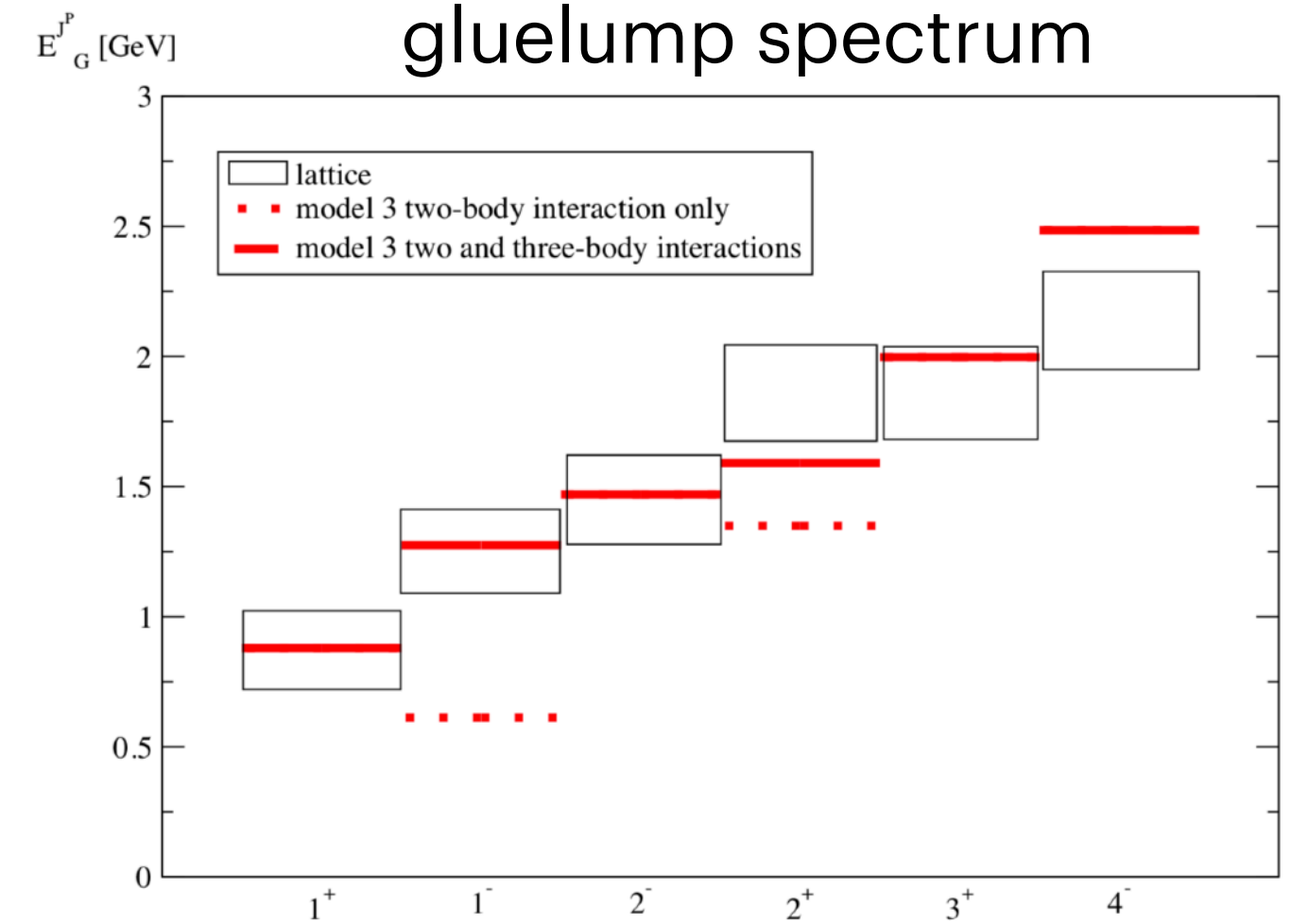
C. Farina et al, arXiv:2005.10850

adiabatic potentials



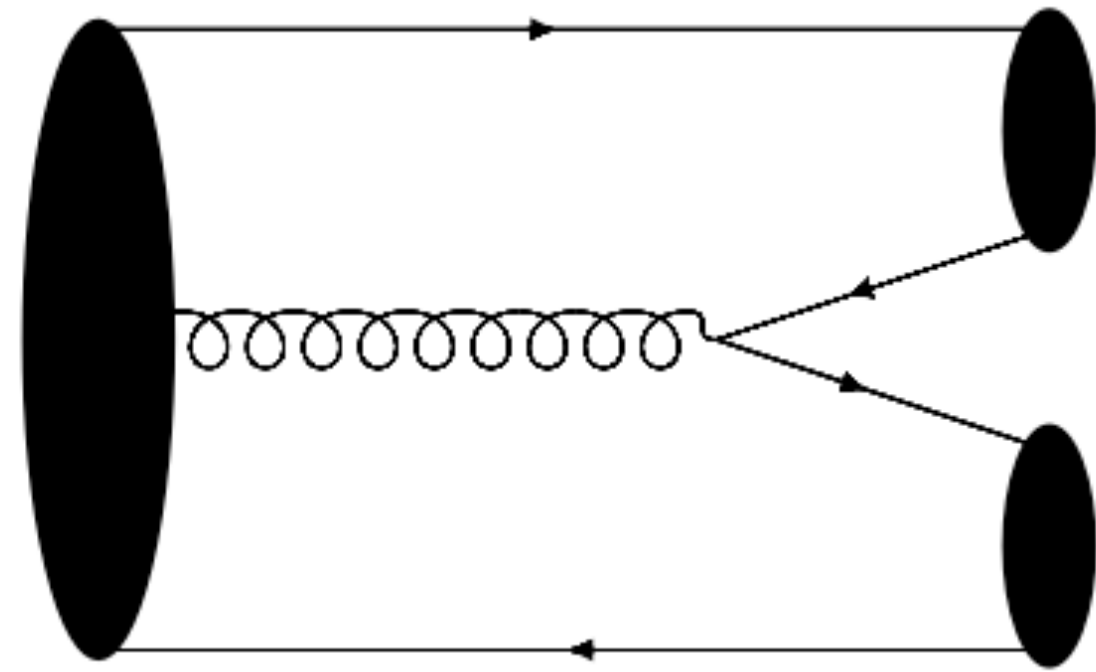
A. Szczepaniak and P. Krupinski, Phys. Rev. D 73, 116002 (2006).

gluelump spectrum



P. Guo, et al, arXiv:0707.3156

hybrid decays



spin selection rule

S+P selection rule

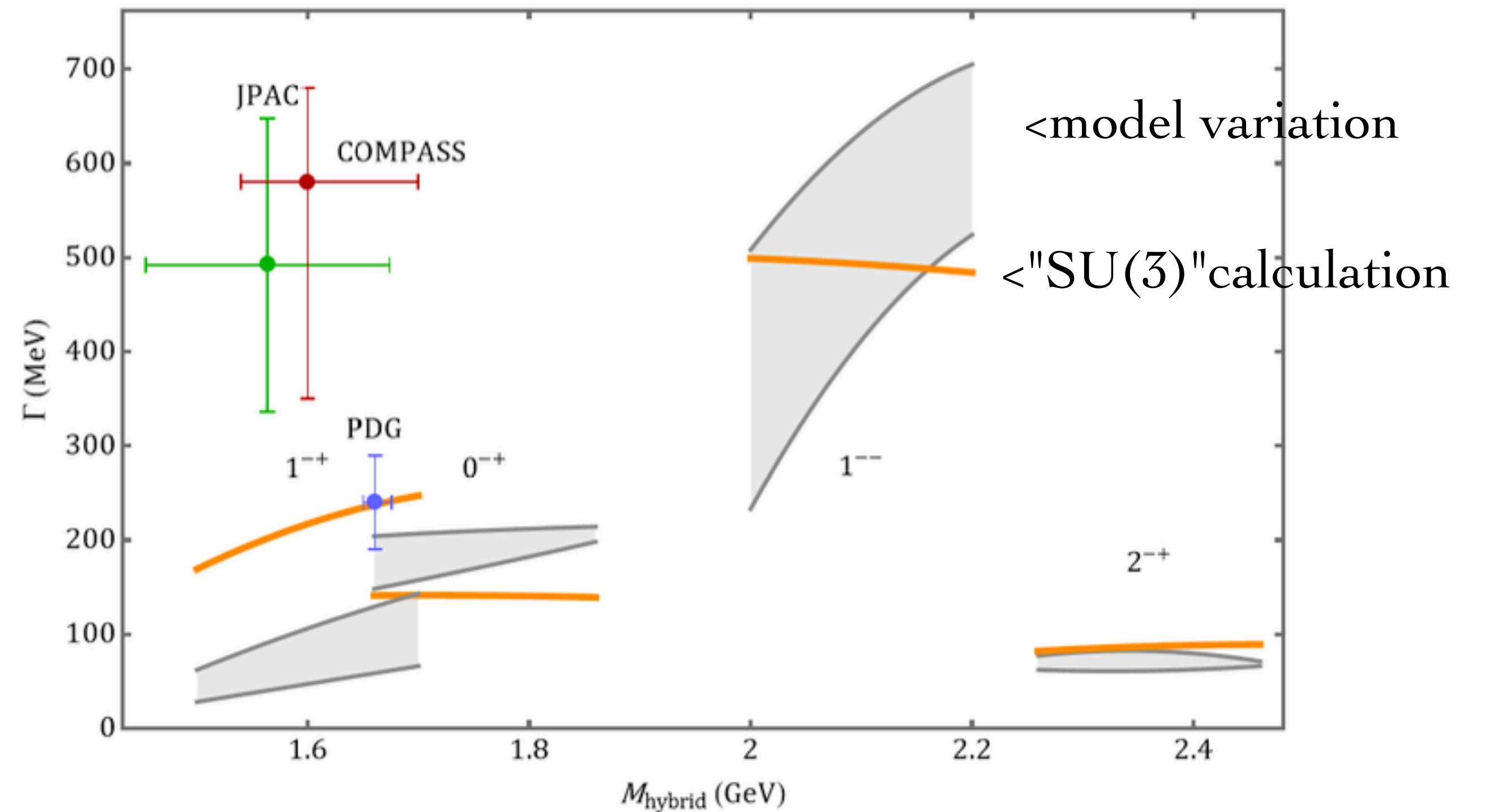
C. Farina & E. Swanson, arXiv:2512.00459

C. Farina & E. Swanson, arXiv:2312.05370

C. Farina et al, arXiv:2005.10850

P. Page, E.S. Swanson, A. Szczepaniak, arXiv:hep-ph/9808346

E.S. Swanson & A. Szczepaniak, arXiv:hep-ph/9704434

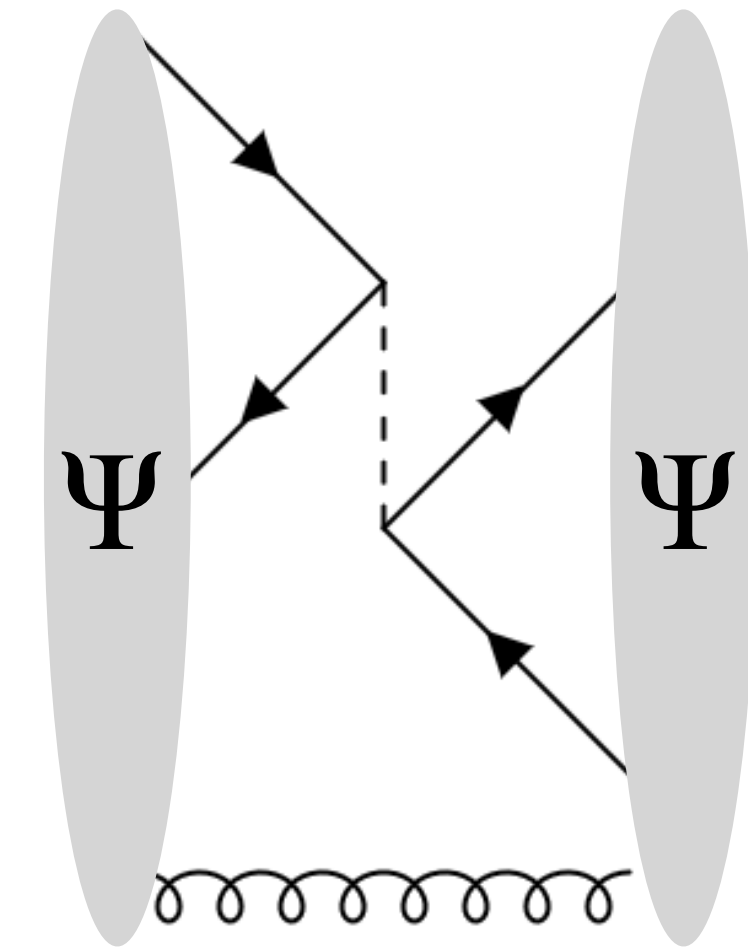


hybrid flavour mixing

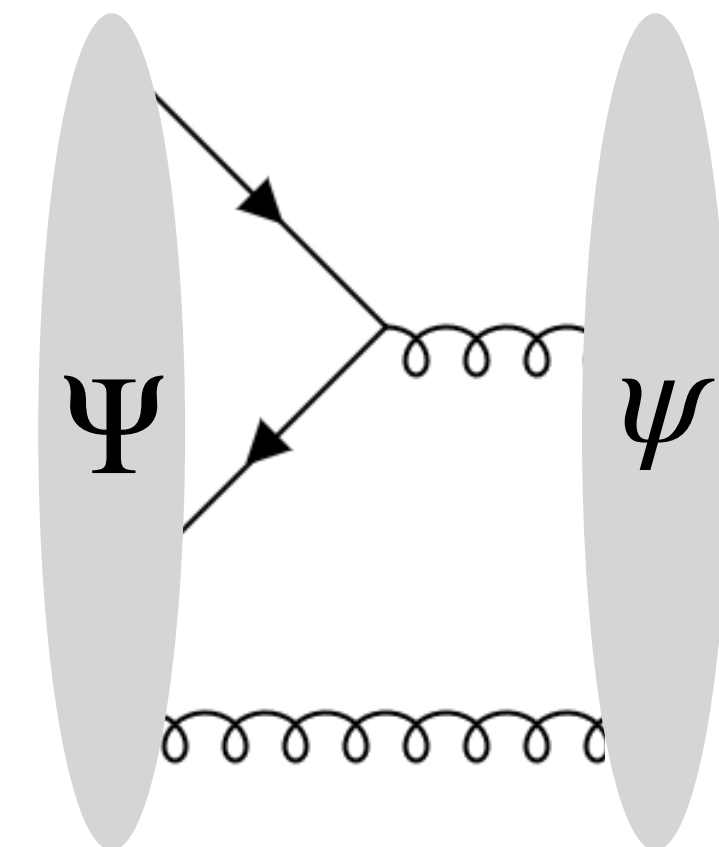
$$H_{uds} = \begin{matrix} & |u\bar{u}\rangle & |d\bar{d}\rangle & |s\bar{s}\rangle & |gg\rangle \\ \begin{pmatrix} m + A_{nn} & A_{nn} & A_{ns} & \mathcal{A}_n^{(0)} \\ A_{nn} & m + A_{nn} & A_{ns} & \mathcal{A}_n^{(0)} \\ A_{ns} & A_{ss} & m + \Delta m + A_{ss} & \mathcal{A}_s^{(0)} \\ \mathcal{A}_n^{(0)} & \mathcal{A}_n^{(0)} & \mathcal{A}_s^{(0)} & M_{gb} \end{pmatrix} \end{matrix}.$$



$$H_{iso} = \begin{pmatrix} m & 0 & 0 & 0 \\ 0 & m + 2A_{nn} & \sqrt{2}A_{ns} & \sqrt{2}\mathcal{A}_n^{(0)} \\ 0 & \sqrt{2}A_{ns} & m + \Delta m + A_{ss} & \mathcal{A}_s^{(0)} \\ 0 & \sqrt{2}\mathcal{A}_n^{(0)} & \mathcal{A}_s^{(0)} & M_{gb} \end{pmatrix}.$$

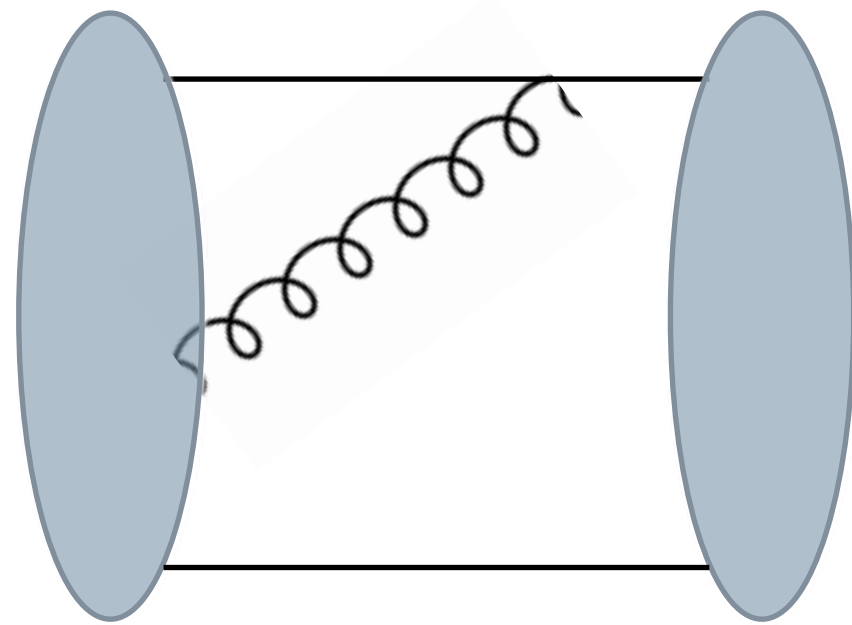


$\ell = 0; S = 1; H_1$ only



[neglect mixing through the continuum ~ the "OZI puzzle"]

hybrid-vector mixing



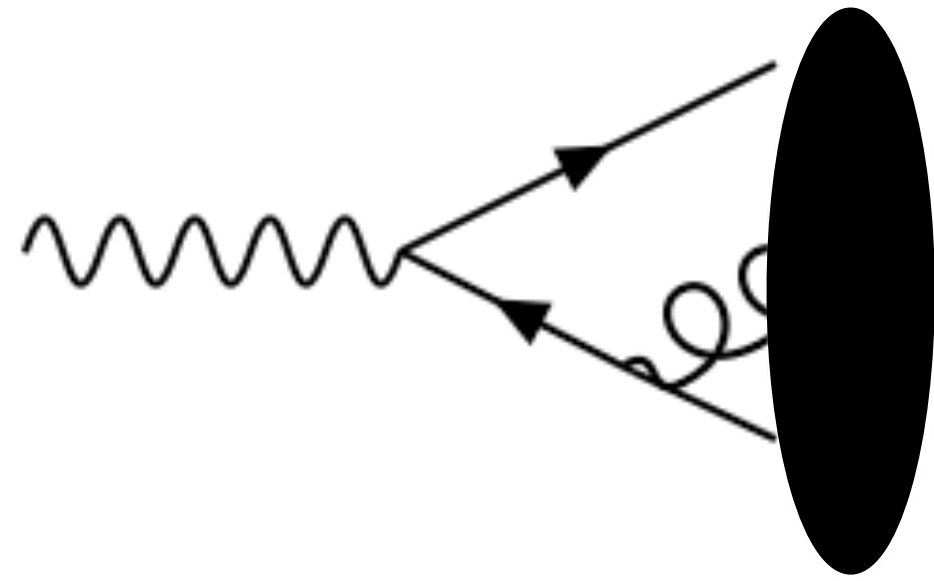
$$\mathcal{H}_{1S} = -ig \begin{cases} 84 \text{ MeV}^2/m_q, & \rho \\ 190 \text{ MeV}^2/m_c, & J/\psi \\ 225 \text{ MeV}^2/m_b, & \Upsilon \end{cases} \approx -i \begin{cases} 210 \text{ MeV}, & \rho \\ 60 \text{ MeV}, & J/\psi \\ 20 \text{ MeV}, & \Upsilon \end{cases}$$

Hybrid configuration content of heavy S-wave mesons,
[MILC] T. Burch & D. Toussaint, Phys. Rev. 68, 094504 (2003).

$$\mathcal{H}_{NRQCD} \approx 170 \text{ MeV} (J/\psi)$$

$$\mathcal{H}_{NRQCD} \approx 70 \text{ MeV} (\Upsilon).$$

vector hybrid decay constant



$$f_H = \frac{1}{\sqrt{M_H}} \sum_{n \neq H} \sqrt{M_n} f_V^{(n)} C_n$$

$$f_V^{(n)} = \sqrt{\frac{3}{M_n}} \int \frac{d^3k}{(2\pi)^3} \psi^{(n)}(\vec{k}) \sqrt{1 + \frac{m_q}{E_k}} \sqrt{1 + \frac{m_{\bar{q}}}{E_{\bar{k}}}} \left(1 + \frac{k^2}{3(E_k + m_q)(E_{\bar{k}} + m_{\bar{q}})} \right).$$

$$C_n = \langle nS | H_1(1^{--}) \rangle$$

$$\delta H = \begin{pmatrix} m_1 & & & & & & \mathcal{H}_{1S} \\ & m_2 & & & & & \mathcal{H}_{2S} \\ & & m_3 & & & & \mathcal{H}_{3S} \\ & & & m_4 & & & \mathcal{H}_{4S} \\ & & & & m_5 & & \mathcal{H}_{5S} \\ & & & & & m_6 & \mathcal{H}_{6S} \\ \mathcal{H}_{1S} & \mathcal{H}_{2S} & \mathcal{H}_{3S} & \mathcal{H}_{4S} & \mathcal{H}_{5S} & \mathcal{H}_{6S} & M_H \end{pmatrix}.$$

$$f_{H_1(1^{--})} \approx 20 \text{ MeV}.$$

hybrid phenomenology

lattice light hybrid spectrum

11

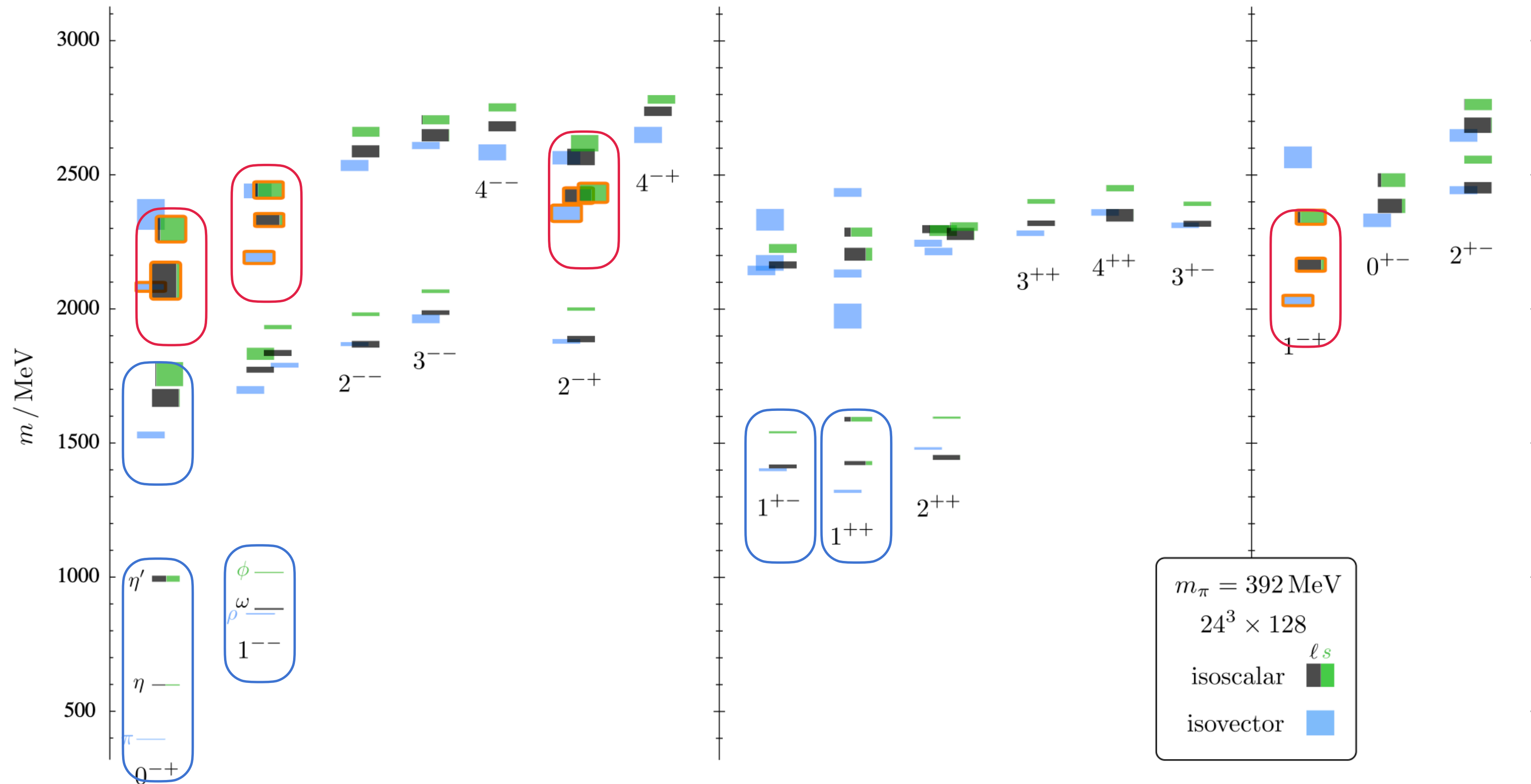
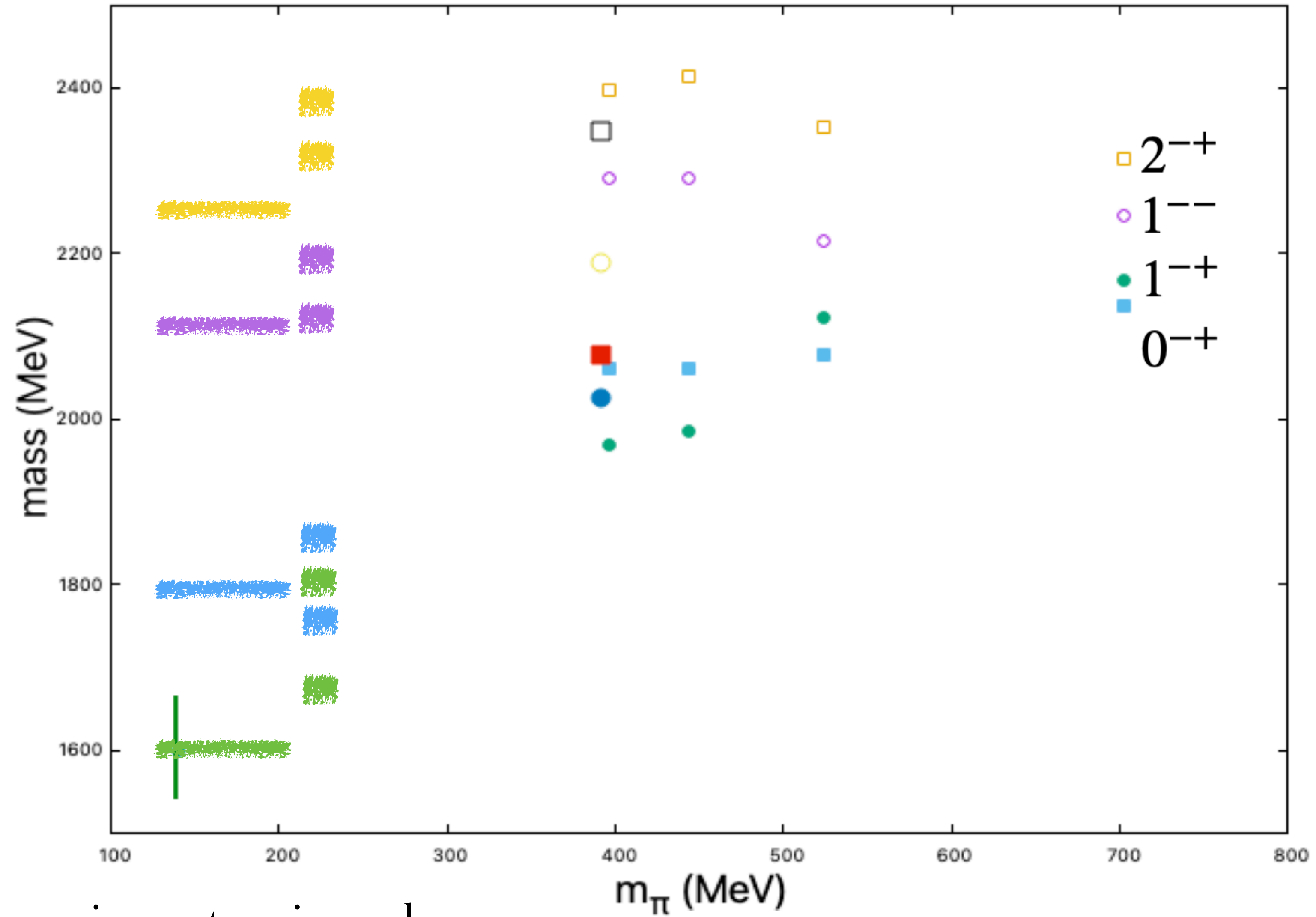


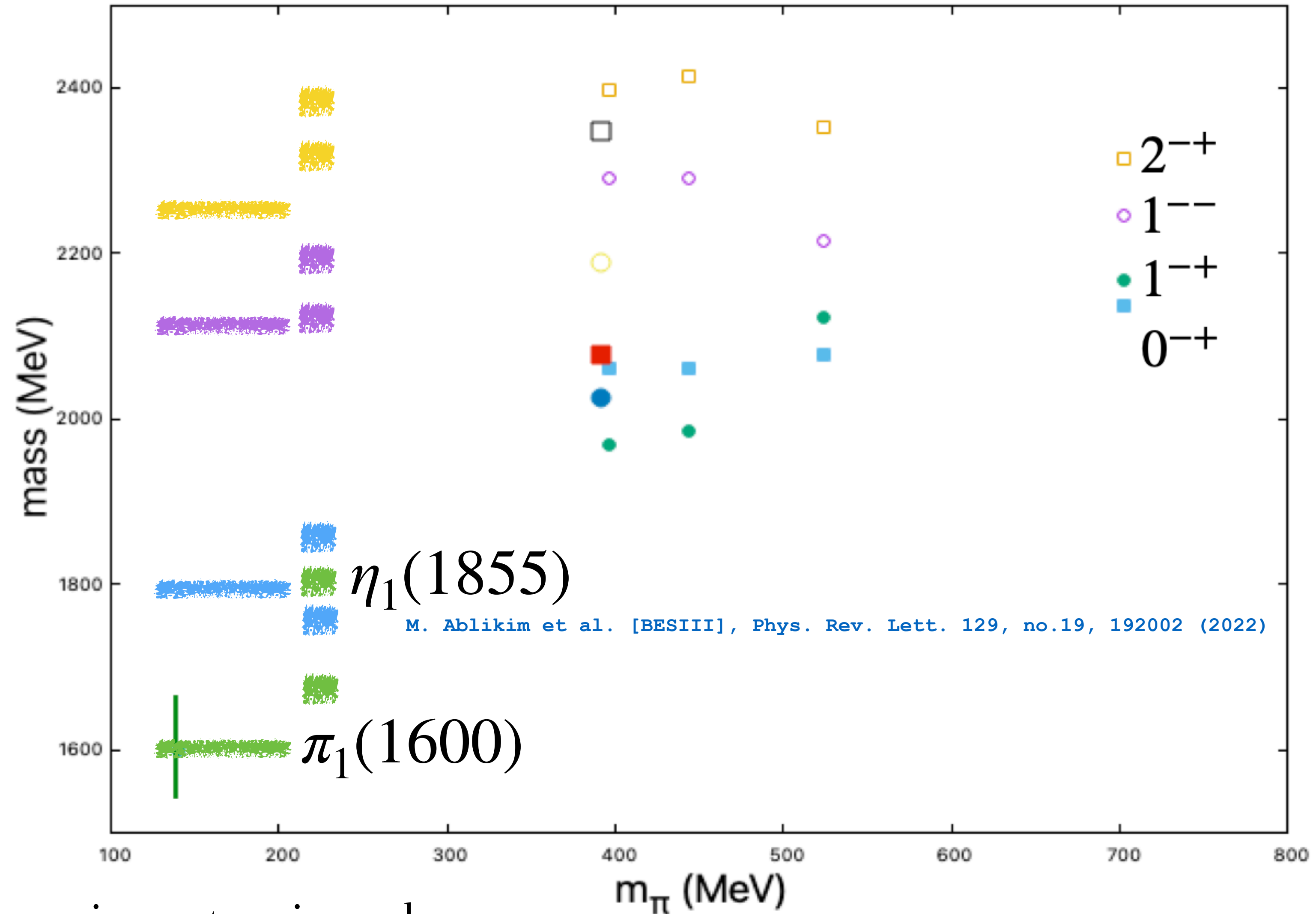
FIG. 11: Isoscalar (green/black) and isovector (blue) meson spectrum on the $m_\pi = 391$ MeV, $24^3 \times 128$ lattice. The vertical height of each box indicates the statistical uncertainty on the mass determination. States outlined in orange are the lowest-lying states having dominant overlap with operators featuring a chromomagnetic construction – their interpretation as the lightest hybrid meson supermultiplet will be discussed later.

lattice light hybrid spectrum



isovector isoscalars

lattice light hybrid spectrum



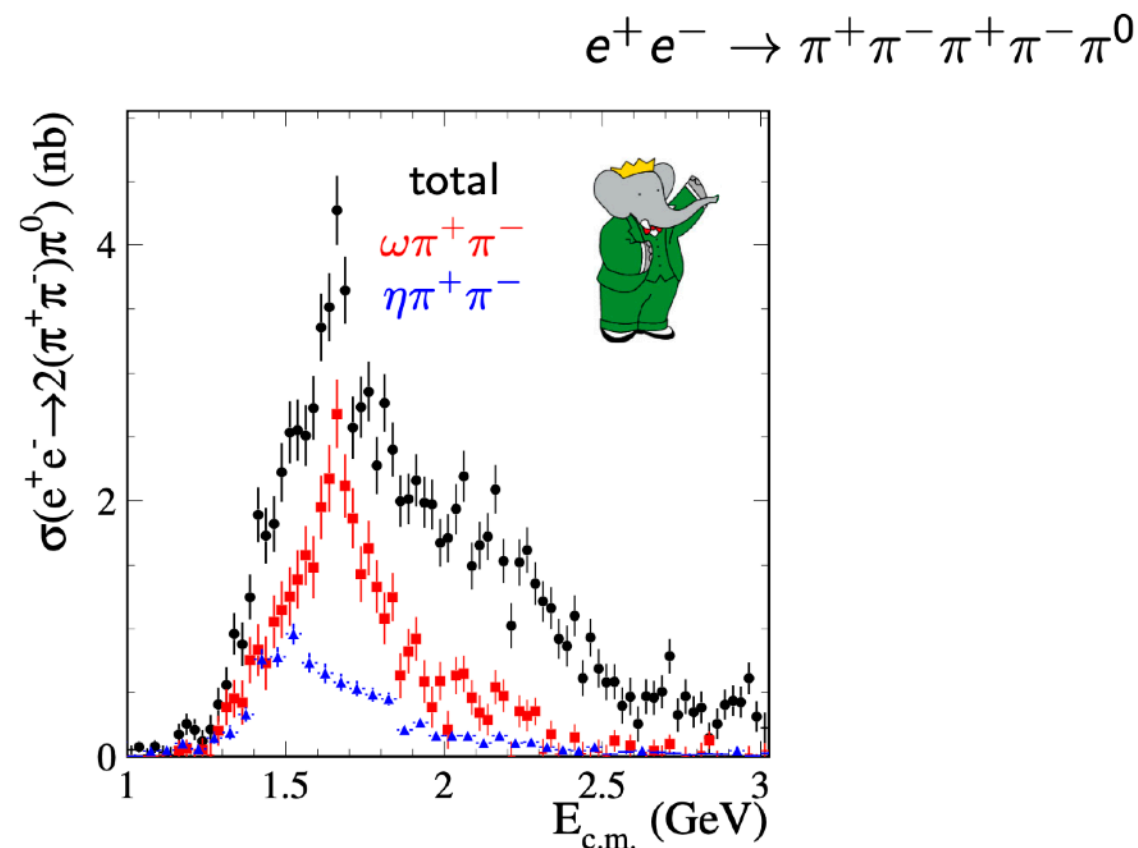
isovector isoscalars

M. Ablikim et al. [BESIII], Phys. Rev. Lett. 129, no.19, 192002 (2022)

$H_1(1^{--})$ 2100/2100/2200

State	Ref.	Mass	Width	Model (iv)	Mass	Model (v)	Mass
$\rho(770)$	PDG	775.2 ± 0.2	147.4 ± 0.8	1^3S_1	720	1^3S_1	730
$\rho(1450)$	PDG	1465 ± 25	400 ± 60	2^3S_1	1440	2^3S_1	1390
$\rho(1570)$	PDG	1570 ± 70	144 ± 90	1^3D_1	1510	1^3D_1	1480
$\rho(1700)$	PDG	1720 ± 20	250 ± 100	$H_1(1^{--})$	1760	3^3S_1	1780
						2^3D_1	1845
$\rho(1900)$	[33]	1900 ± 30	50 ± 30	3^3S_1	1850	4^3S_1	2090
$\rho(2150)$	[34]	2034 ± 16	234 ± 39	2^3D_1	1910	$H_1(1^{--})$	2100
				4^3S_1	2170	3^3D_1	2140
				3^3D_1	2220		

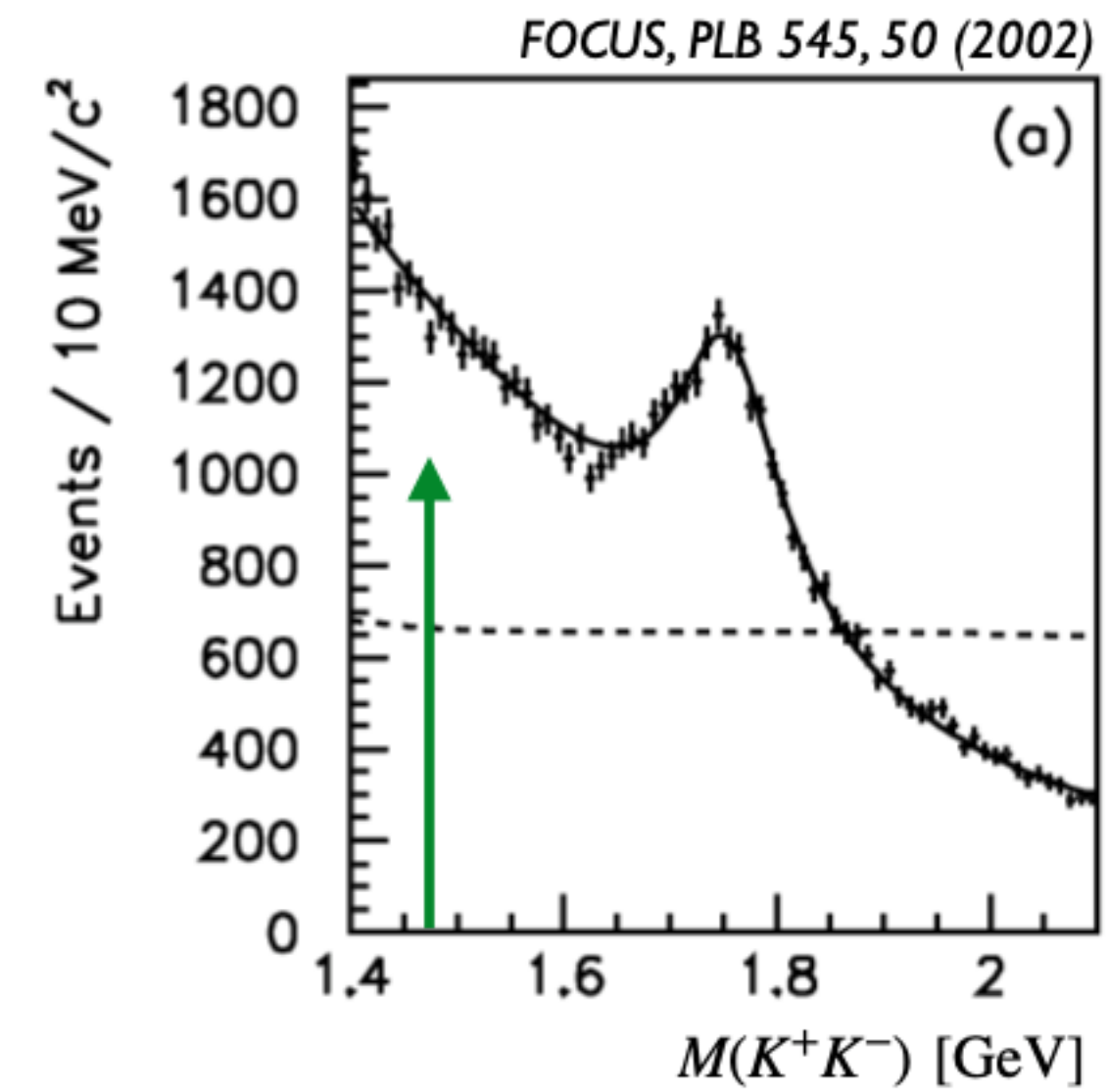
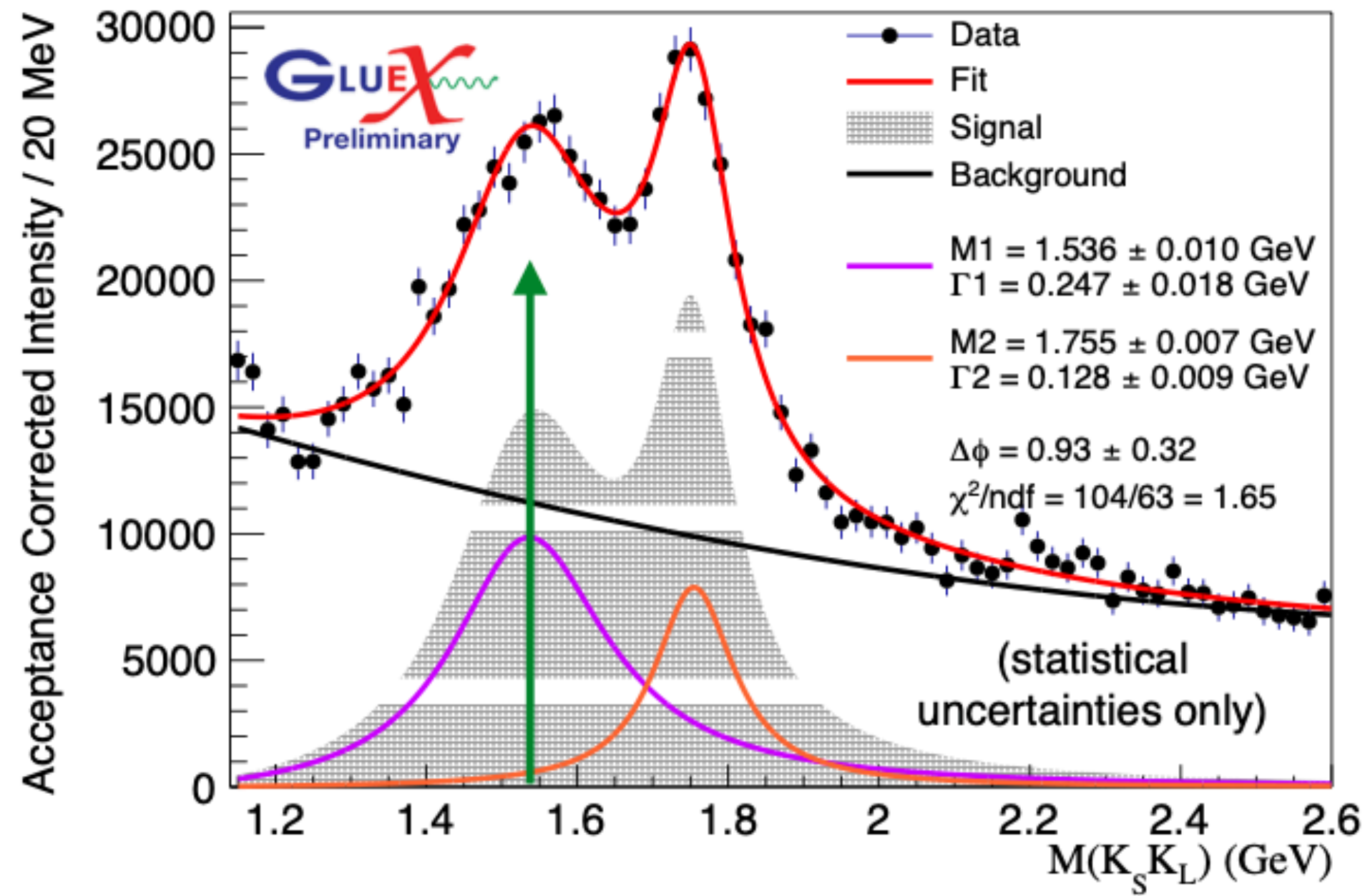
both scenarios find support in LGT and QMs



Model	$m(m_s)$ (MeV)	a_s	σ (GeV ²)	\mathcal{C} (MeV)	α_H	b_0 (GeV ⁻¹)	r_0 (GeV ⁻¹)	ϵ	Rel error	Avg deviation (MeV)
i. [hyp]	335	0.59	0.16	-697	0	9%	94
ii. [SI]	300	1.52	0.071	110	0	7%	66
iii. [SD]	400	1.8	0.06	230	1.3	0.60	4.7	0	5%	54
iv. [SD]	330	2.1	0.055	385	1.3	0.53	5.2	0	5%	54
v. [30]	375 (525)	1.53	0.059	168	1.2	0.56	7.5	0.25	7%	79
Rel/SI	200	0.59	0.14	-246	0	6%	59
Rel/SD	400	0.72	0.14	-359	1.1	0.20	4.4	0	5%	55

$H_1(1^{--})$ 2100/2100/2200

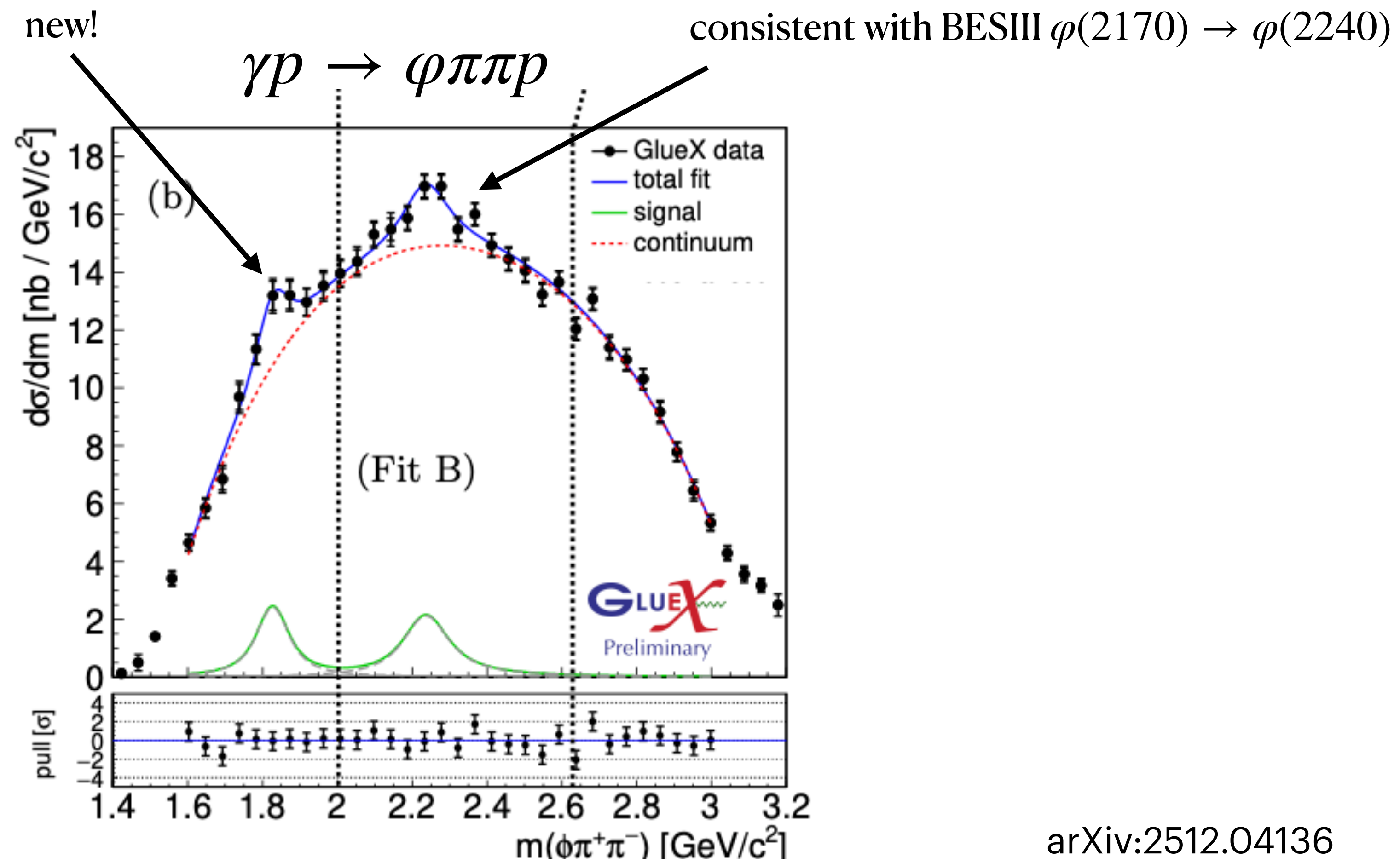
$$\gamma p \rightarrow K_S K_L p$$



Second peak is consistent with FOCUS X(1750). Where is the first peak!?

$H_1(1^{--})$ 2100/2100/2200

Seek BaBar/Belle $\phi(2170)$



arXiv:2512.04136

$H_1(0^{-+})$ 1800/1780/1860

isoscalars

source	π masses				η masses					
RPP[29]	139	1300	1800	–	548	958	1294	1409	1475	2221
LGT[11]	391	1520	2100 ^a	2350	600	1000	1680	1750	2100 ^a	2300 ^a
GI[38]	150	1300	1880		520	960	1440	1630		
fit A	560	1365	1720	1990	560	730	1365	1580	1990	2200
fit A'	600	1390	1750	2030	600	820	1390	1580	1750	1940

0^{-+} (1750 – 1850)			
	$f_0(500)\eta$	$33.3 \cos^2 \theta$	$30.2 \sin^2 \theta$
	$K^*\bar{K}$	$(\cos \theta \sqrt{0.8} - \sin \theta \sqrt{2})^2$	$(\sin \theta \sqrt{1.1} + \cos \theta \sqrt{2.7})^2$
	$a_2(1320)\pi$	$1.0 \cos^2 \theta$	$2.5 \sin^2 \theta$
	$f_0(980)\eta$	$42.8 \cos^2 \theta$	$44.9 \sin^2 \theta$
	$a_0(1450)\pi$	$34.5 \cos^2 \theta$	$54.6 \sin^2 \theta$
	$\phi\eta$	x	x
	$f_2(1270)\eta$	\emptyset	x
	$\phi\eta'$	\emptyset	\emptyset
	$K^*(1410)\bar{K}$	\emptyset	\emptyset
	$f_0(1370)\eta$	\emptyset	\emptyset
	$K_0^*(1430)\bar{K}$	\emptyset	\emptyset
	$K_2^*(1430)\bar{K}$	\emptyset	\emptyset
	$f_2'(1525)\eta$	\emptyset	\emptyset
Γ_{tot}		$112 \cos^2(\theta) - 2.5 \cos(\theta) \sin(\theta) + 2 \sin^2(\theta)$	$2.7 \cos^2(\theta) + 3.4 \cos(\theta) \sin(\theta) + 133 \sin^2(\theta)$

^agluonic content

not a good agreement!

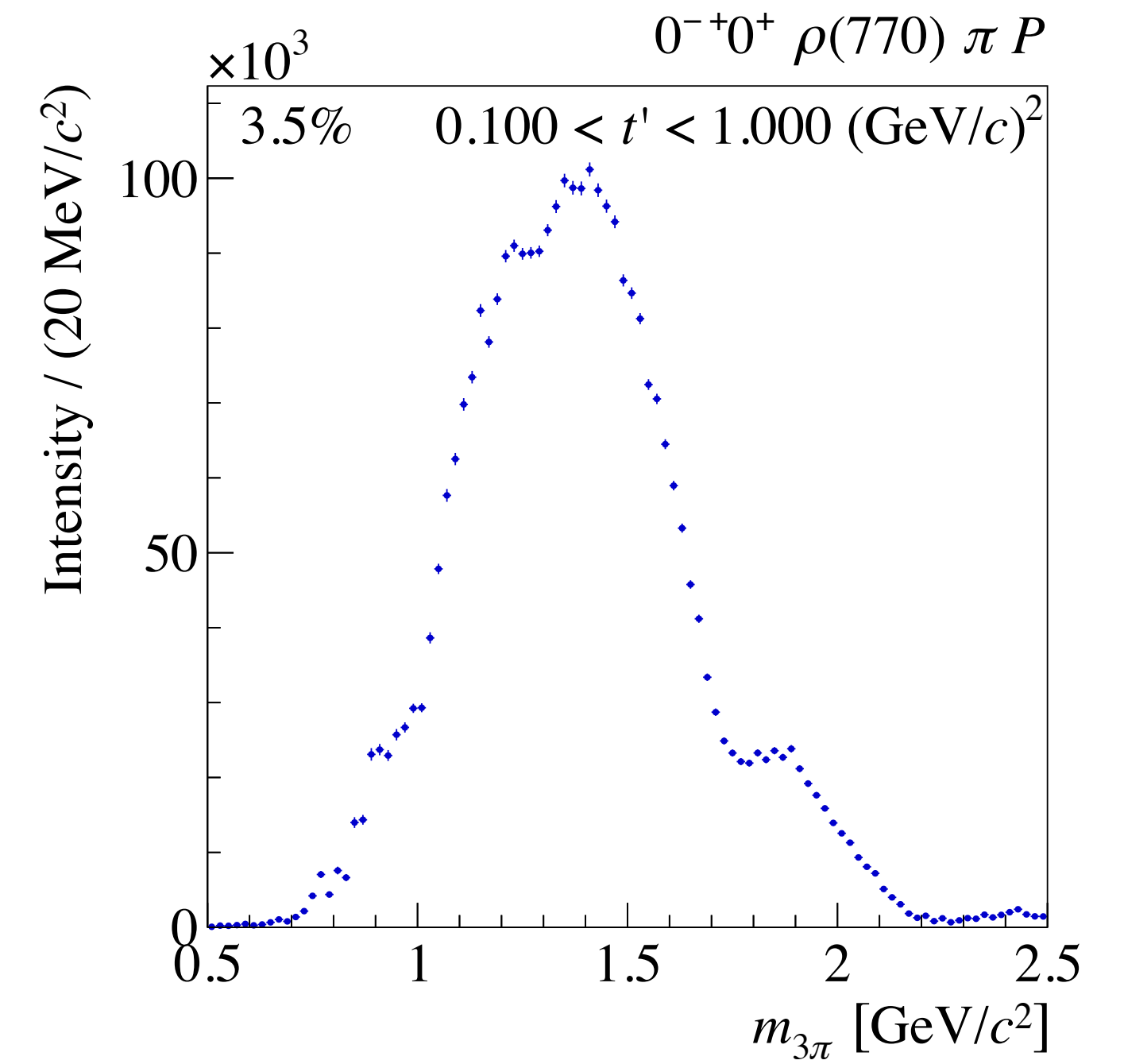
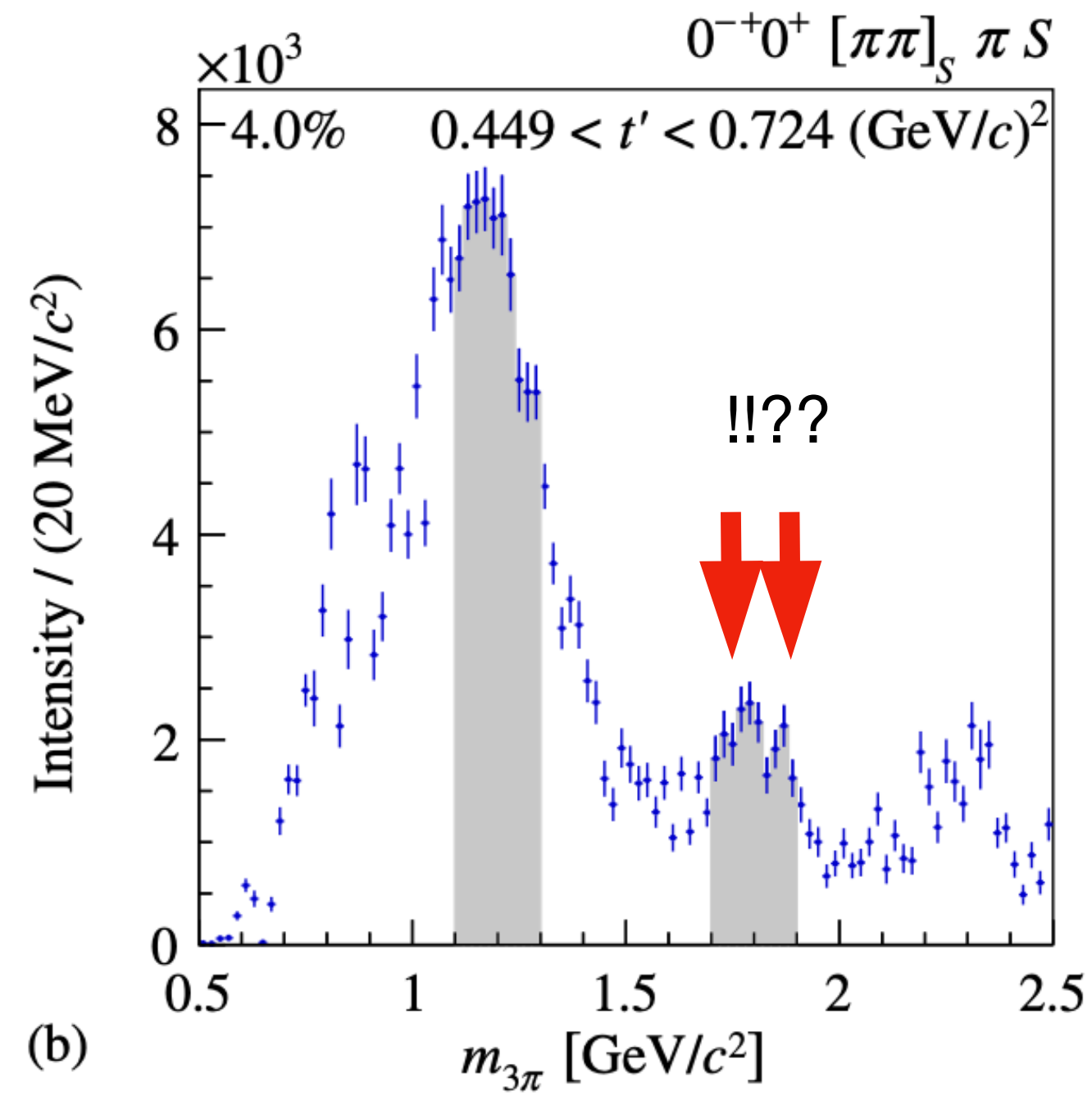
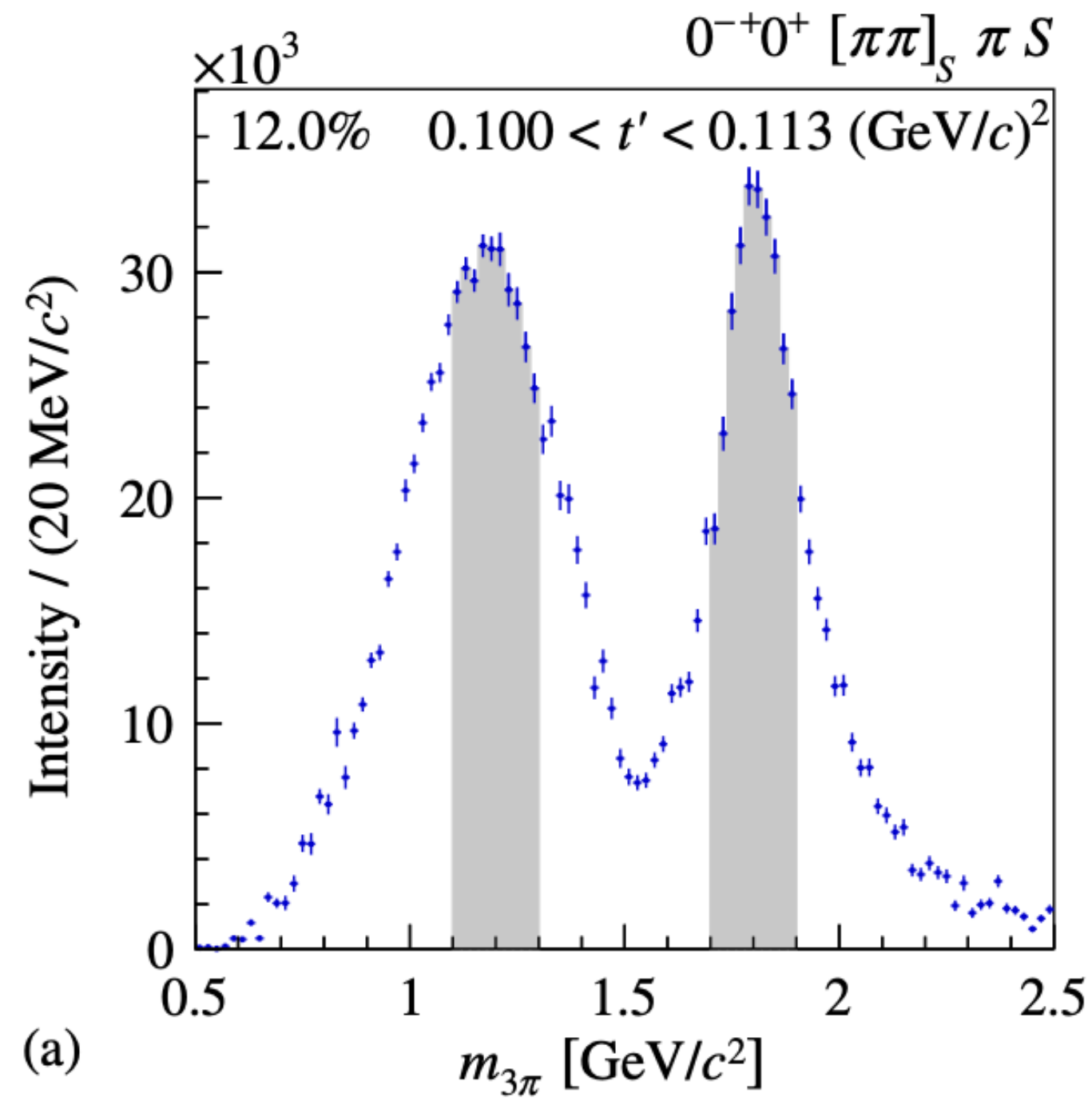
state	$f_0(500)\pi$	$\rho\pi$	$f_0(980)\pi$	$K^*\bar{K}$	$b_1(1235)\pi$	$f_2(1270)\pi$	$f_0(1370)\pi$	$\omega\rho$	Γ_{tot}	Γ_{PDG}
$\pi(2S)$	1	230	2	\emptyset	–	\emptyset	\emptyset	\emptyset	233	200-600
$\pi(3S; 1850)$	12	117	x	64	–	30	1	109	333+	215
$H_1(\pi)$	51	9	73	0	57	x	22	x	212	–

$$\Gamma_{s\bar{s}g} \sim 10s \text{ MeV}$$

ratios of partial widths are a powerful discriminant

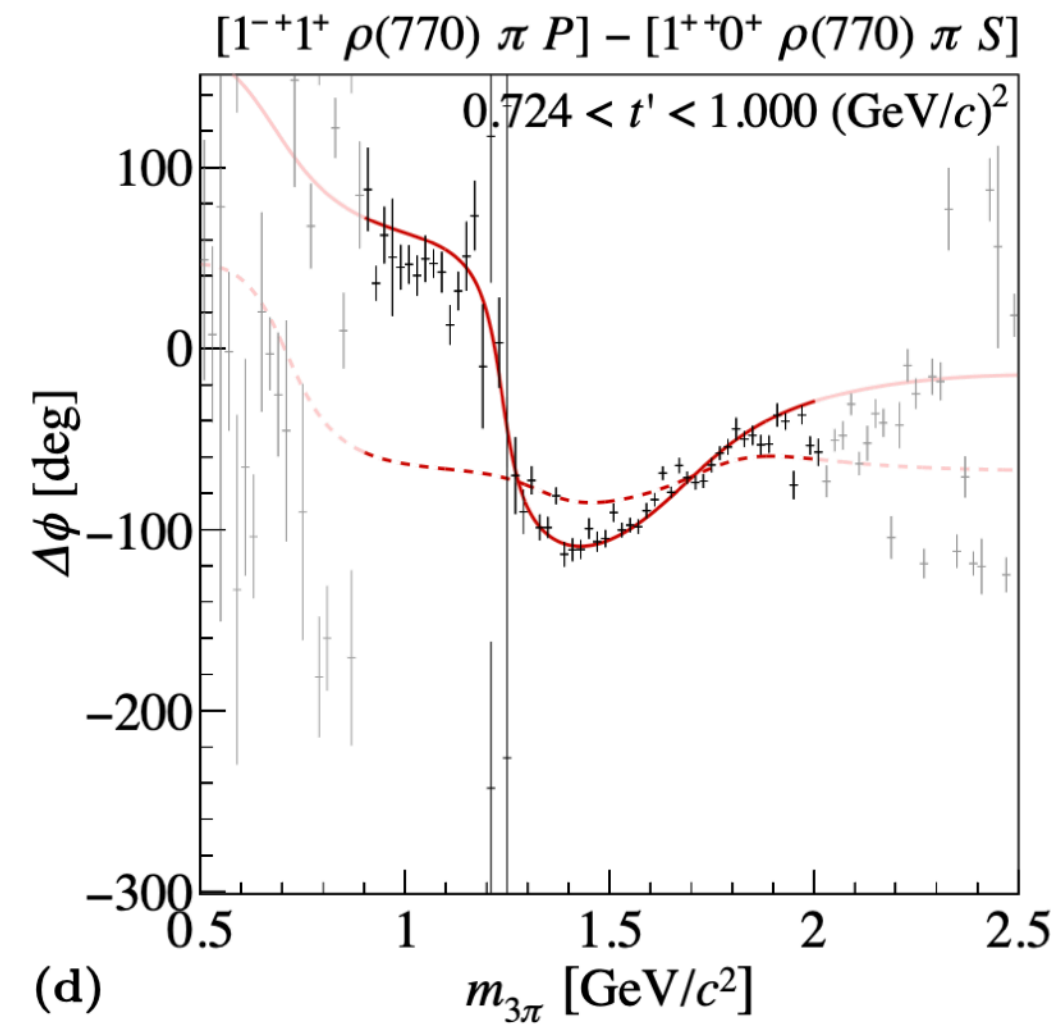
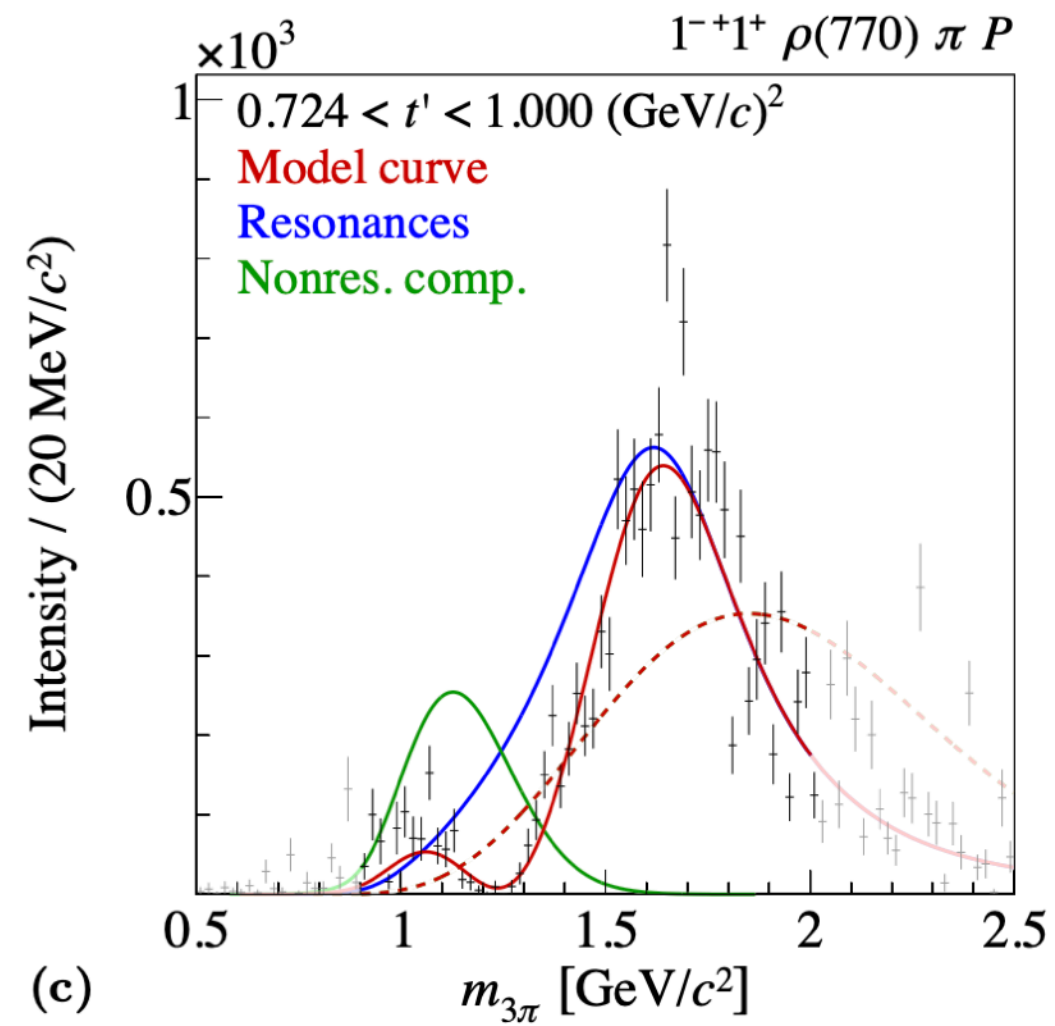


$H_1(0^{-+})$ 1800/1780/1860



$H_1(1^{-+})$ 1600/1680/1820

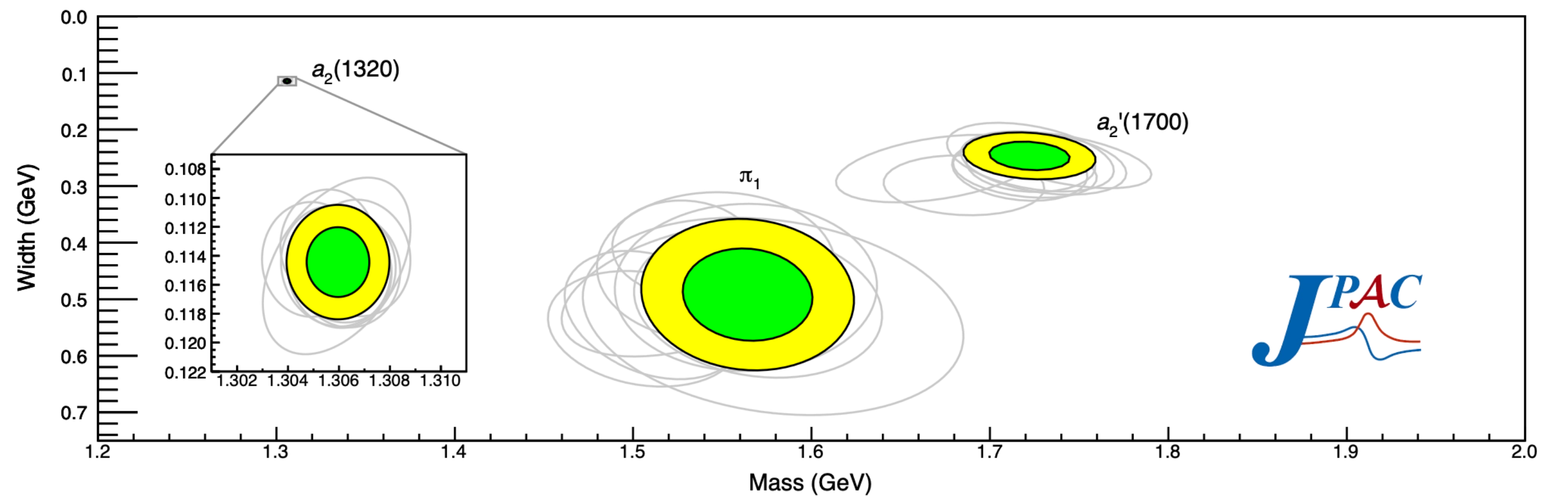
COMPASS reanalysis



JPAC reanalysis

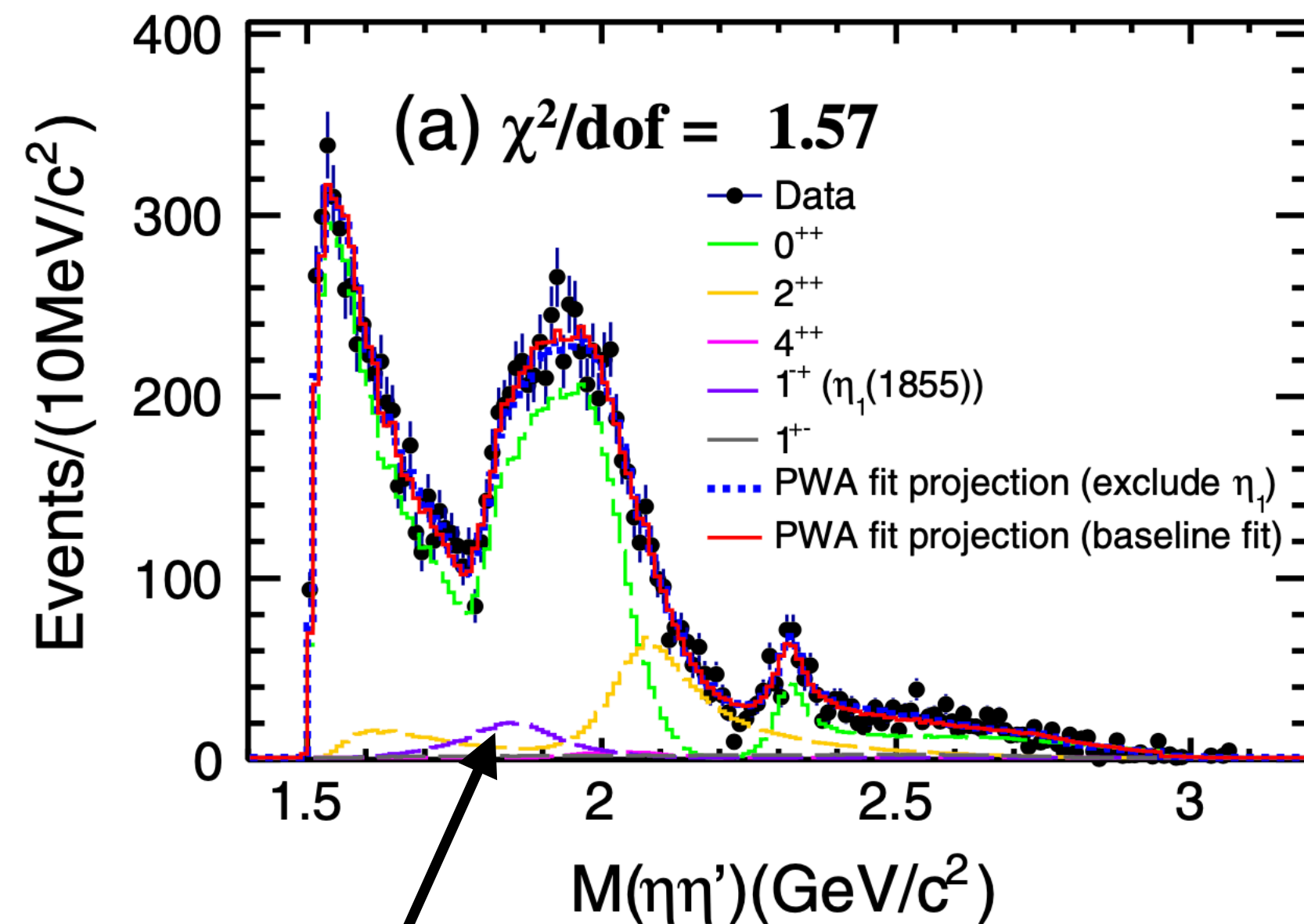
A. Rodas et al. [JPAC],
Determination of the pole position of the lightest hybrid meson candidate
 Phys. Rev. Lett. **122**, 042002 (2019).

COMPASS 1802.05913; 2108.01744



$H_1(1^{-+})$ 1600/1680/1820

BESIII discover a possible partner state!



$\eta_1(1855)$

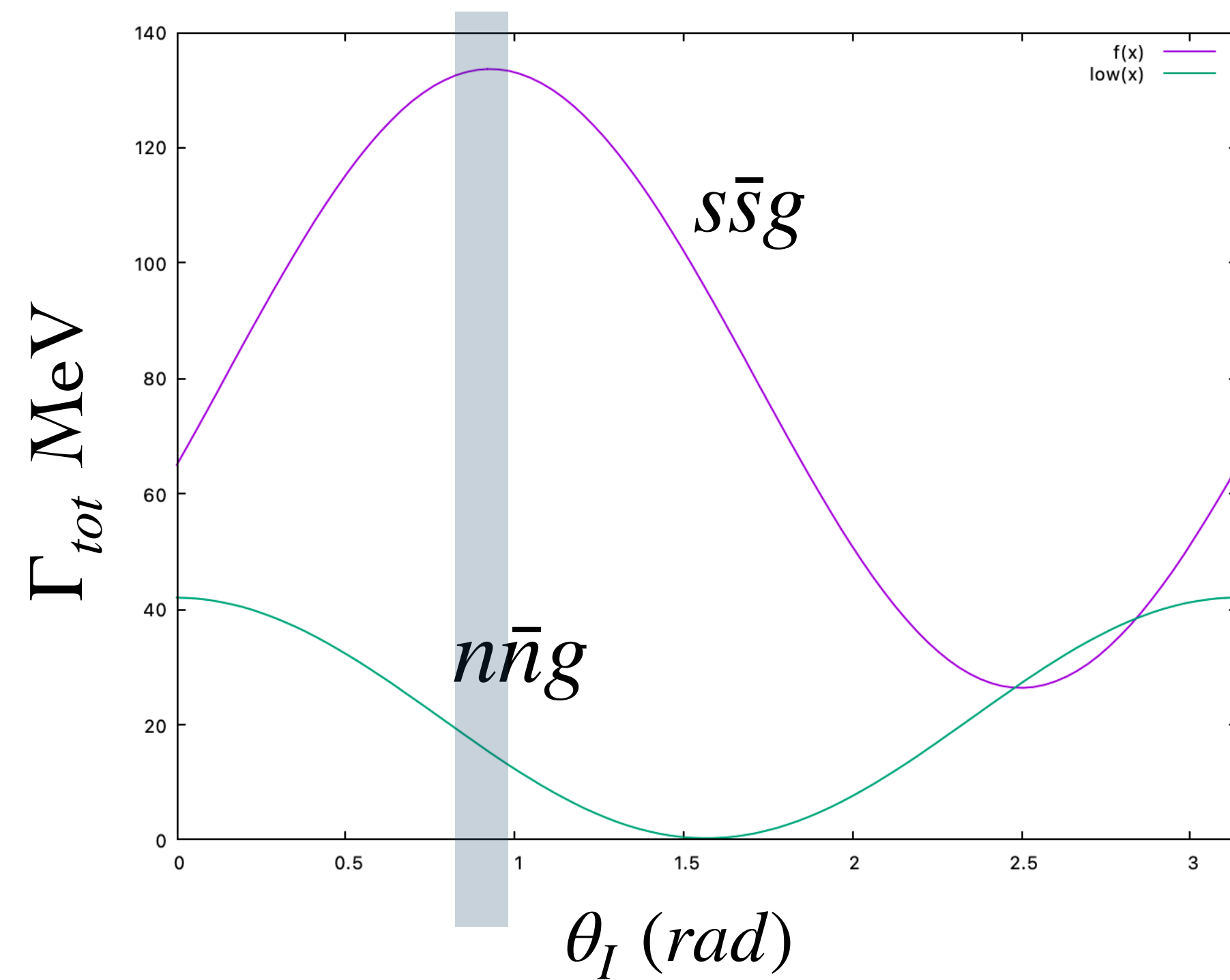
$$M = 1855 \pm 9_{-1}^{+6} \text{ MeV}$$

$$\Gamma = 188 \pm 18_{-8}^{+3} \text{ MeV}$$

BESIII, PRL 129, 192002 (2022).

$H_1(1^{-+})$ 1600/1680/1820

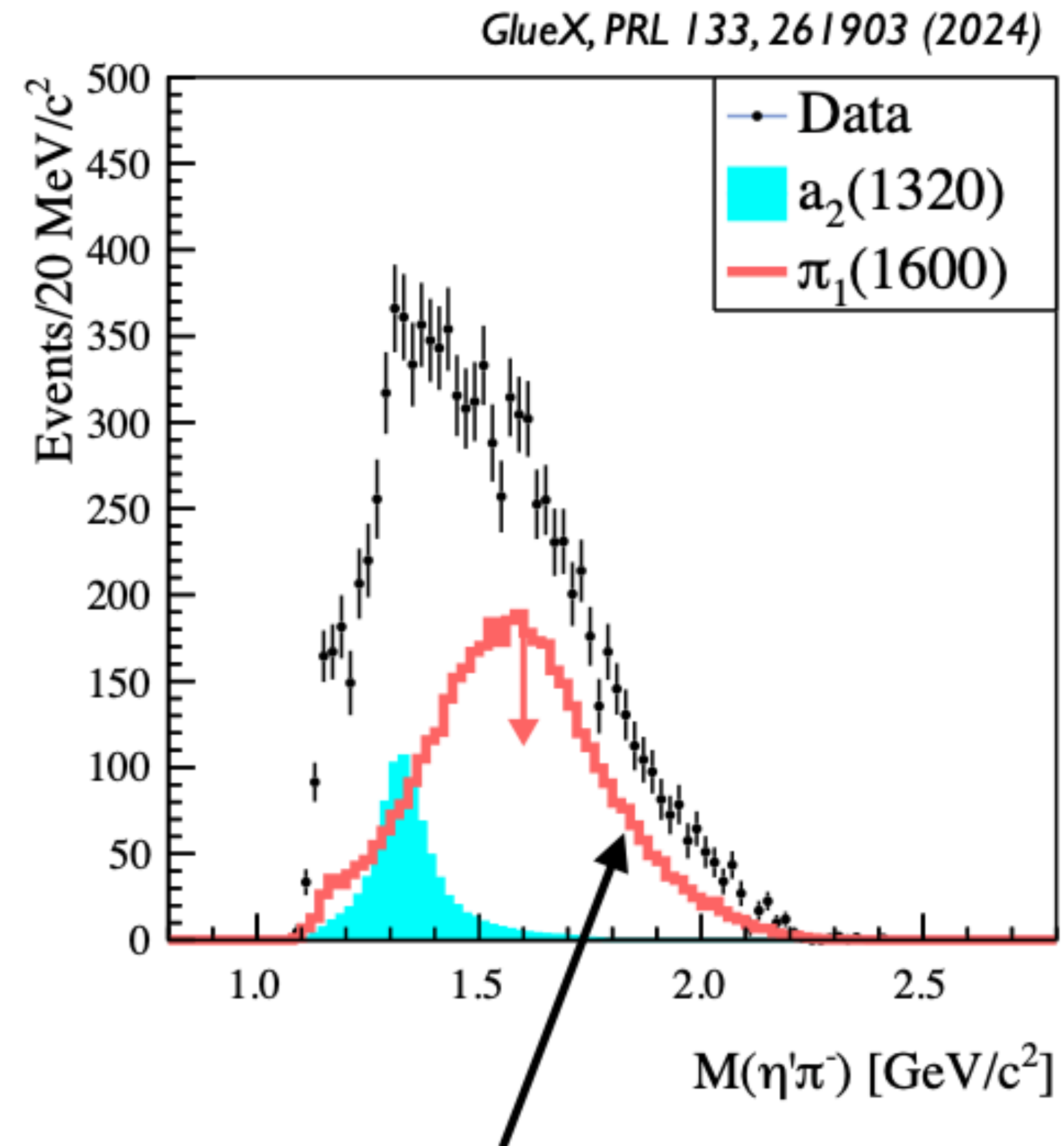
Other partners



With these observations one expects a light isoscalar exotic with mass near 1680 MeV with a total width of approximately 30 MeV and dominant decays to $a_1\pi$ and $K_1\bar{K}$.

$H_1(1^{-+})$ 1600/1680/1820

GlueX search reports an upper limit
(uses HadSpec decay calculation)



Upper limit using LQCD
[HadSpec, PRD 103, 054502 (2021)]
and GlueX data to constrain $\pi_1 \rightarrow b_1\pi$

$H_1(2^{-+})$ 2250/2300/2400

$$\Gamma_{tot}(H_1(2^{-+}); I = 1) \approx 80 \text{ MeV}$$

$$\theta_I \approx 10(5) \text{ deg}$$

$$\Gamma_{tot}(H_1(2^{-+}); I = 0) \approx 75 \text{ MeV}$$

$$\Gamma_{tot}(H_1(2^{-+}); I = 0) \approx 60 \text{ MeV}$$

source	π_2 masses				η_2 masses	
RPP[29]	1670	1874	1963 ^a	2090 ^a	1617	1842
LGT[11]	1900	2350 ^b	2550	–	1900	2000
GI[38]	1680	2130	–	–	1680	1890
fit A	1580	1880	2130	2350	1580	1785
fit A'	1590	1865	2100	2300	1590	1820

^anot established

^bgluonic content

2 ⁻⁺ (2360)	$f_0(500)\pi$	0.5			
	$\rho\pi$	3.7			
	$K^*\bar{K}$	0.8			
	$f_2(1270)\pi$	19	(18,43)	(20,15)	(32,34,36)
	$f_1(1285)\pi$	0.6			
	$f_0(980)\pi$	0.4			
	$f_0(1370)\pi$	x			
	$\rho(1450)\pi$	7.2			
	$K_{1L}\bar{K}$	x			
	$K_{1H}\bar{K}$	0.1			
	$a_1(1260)\eta$	0.2			
	$a_2(1320)\eta$	12	(10,23)	(13,11)	(15,16,16)
	$K^*(1410)\bar{K}$	0.3			
	$K_0^*(1430)\bar{K}$	x			
	$K_2^*(1430)\bar{K}$	39	(33,66)	(37,39)	(32,34,36)
$a_0(1450)\eta$	x				
Γ_{tot}	83.8	(61,132)	(84,82)	(82,87,89)	

$H_1(2^{-+})$ 2250/2300/2400

state	$\pi\rho$	$\omega\rho$	$f_2\pi$	$K^*\bar{K}$	$\rho(1450)\pi$	$K_2^*\bar{K}$	Γ_{tot}	Γ_{PDG}
$\pi_2(1S)$	117	25	85	20	3	–	250	258(8)
$\pi_2(2S)$	86	24	31	3	19	9	172+	237(30)
$\pi_2(3S; 2250)$	12	10	16	6	35	8	87+	–
$H_1(\pi_2)$	4	0	20	0	7	40	71	–

perhaps $f_2\pi/\rho\pi$ is the best diagnostic

S-D-G amplitude ratios for $\pi_2 \rightarrow f_2\pi$

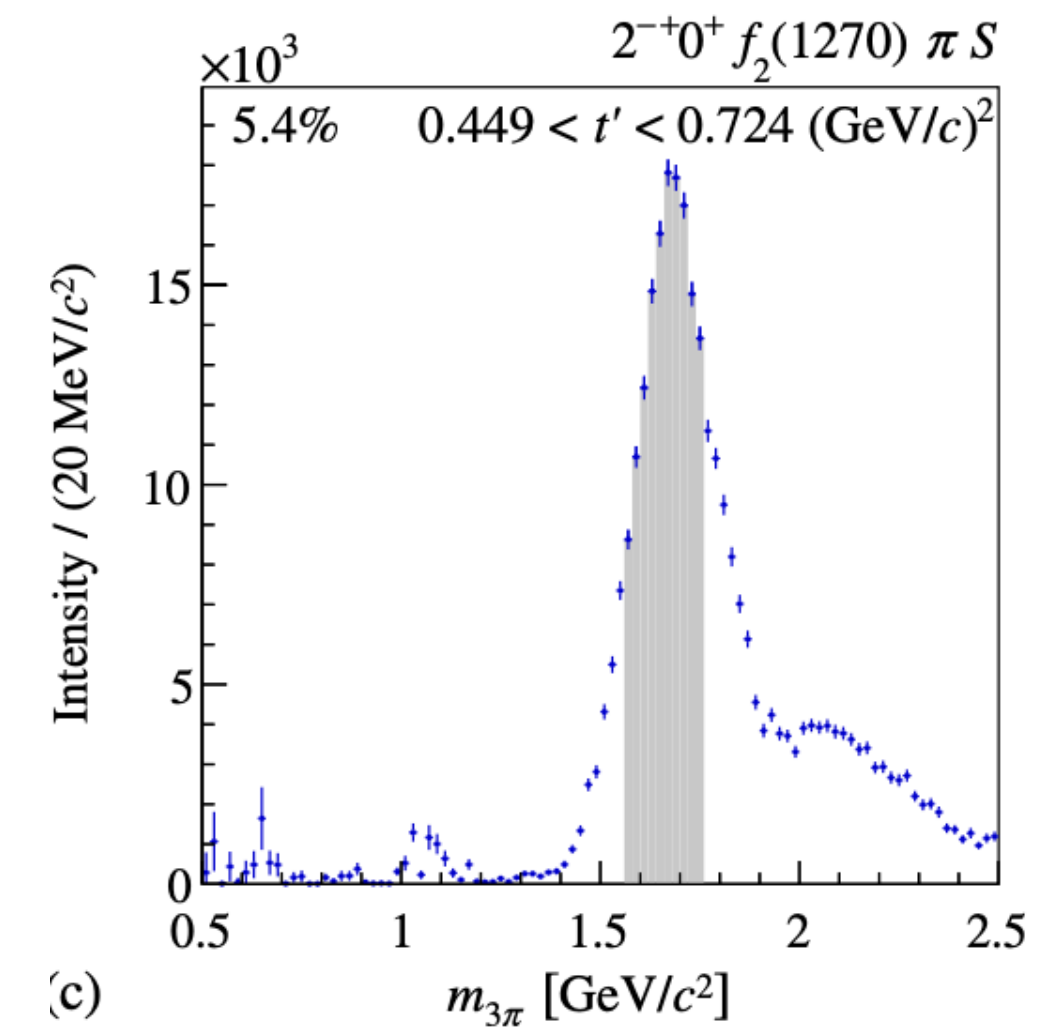
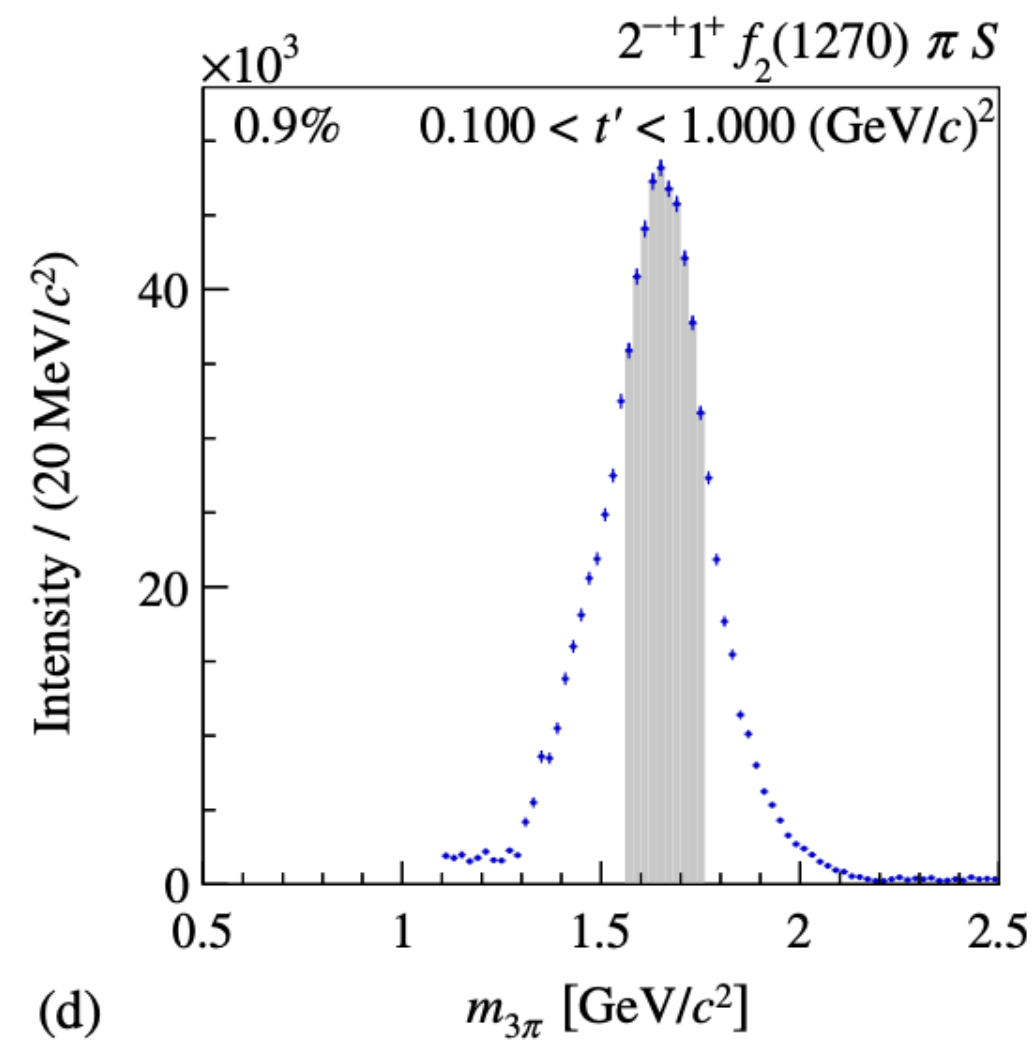
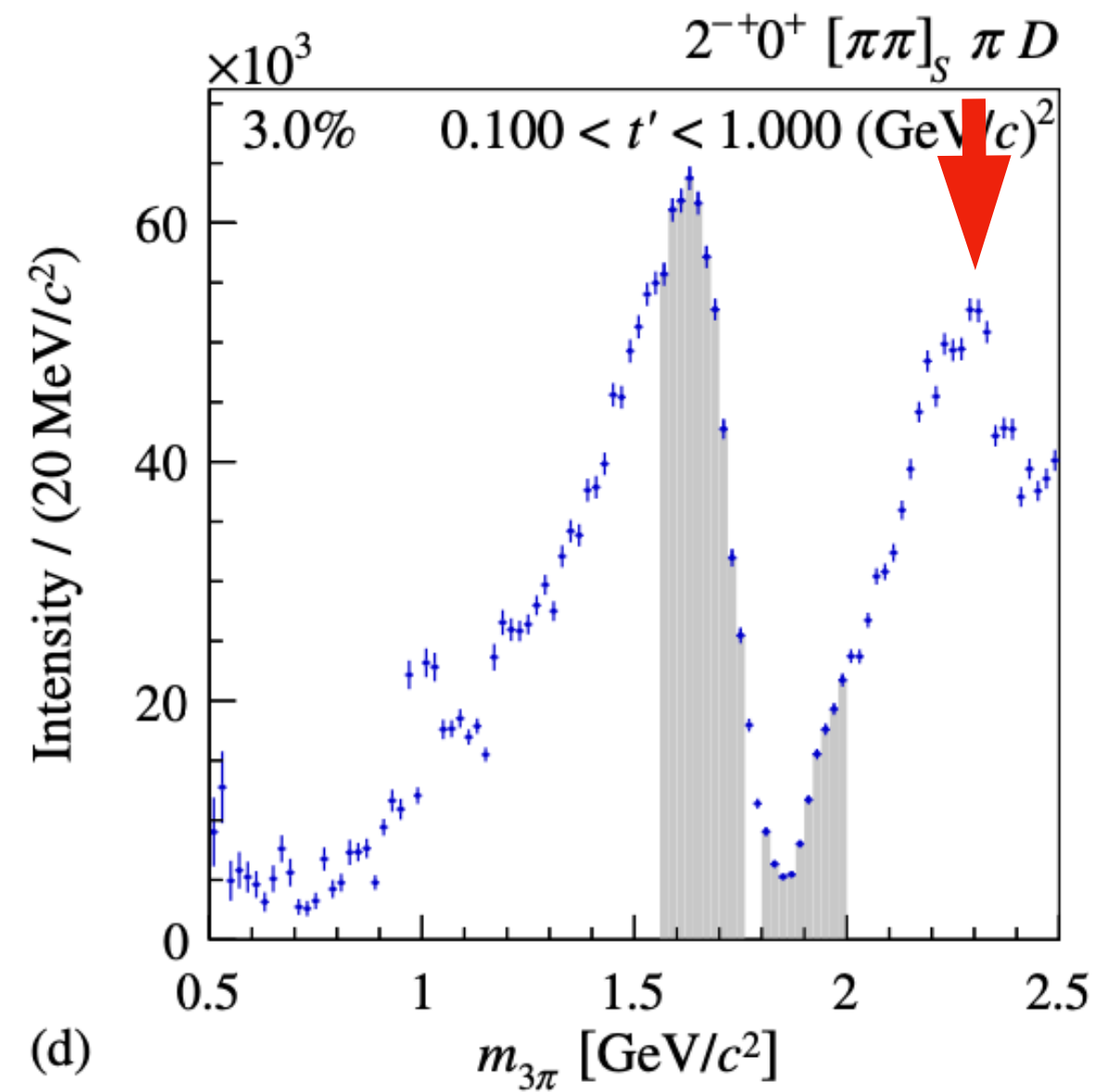
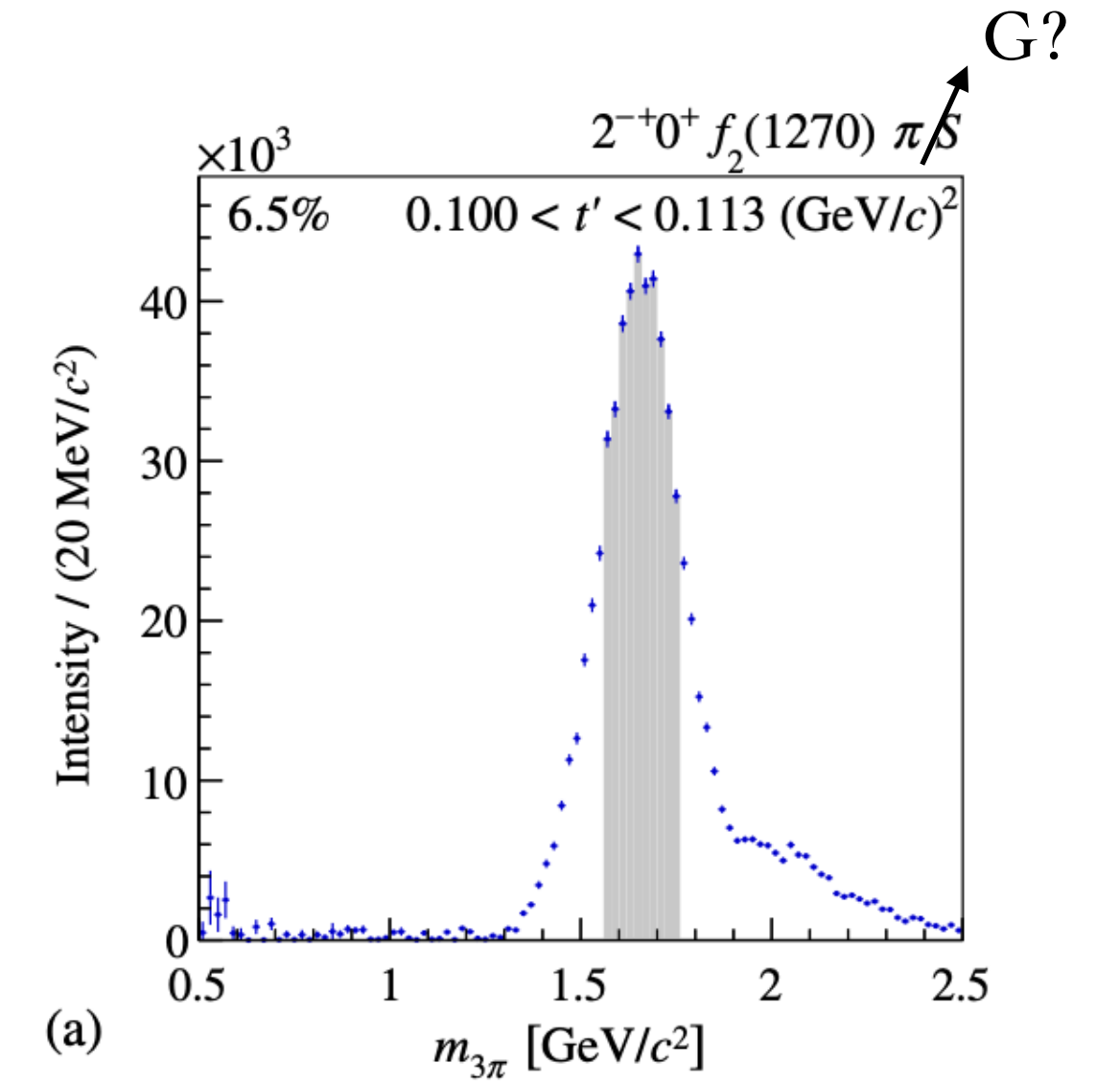
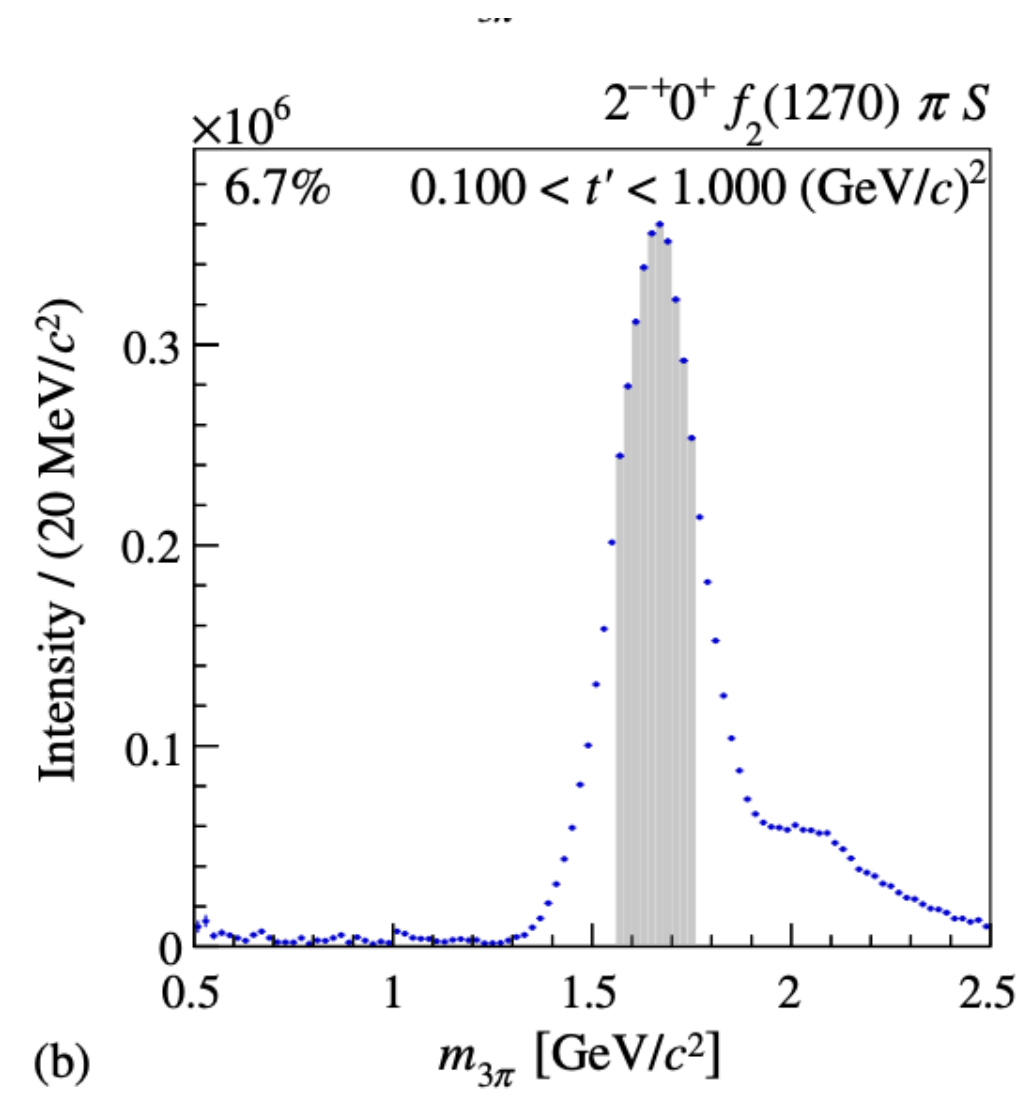
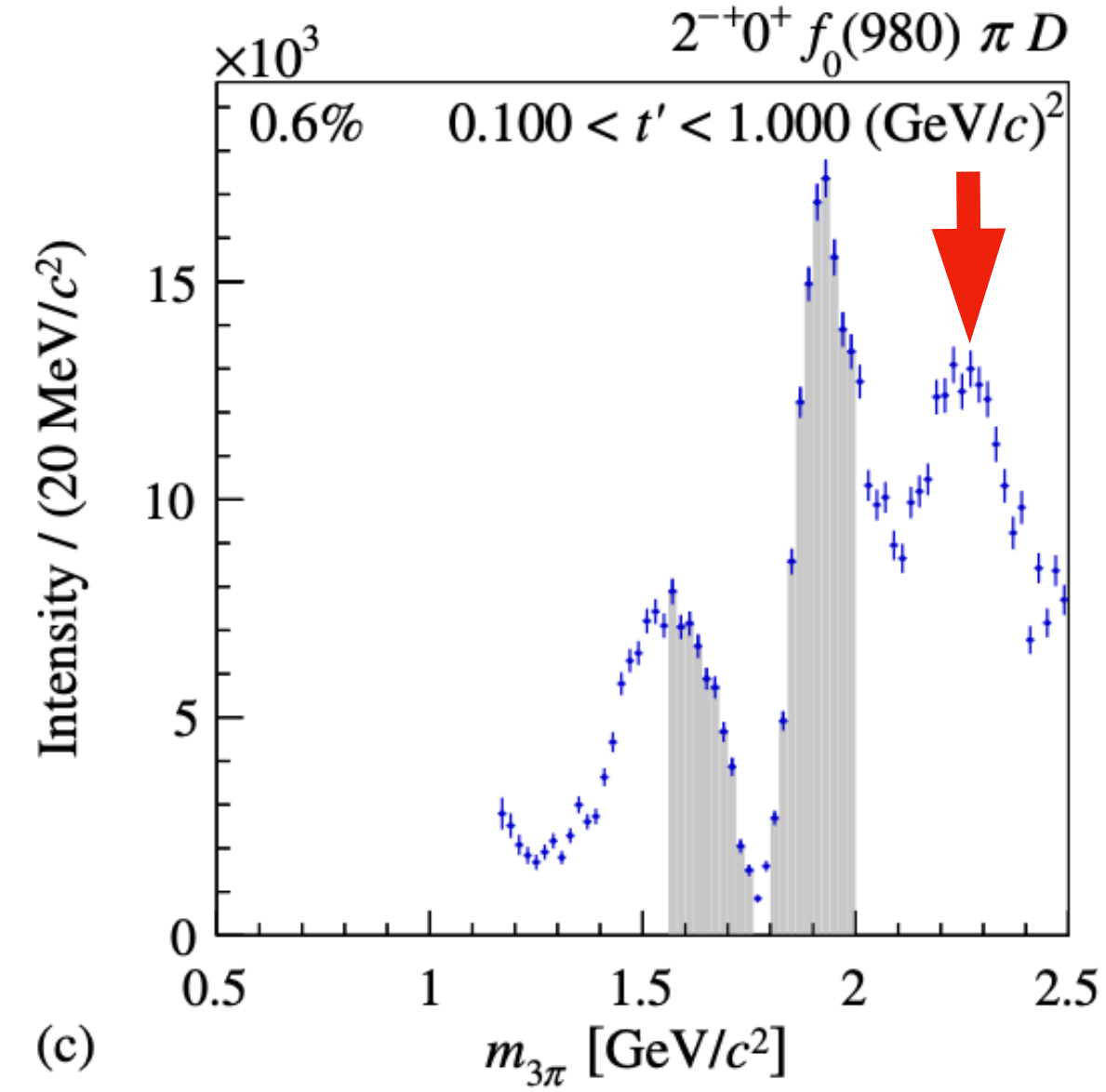
are approximately 1:0:0 for a hybrid but are 1 : 0.14 : 0.003 for the $q\bar{q}(1S)$ state, 1 : 0.62 : 0.06 for the $q\bar{q}(2S)$ state, and -0.24 : 1 : -0.38 for a $q\bar{q}(3S)$ state near 2250 MeV.

COMPASS: $\pi_2(1670) \rightarrow (f_2\pi)_S \rightarrow 2S$

$\pi_s(1880) \rightarrow (f_2\pi)_D \rightarrow 3S$



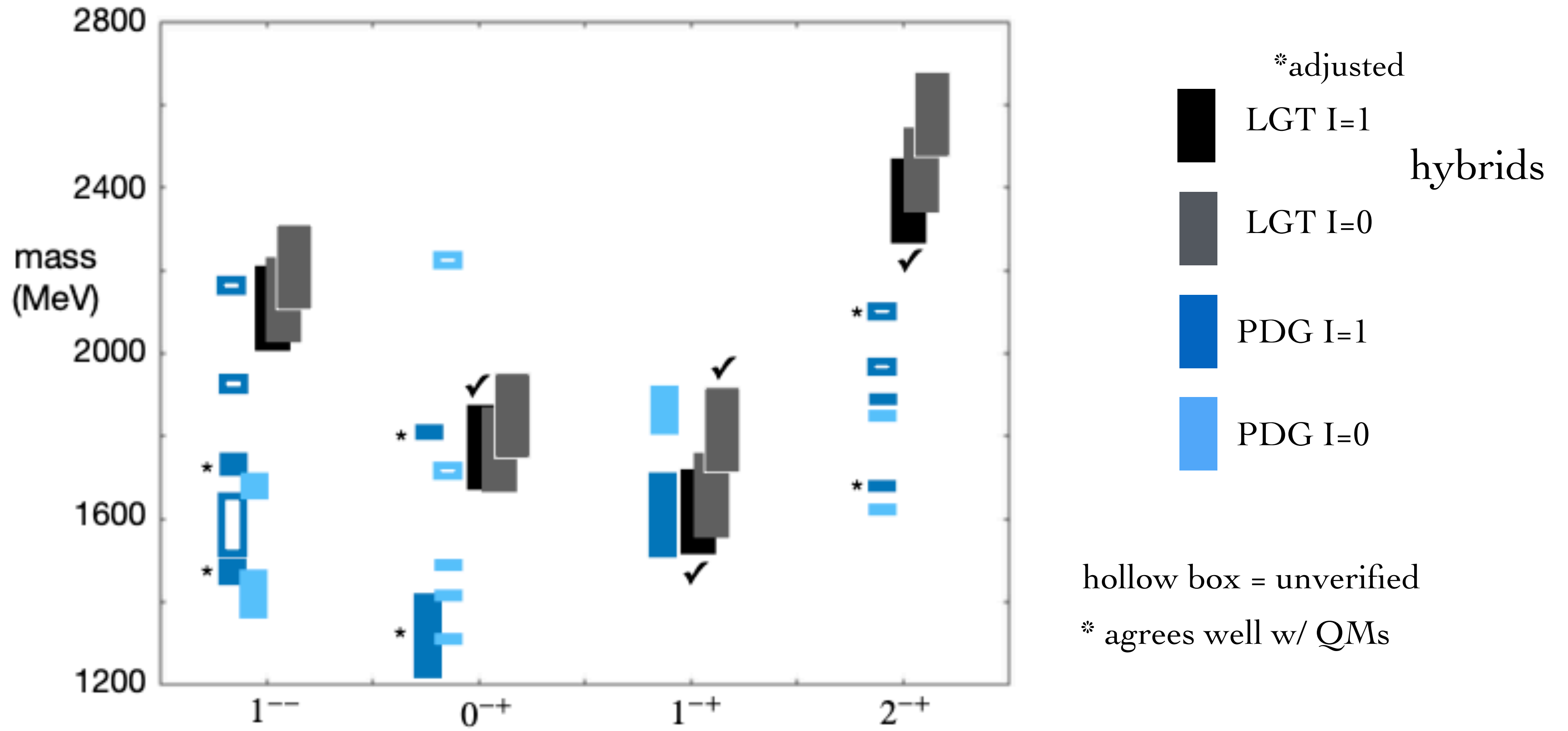
$H_1(2^{-+})$ 2250/2300/2400



summary

- vector hybrid near 2100 w/ decay constant ~ 20 MeV. Likely too broad to be easily seen
- partner states at 2100-2250 and 2220-2350.
- with a $\pi_1(1600)$ expect partner states at 1750-1780 and ~ 1900 (near the $\eta_1(1855)$).
- the $\pi(1800)$ is a good hybrid candidate. The $\pi(3S)$ should be nearby
- it is possible that the hybrid $\pi_2(2360)$ has been observed!
- think about strange hybrids!
- perhaps the low lying hybrid spectrum is emerging (& consistent with expectations)

A Speculative Summary



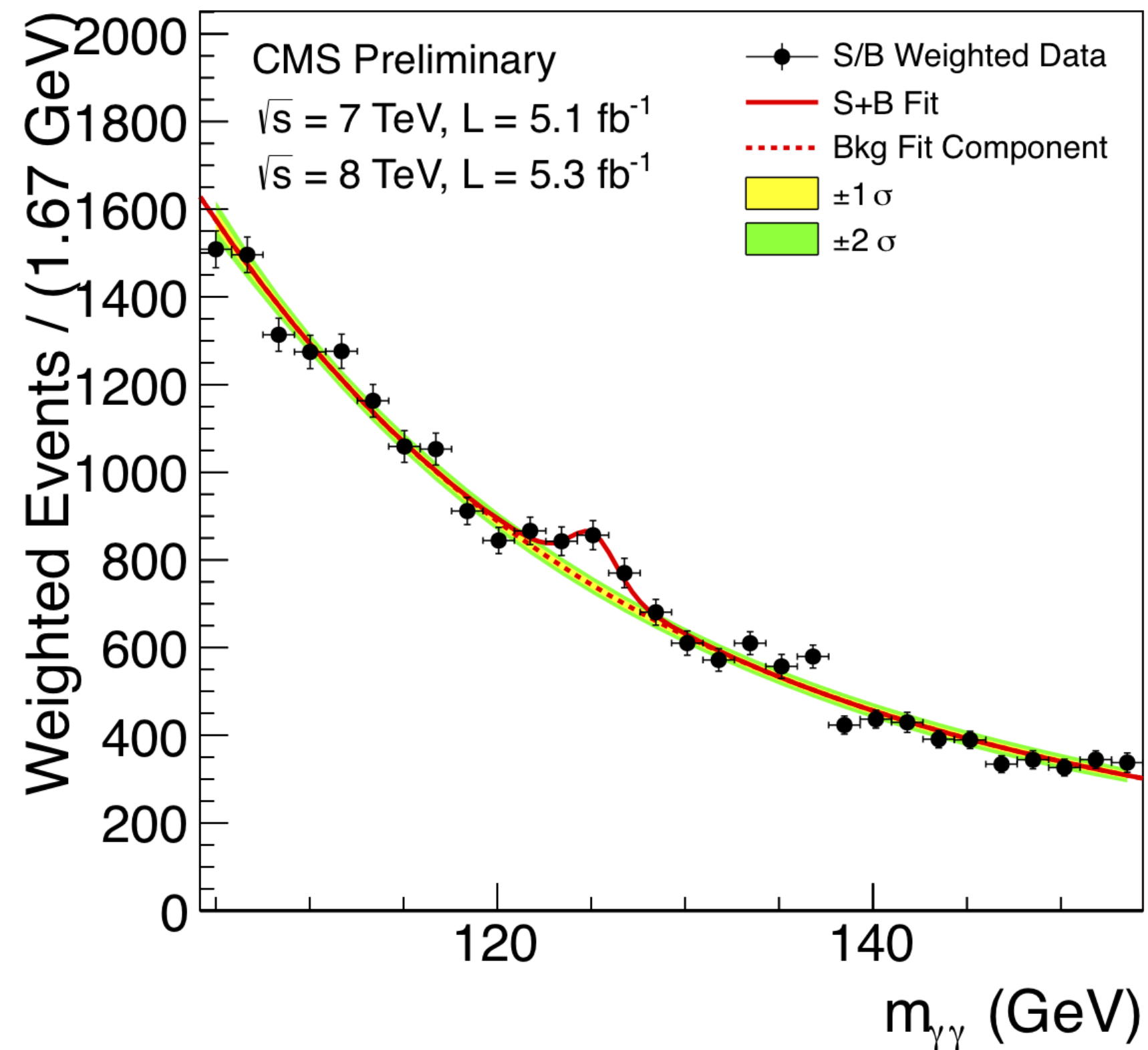
discovering exotic particles

the traditional approach to "discovery"

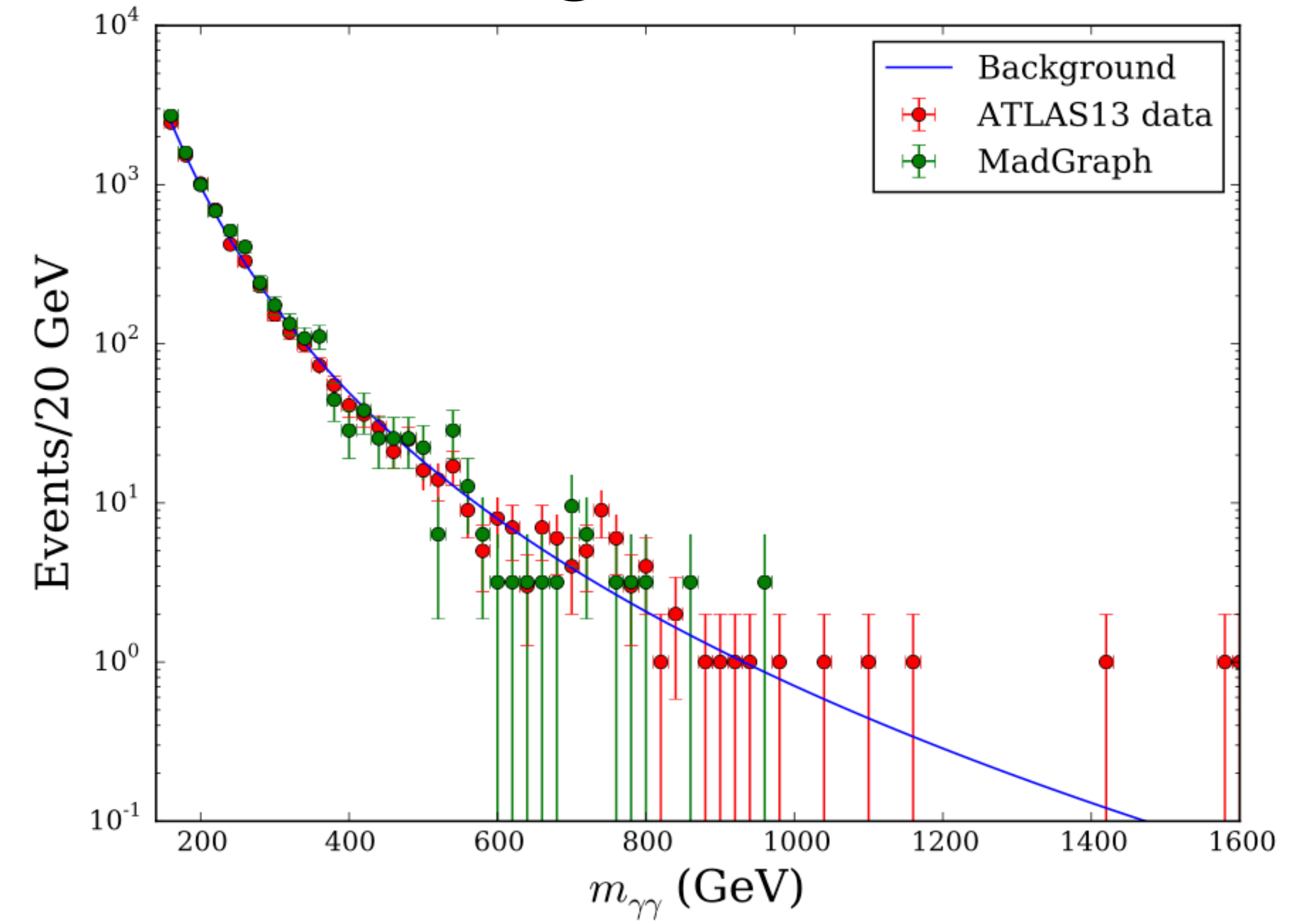
- i. fit two model amplitudes to data, $M_0(\dots)$ and $M_1(\dots; m_R, \Gamma_R)$
- ii. determine the P-value ~ the probability of obtaining the observed effect (or greater) given that the null hypothesis is true.

$$2 \log \frac{L_1}{L_0} \rightarrow \chi_{d_1-d_0}^2, \quad L_0 \subset L_1. \text{ (Wilks's theorem)}$$
- iii. declare a discovery if $P < 3 \cdot 10^{-7} \rightarrow 5\sigma$ [$P < 0.0027 \rightarrow 3\sigma$]. The new physics is described by the fit parameters, m_R, Γ_R .

Higgs



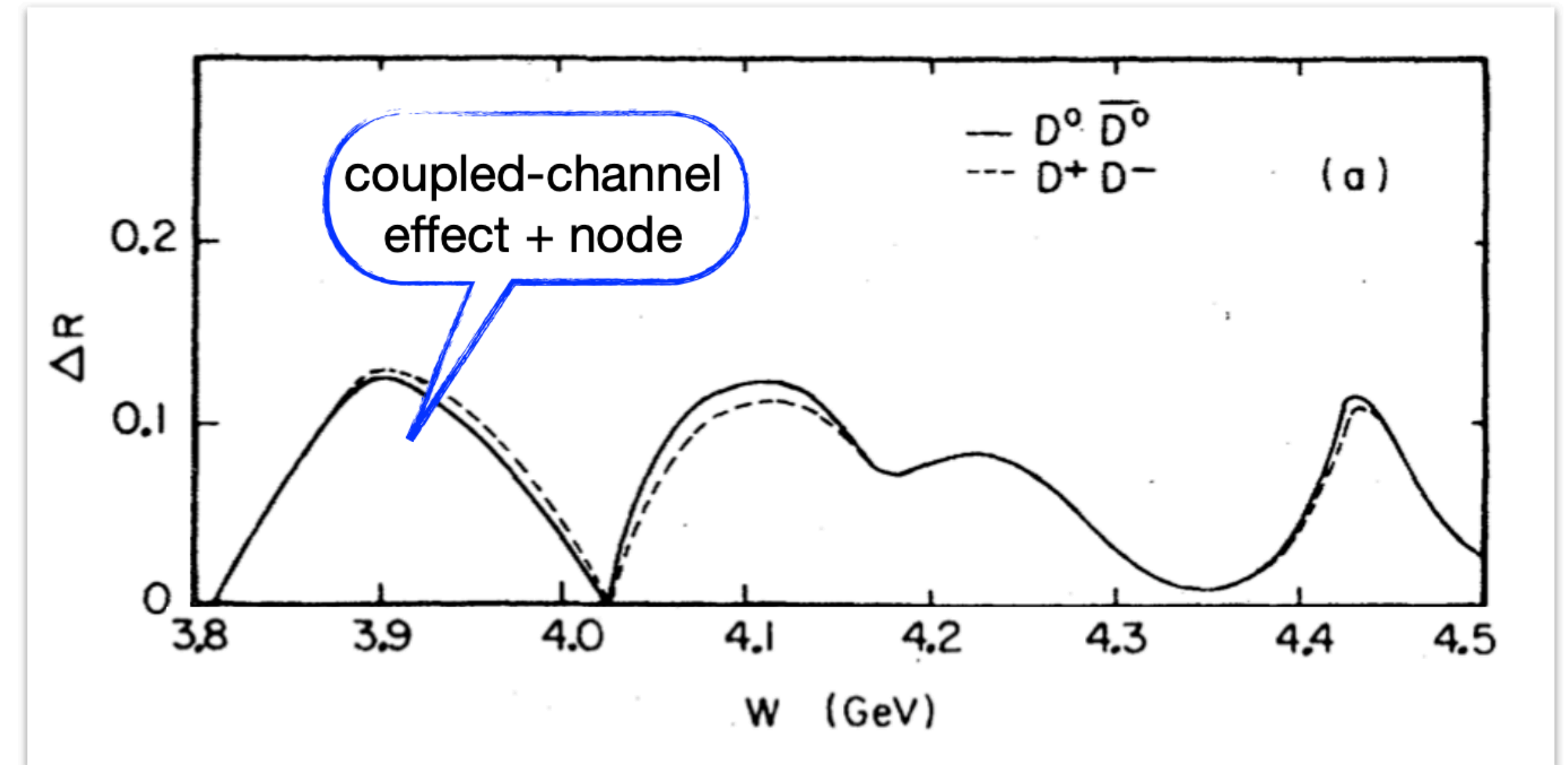
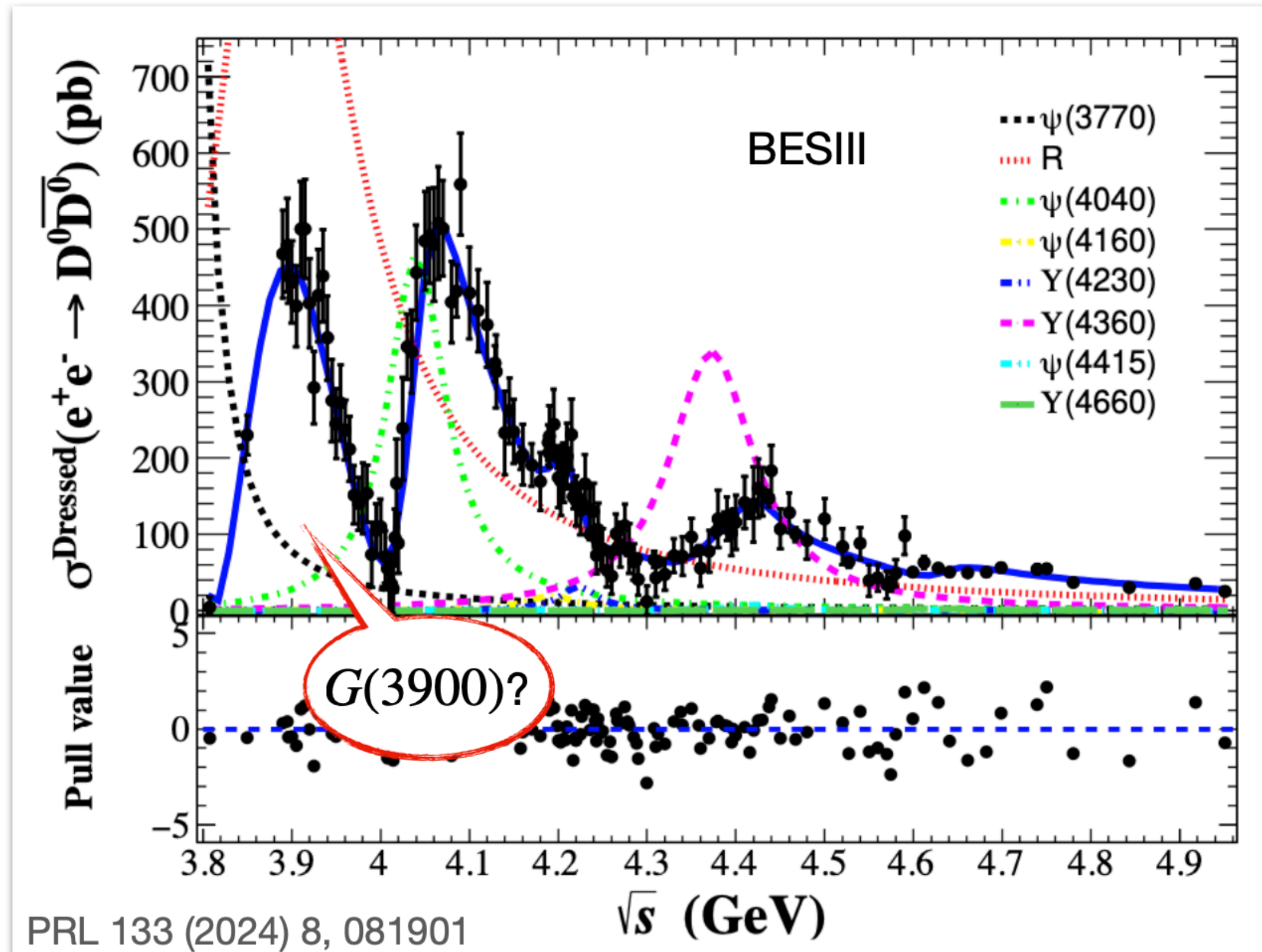
digamma



G(3900)

$$e^+e^- \rightarrow D\bar{D}$$

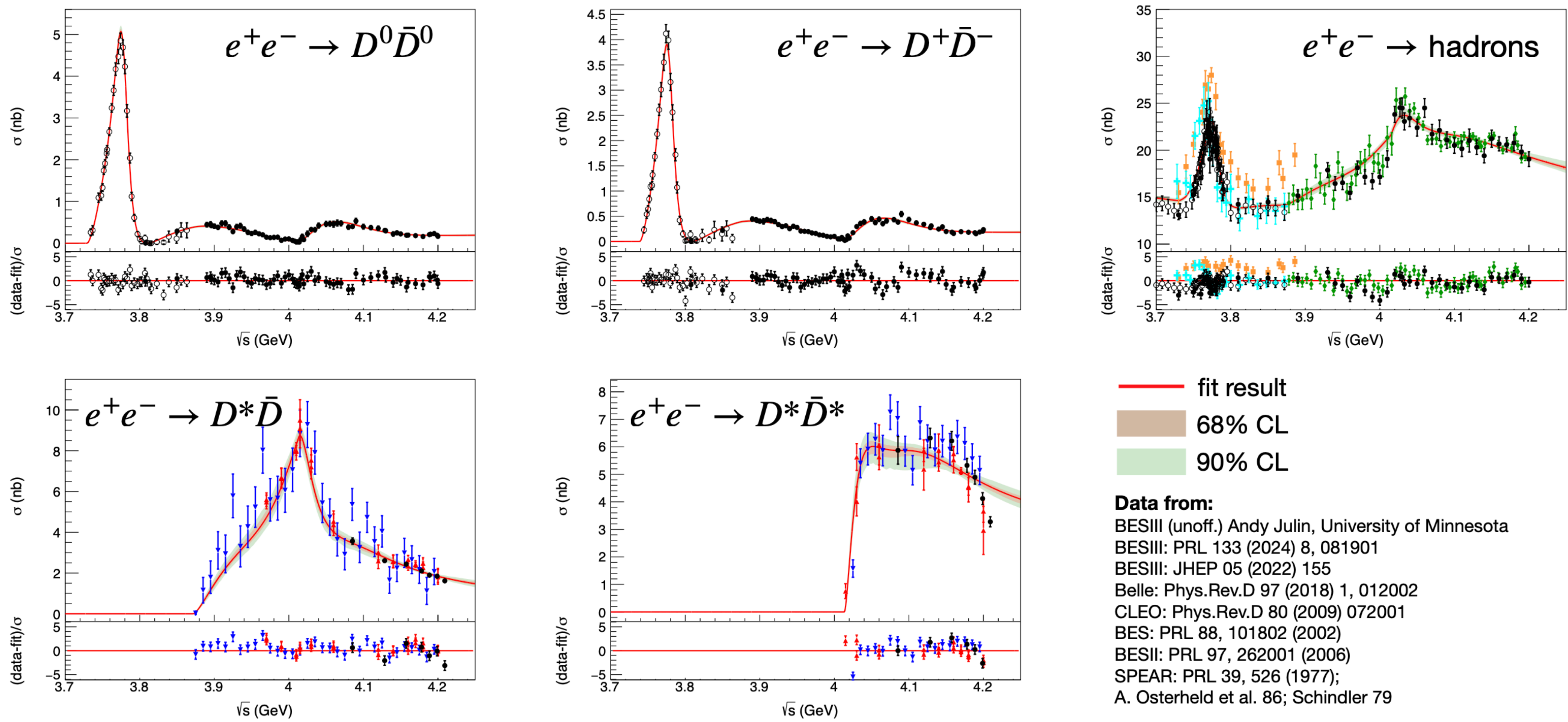
Eichten et al., Phys. Rev. D 21 (1980) 203

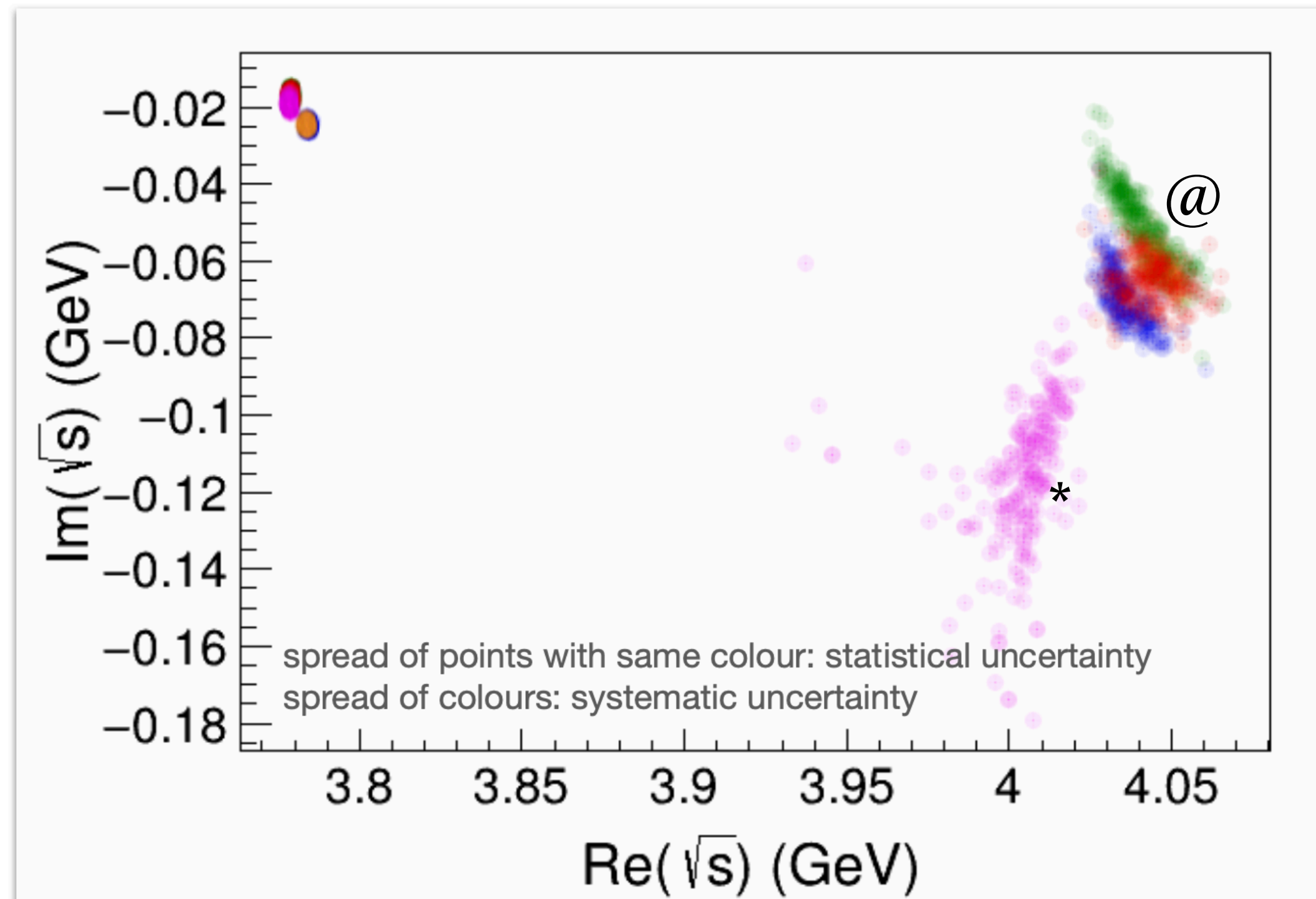


In our calculation there is some weak structure in the 3.9–4.0 GeV region. It does not arise from a $c\bar{c}$ resonance, but from the opening of the $D\bar{D}^* + D^*\bar{D}$ channel and a decrease in the $D\bar{D}$ channel due to a nearby zero in the $3S$ decay amplitude.

N. Hüsken, R.F. Lebed, R.E. Mitchell, E.S. Swanson, Y-Q Wang, 2404.03896

Refit with additional information and coupled channels





the devil's in the details:

- given our choices of K_{ij} , $n(s)$, Σ , we find described without a $G(3900)$ pole
- details we varied:
 - isospin-constraints between D^+D^- and $D^0\bar{D}^0$
 - threshold opening of channel that absorbs *missing* intensity
 - node in $\hat{n}(s) = n(s) \cdot (1 - k^2/k_0^2)$
- details we did not vary:
 - type of barrier factor $b_L(s)$
 - choice of Σ
- very valuable that other groups use different assumptions entirely different approaches

PRD 109 (2024) 11, 114015
 PRL 133 (2024) 24, 241903
 PRD 112 (2025) 5, 054027
 PRD 112 (2025), 016015
 arXiv:2509.17679

some find a $G(3900)$,
 some do not...

+ more in other energy regions

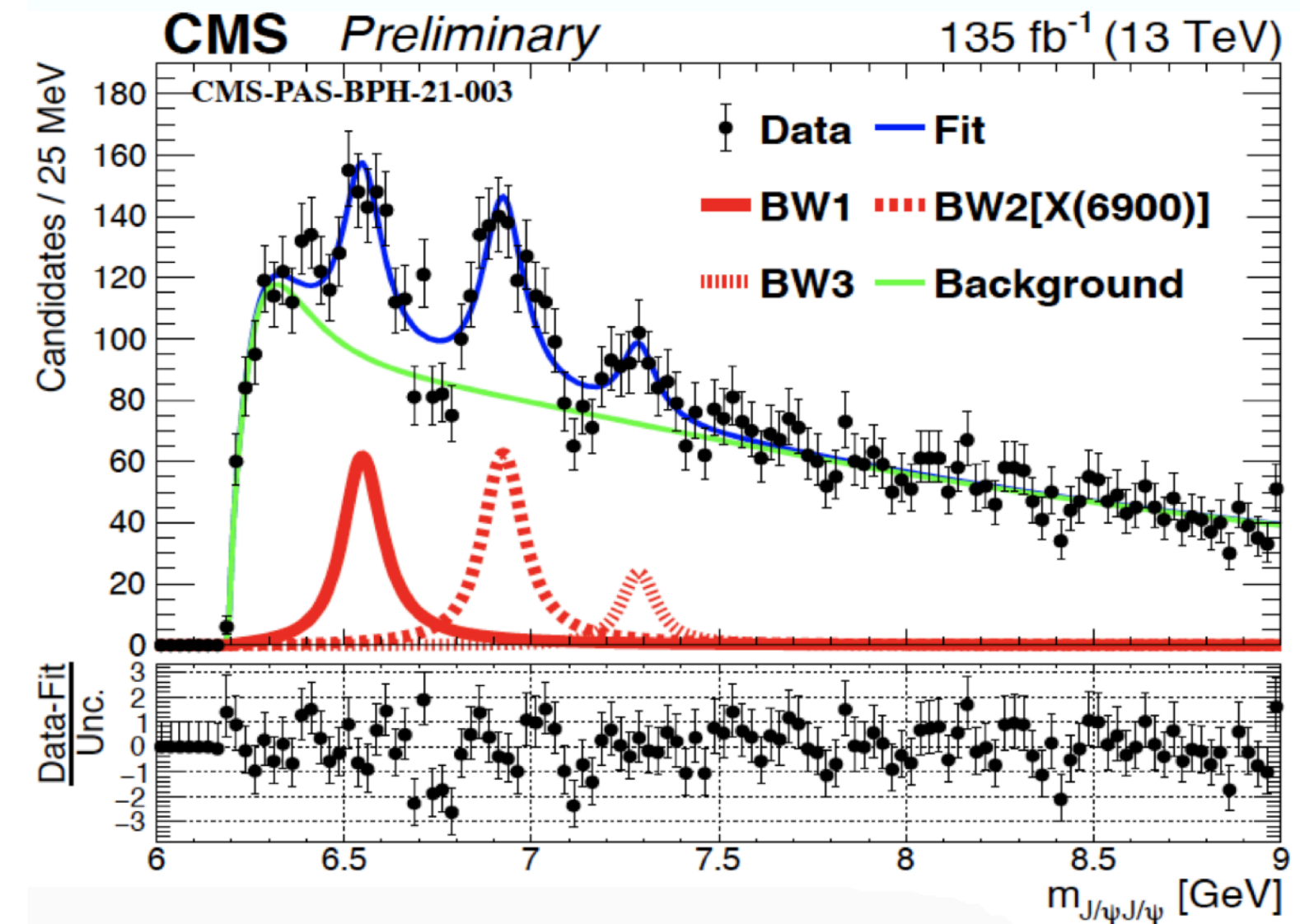
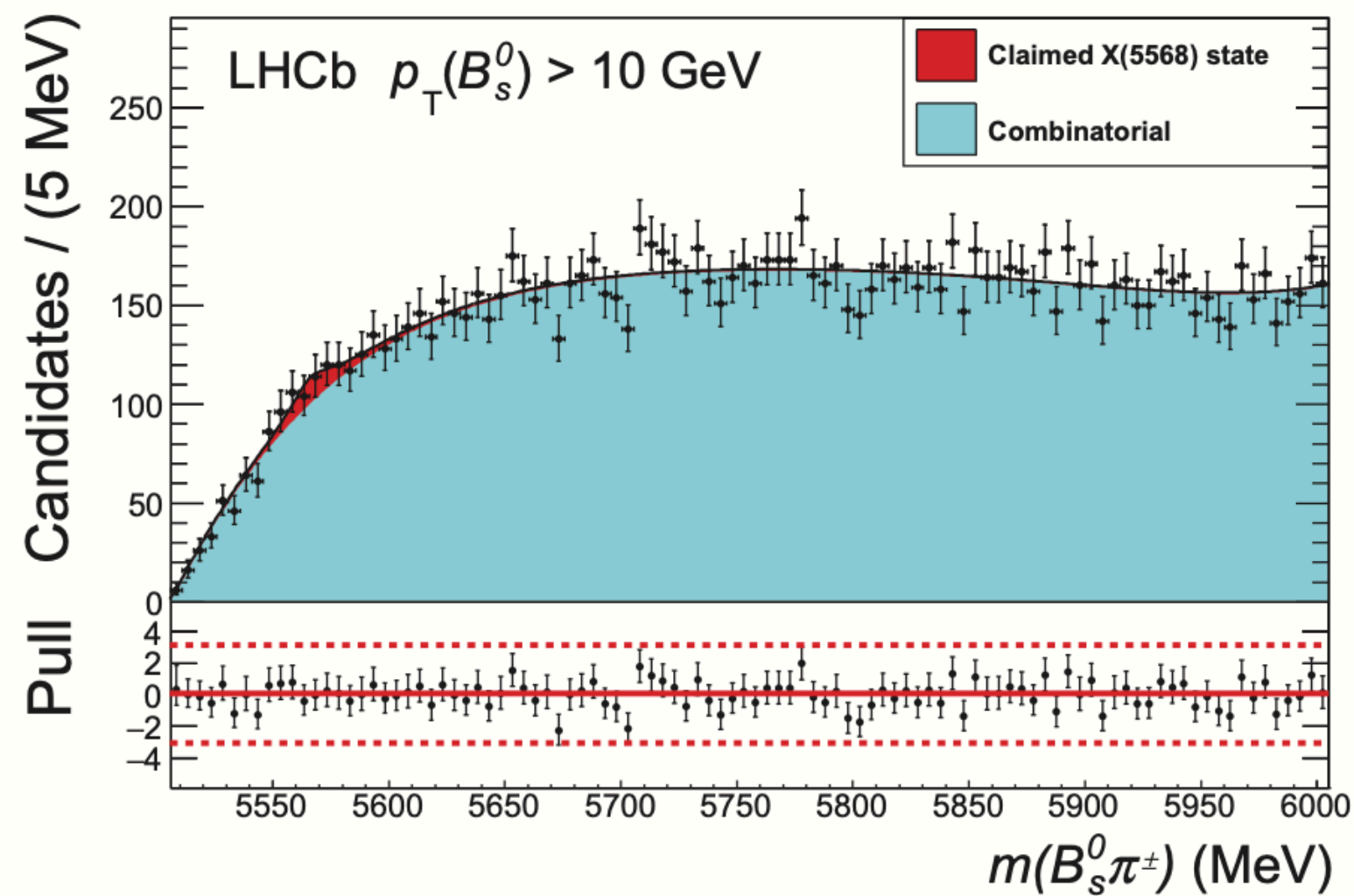
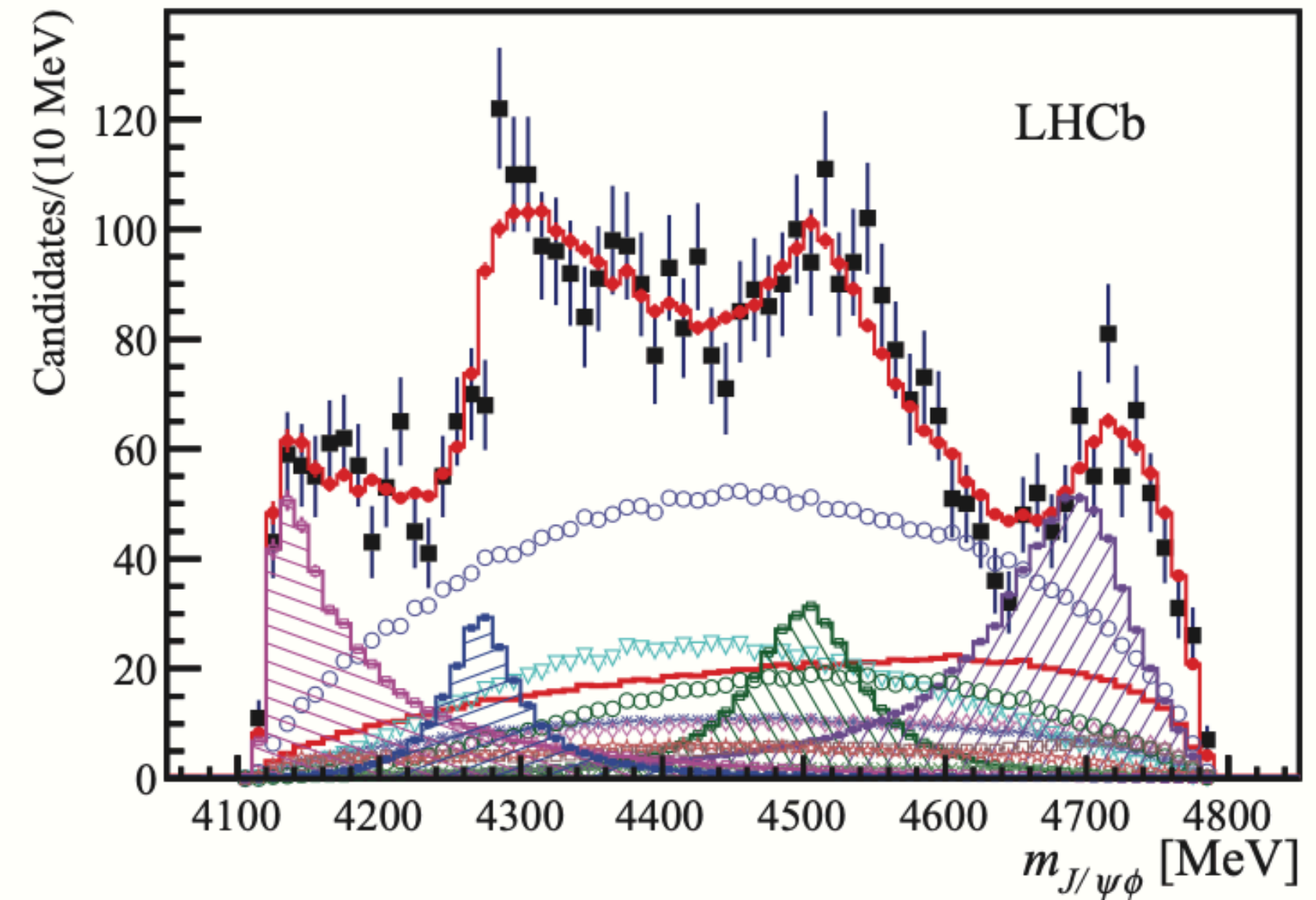
(@) 4 models w/ 24-30 parameters

(*) model with node

problems with the traditional approach

problems with the traditional approach

- i. **fluctuations in the data set may be important**
- ii. models M0 and M1 are wrong!
- iii. systematic errors are often underestimated
- iv. problematic overfitting
- v. we wish to extract model structure, not assume it



problems with the traditional approach

i. fluctuations in the data set may be important

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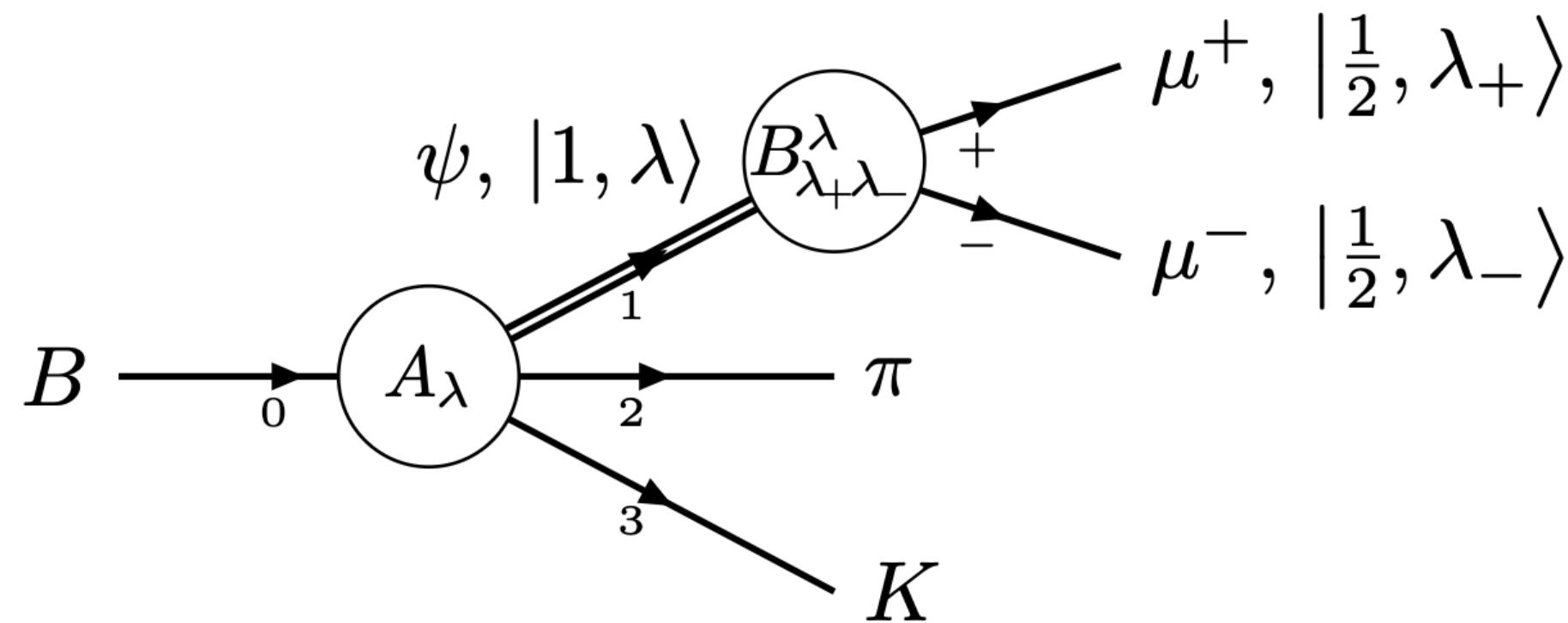
v. we wish to extract model structure, not assume

$$\text{BW}(m|M_0, \Gamma_0) = \frac{1}{M_0^2 - m^2 - iM_0\Gamma(m)},$$

$$\Gamma(m) = \Gamma_0 \left(\frac{q}{q_0}\right)^{2L_{\Lambda^*}+1} \frac{M_0}{m} B'_{L_{\Lambda^*}}(q, q_0, d)^2.$$

$$R_{\Lambda_n^*}(m_{Kp}) = B'_{L_{\Lambda_b^0}}(p, p_0, d) \left(\frac{p}{M_{\Lambda_b^0}}\right)^{L_{\Lambda_b^0}} \text{BW}(m_{Kp}|M_0^{\Lambda_n^*}, \Gamma_0^{\Lambda_n^*}) B'_{L_{\Lambda_n^*}}(q, q_0, d) \left(\frac{q}{M_0^{\Lambda_n^*}}\right)^{L_{\Lambda_n^*}}.$$

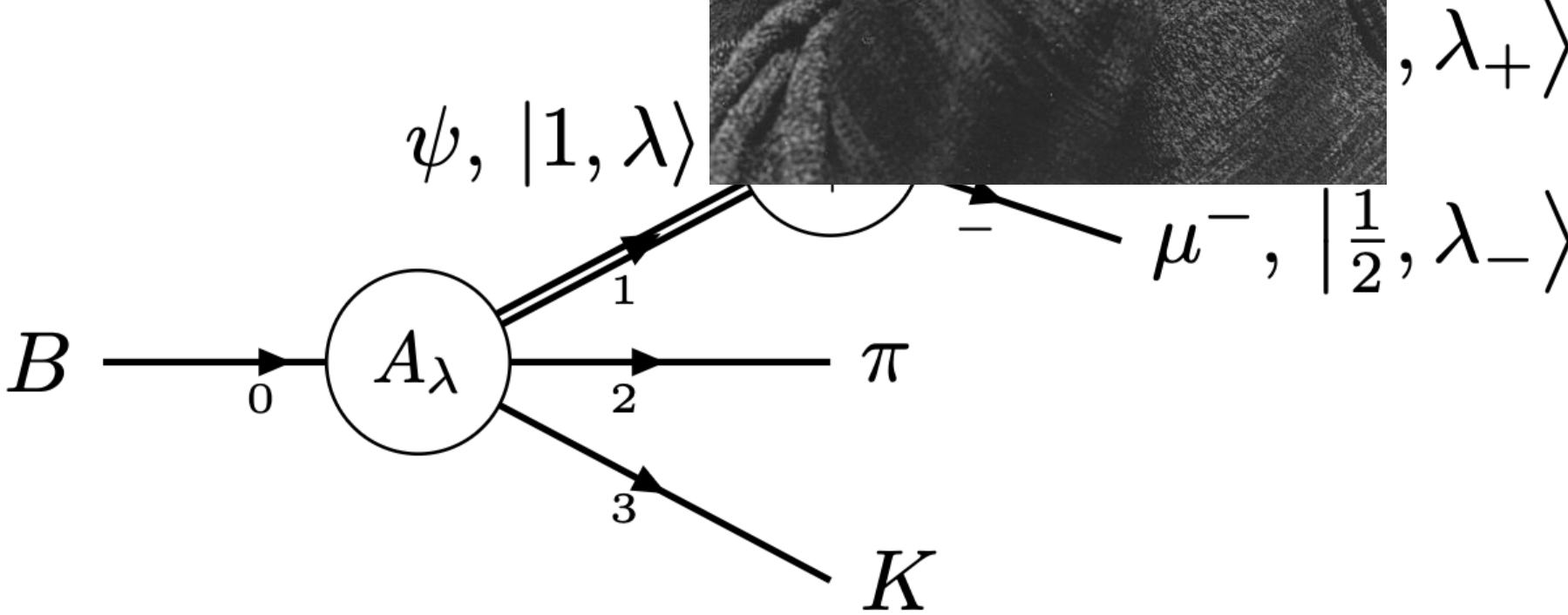
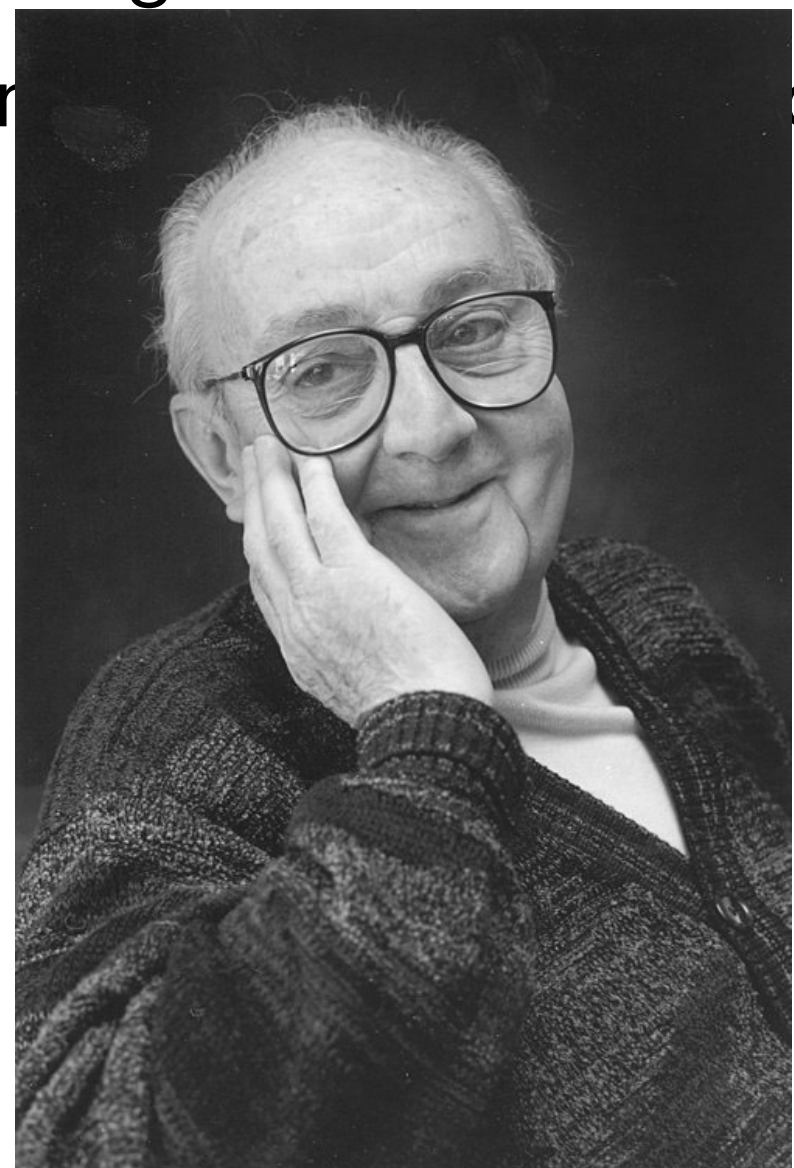
$$\begin{aligned} \mathcal{M}_{\lambda_{\Lambda_b^0}^{\Lambda^*}, \lambda_p, \Delta\lambda_\mu}^{\Lambda^*} &= \sum_n R_{\Lambda_n^*}(m_{Kp}) \mathcal{H}_{\lambda_p}^{\Lambda_n^* \rightarrow Kp} \sum_{\lambda_\psi} e^{i\lambda_\psi \phi_\mu} d_{\lambda_\psi, \Delta\lambda_\mu}^1(\theta_\psi) \\ &\times \sum_{\lambda_{\Lambda^*}} \mathcal{H}_{\lambda_{\Lambda^*}, \lambda_\psi}^{\Lambda_b^0 \rightarrow \Lambda_n^* \psi} e^{i\lambda_{\Lambda^*} \phi_K} d_{\lambda_{\Lambda_b^0}^{\frac{1}{2}}, \lambda_{\Lambda^*} - \lambda_\psi}(\theta_{\Lambda_b^0}) d_{\lambda_{\Lambda^*}, \lambda_p}^{J_{\Lambda_n^*}}(\theta_{\Lambda^*}). \end{aligned}$$



- unitary?
- correct analytic structure?
- crossing symmetric?
- non-perturbative?

problems with the traditional approach

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- v. we wish to extract more information



$$\text{BW}(m|M_0, \Gamma_0) = \frac{1}{M_0^2 - m^2 - iM_0\Gamma(m)},$$

$$\Gamma(m) = \Gamma_0 \left(\frac{q}{q_0}\right)^{2L_{\Lambda^*}+1} \frac{M_0}{m} B'_{L_{\Lambda^*}}(q, q_0, d)^2.$$

assume $R_{\Lambda_n^*}(m_{Kp}) = B'_{L_{\Lambda_b^0}}(p, p_0, d) \left(\frac{p}{M_{\Lambda_b^0}}\right)^{L_{\Lambda_b^0}^*} \text{BW}(m_{Kp}|M_0^{\Lambda_n^*}, \Gamma_0^{\Lambda_n^*}) B'_{L_{\Lambda_n^*}}(q, q_0, d) \left(\frac{q}{M_0^{\Lambda_n^*}}\right)^{L_{\Lambda_n^*}^*}.$

$$\mathcal{M}_{\lambda_{\Lambda_b^0}^0, \lambda_p, \Delta\lambda_\mu}^{\Lambda^*} = \sum_n R_{\Lambda_n^*}(m_{Kp}) \mathcal{H}_{\lambda_p}^{\Lambda_n^* \rightarrow Kp} \sum_{\lambda_\psi} e^{i\lambda_\psi \phi_\mu} d_{\lambda_\psi, \Delta\lambda_\mu}^1(\theta_\psi) \times \sum_{\lambda_{\Lambda^*}} \mathcal{H}_{\lambda_{\Lambda^*}, \lambda_\psi}^{\Lambda_b^0 \rightarrow \Lambda_n^* \psi} e^{i\lambda_{\Lambda^*} \phi_K} d_{\lambda_{\Lambda_b^0}^0, \lambda_{\Lambda^*} - \lambda_\psi}^{\frac{1}{2}}(\theta_{\Lambda_b^0}^0) d_{\lambda_{\Lambda^*}, \lambda_p}^{J_{\Lambda_n^*}}(\theta_{\Lambda^*}).$$

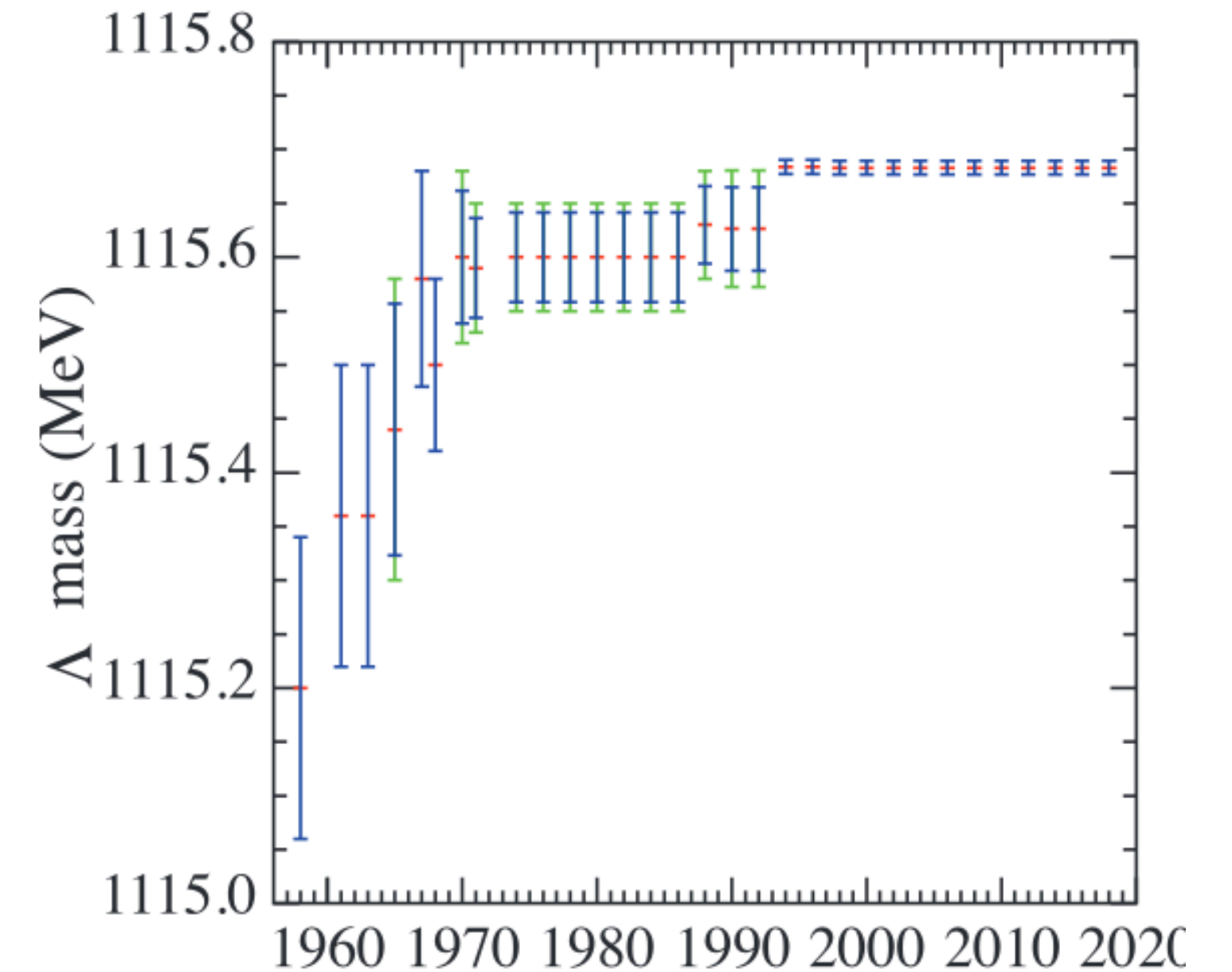
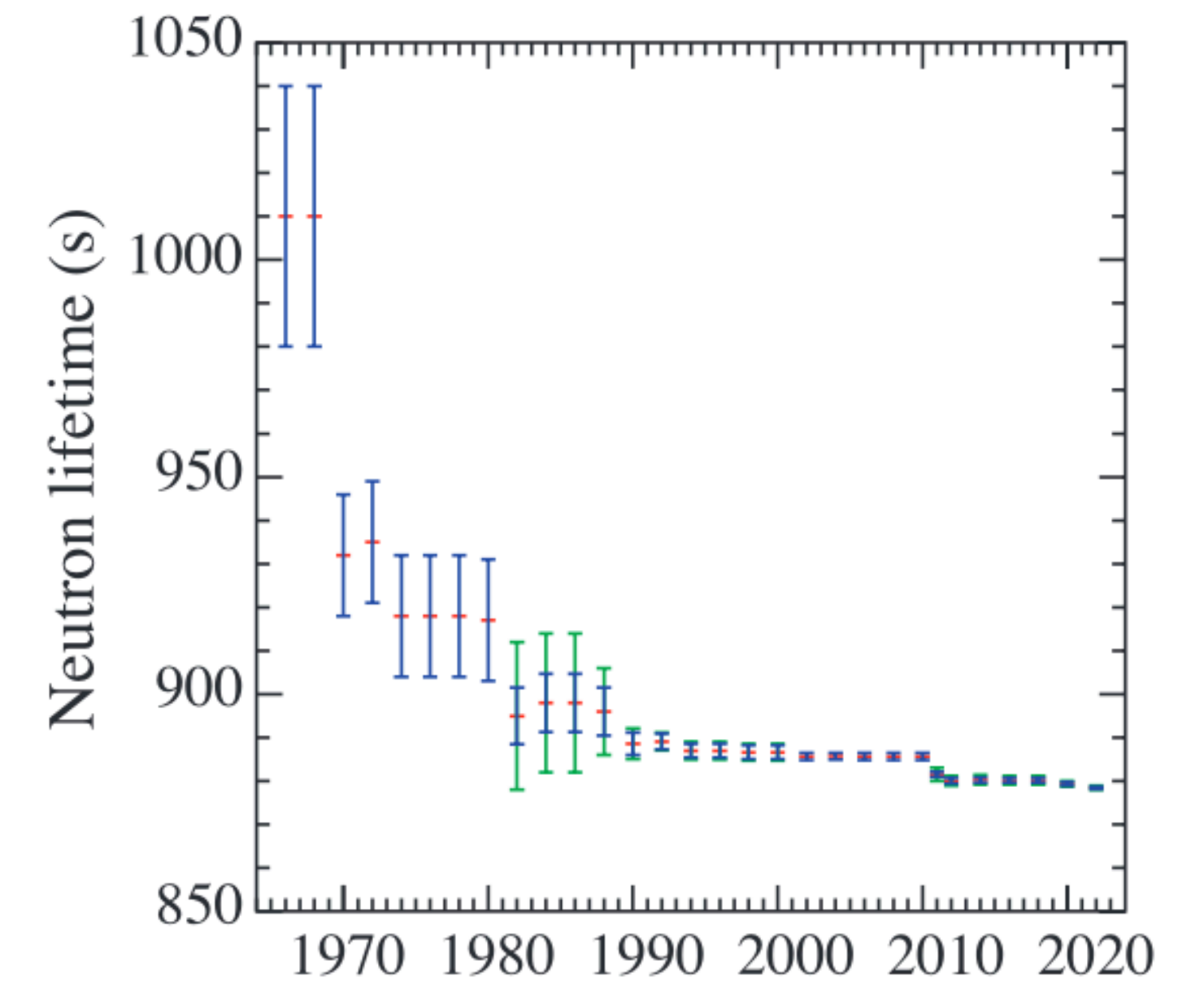
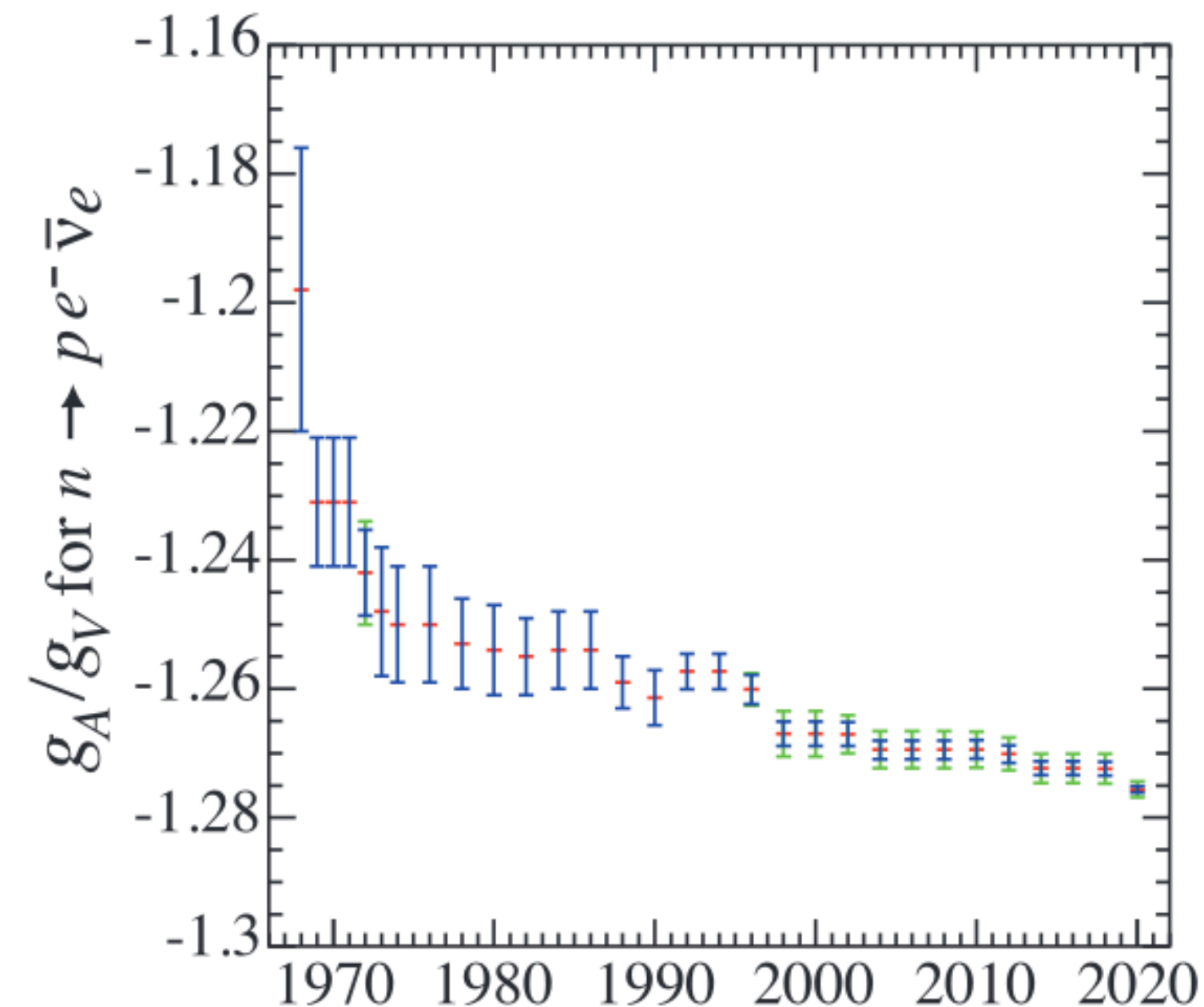
- unitary?
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- v. we wish to extract model structure, not assume it

i.e. over-confidence in one's model.

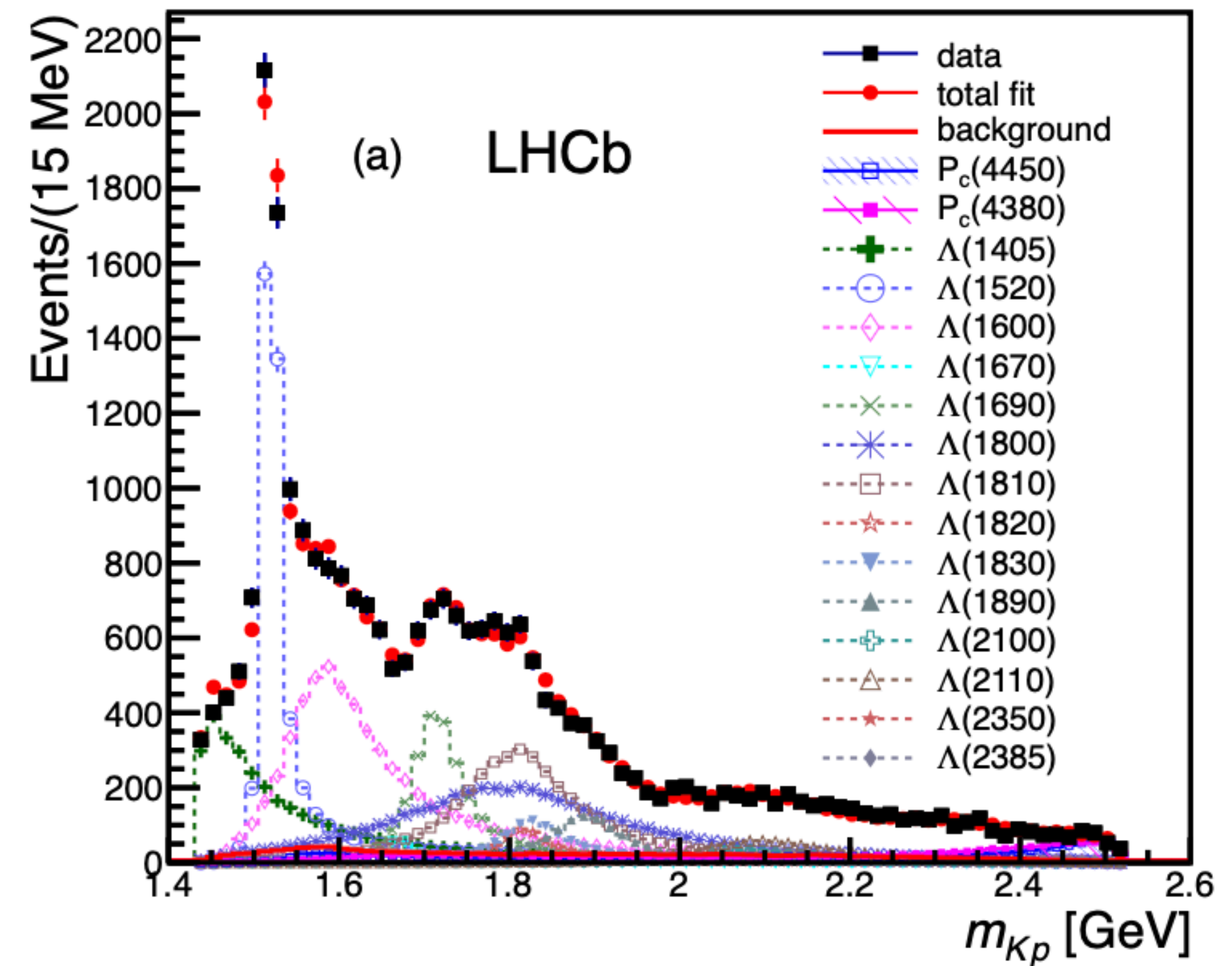
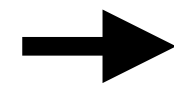
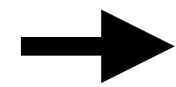
i.e. one acts as if the model generated the data.



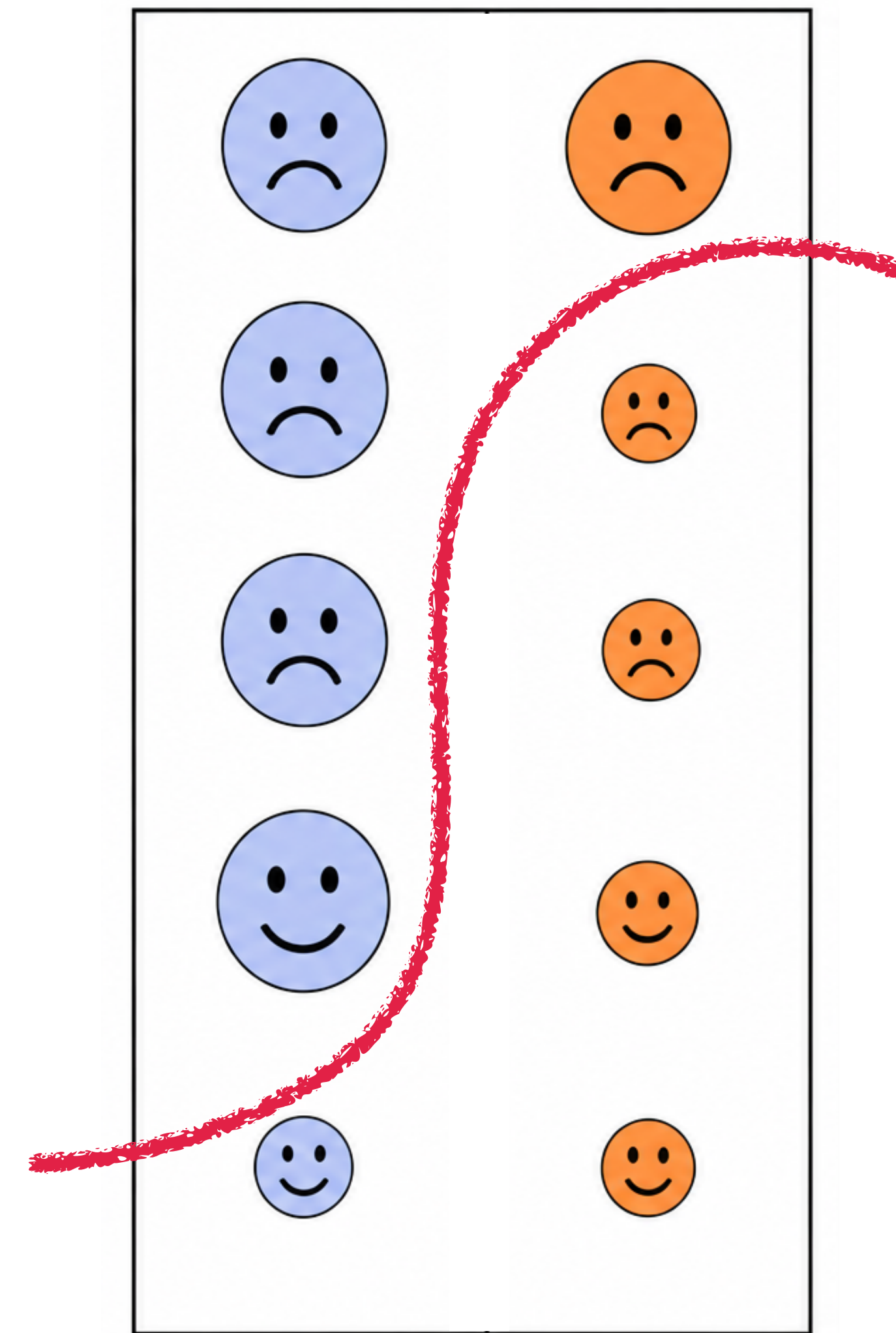
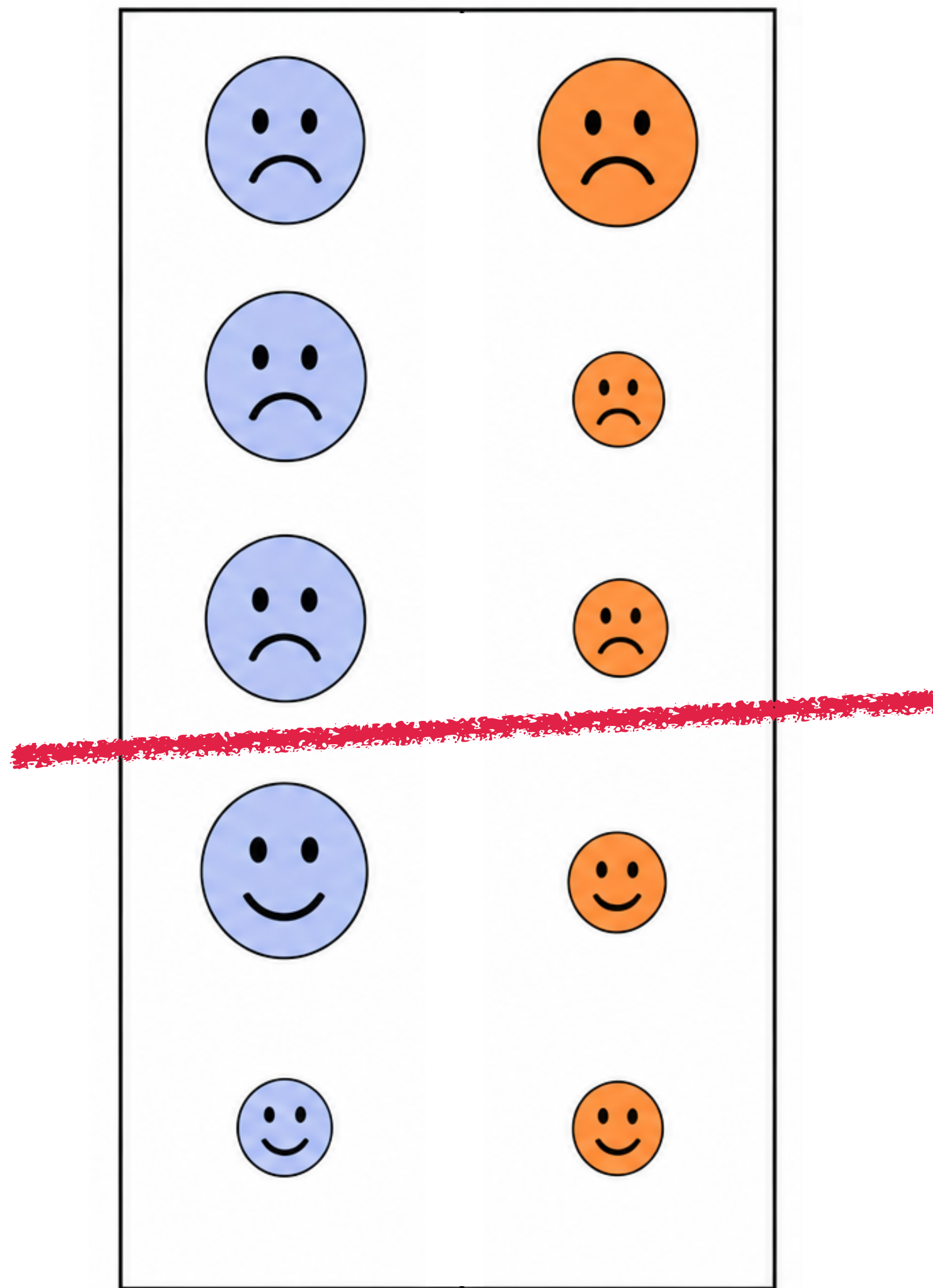
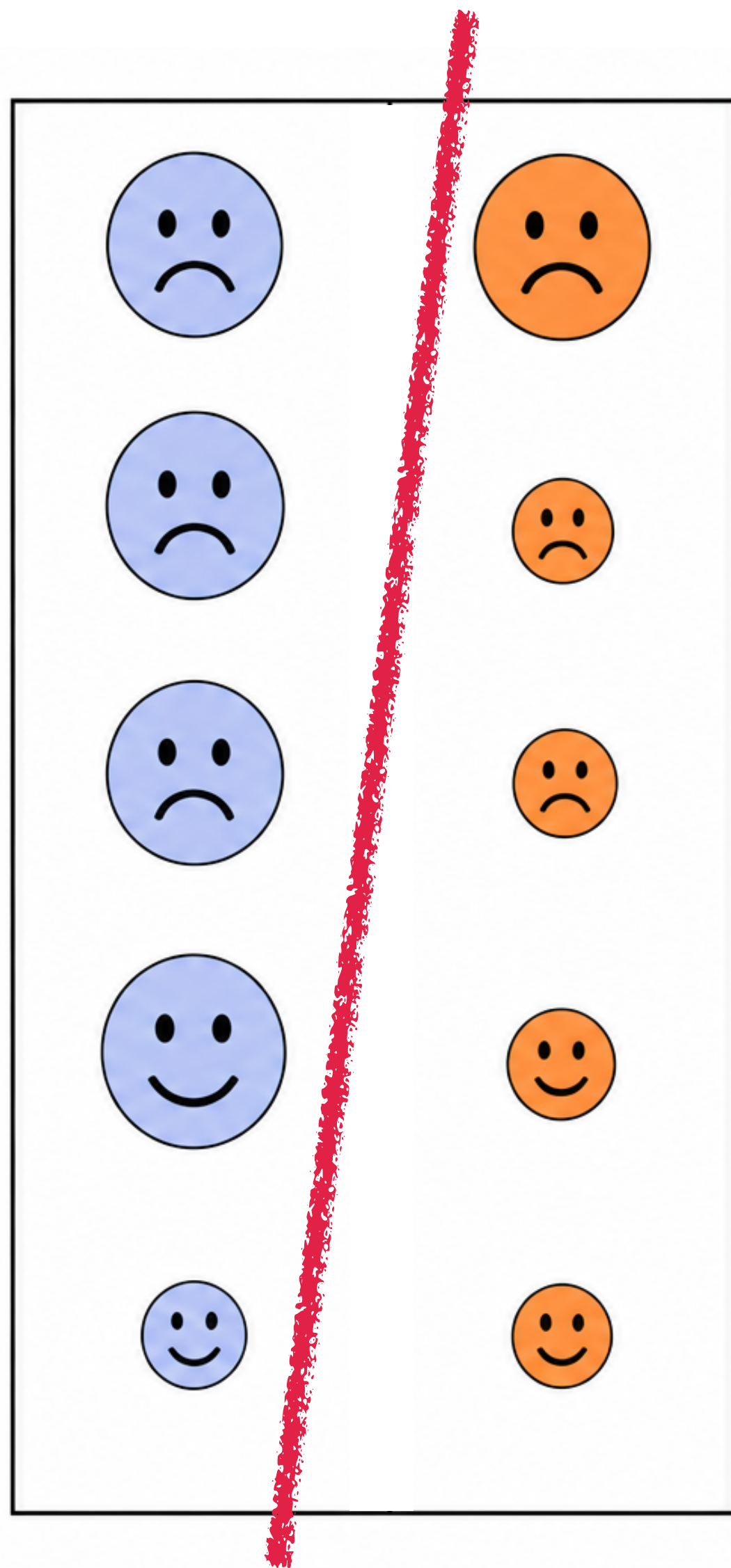
problems with the traditional approach

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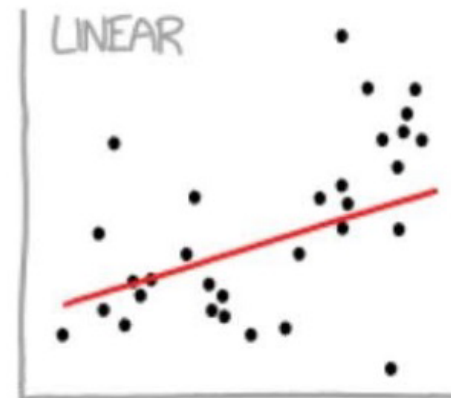
"With four parameters I can fit an elephant, and with five I can make him wiggle his trunk."



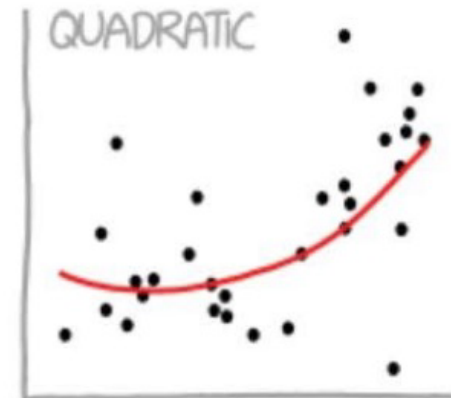
Your fit heuristic is domain-dependent!



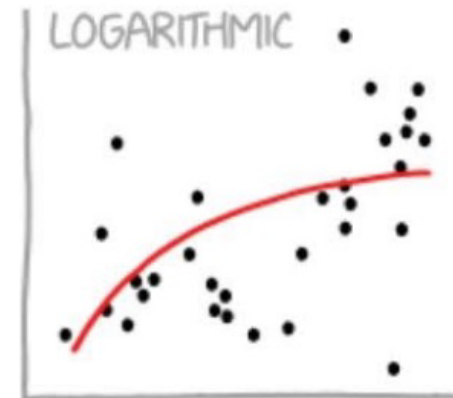
CURVE-FITTING METHODS AND THE MESSAGES THEY SEND



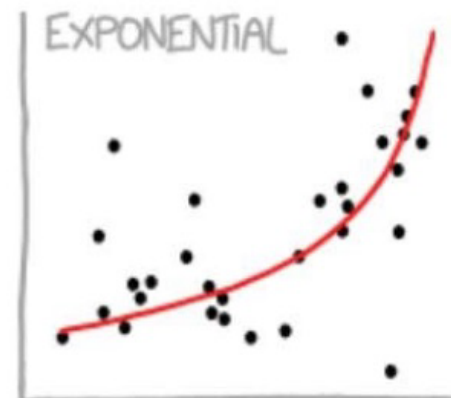
LINEAR
"HEY, I DID A REGRESSION."



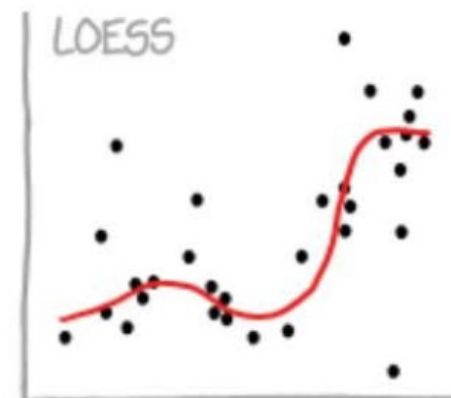
QUADRATIC
"I WANTED A CURVED LINE, SO I MADE ONE WITH MATH."



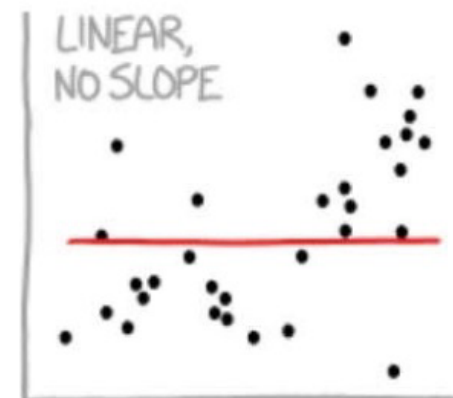
LOGARITHMIC
"LOOK, IT'S TAPERING OFF!"



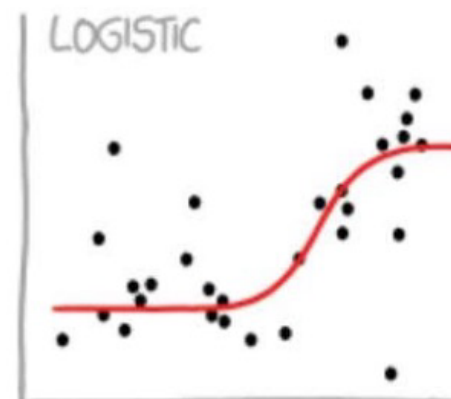
EXPONENTIAL
"LOOK, IT'S GROWING UNCONTROLLABLY!"



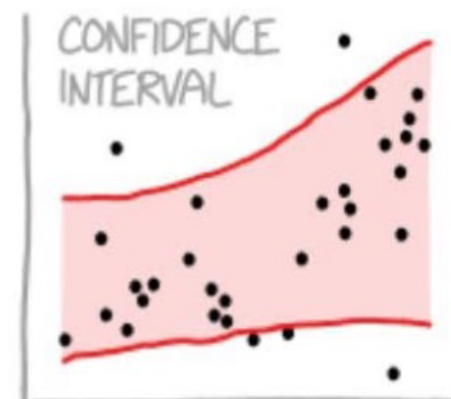
LOESS
"I'M SOPHISTICATED, NOT LIKE THOSE BUMBLING POLYNOMIAL PEOPLE."



LINEAR, NO SLOPE
"I'M MAKING A SCATTER PLOT BUT I DON'T WANT TO."



LOGISTIC
"I NEED TO CONNECT THESE TWO LINES, BUT MY FIRST IDEA DIDN'T HAVE ENOUGH MATH."



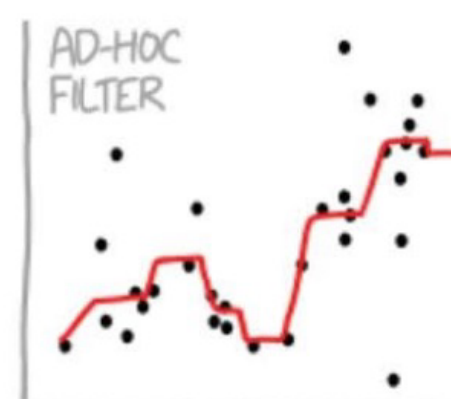
CONFIDENCE INTERVAL
"LISTEN, SCIENCE IS HARD. BUT I'M A SERIOUS PERSON DOING MY BEST."



PIECEWISE
"I HAVE A THEORY, AND THIS IS THE ONLY DATA I COULD FIND."



CONNECTING LINES
"I CLICKED 'SMOOTH LINES' IN EXCEL."



AD-HOC FILTER
"I HAD AN IDEA FOR HOW TO CLEAN UP THE DATA. WHAT DO YOU THINK?"



HOUSE OF CARDS
"AS YOU CAN SEE, THIS MODEL SMOOTHLY FITS THE— WAIT NO NO DON'T EXTEND IT AAAAAA!!!"

problems with the traditional approach

- i. fluctuations in the data set may be important
- ii. models M0 and M1 are wrong!
- iii. systematic errors are often underestimated
- iv. problematic overfitting**
- v. we wish to extract model structure, not assume it

a. LASSO

b. AIC

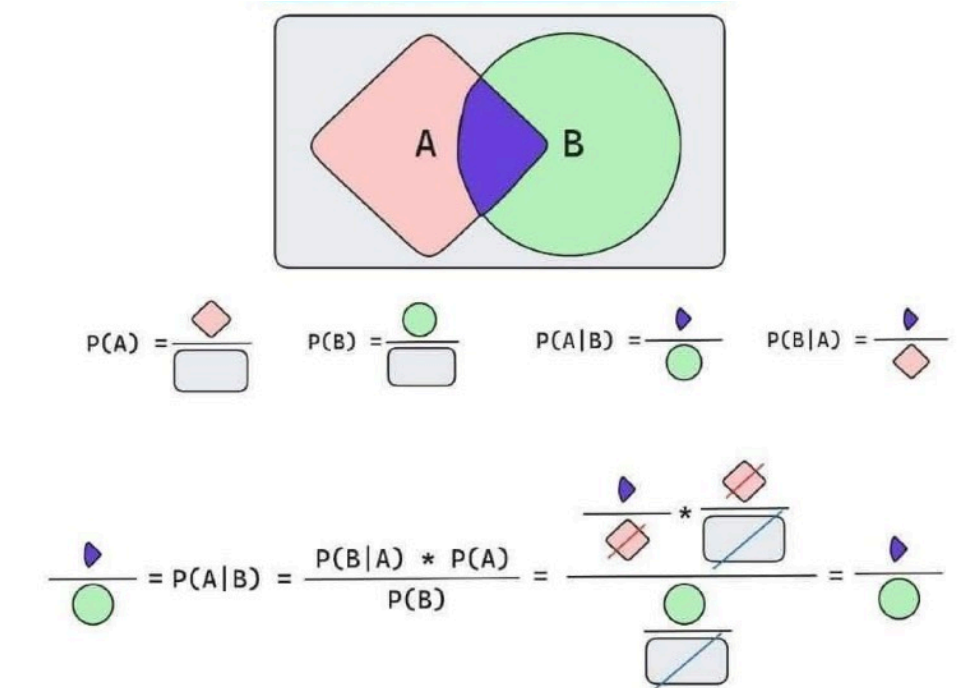
$$E_g(2k - 2 \log f(\mathcal{D} | \hat{\theta})) \rightarrow -2E_g(\log(f(\mathcal{D} | \theta)))$$

c. BIC

$$p(M | \mathcal{D}) \approx \exp \left[-\frac{1}{2}(k \ln n - 2 \log L(\hat{\theta})) + O(1) \right] p(M)$$

d. Bayesian Model Averaging :

$$p(M | \mathcal{D}) = \int d\theta p(\mathcal{D} | \theta; M) p(\theta | M) p(M)$$



Priors are problematic! What is a uniform prior? Lack of reparametrization invariance. What is the totality of causative agents?

additional concerns

vi. model parameters are (approximately?) meaningless

vii. we are not interested in the model, but rather the analytic structure

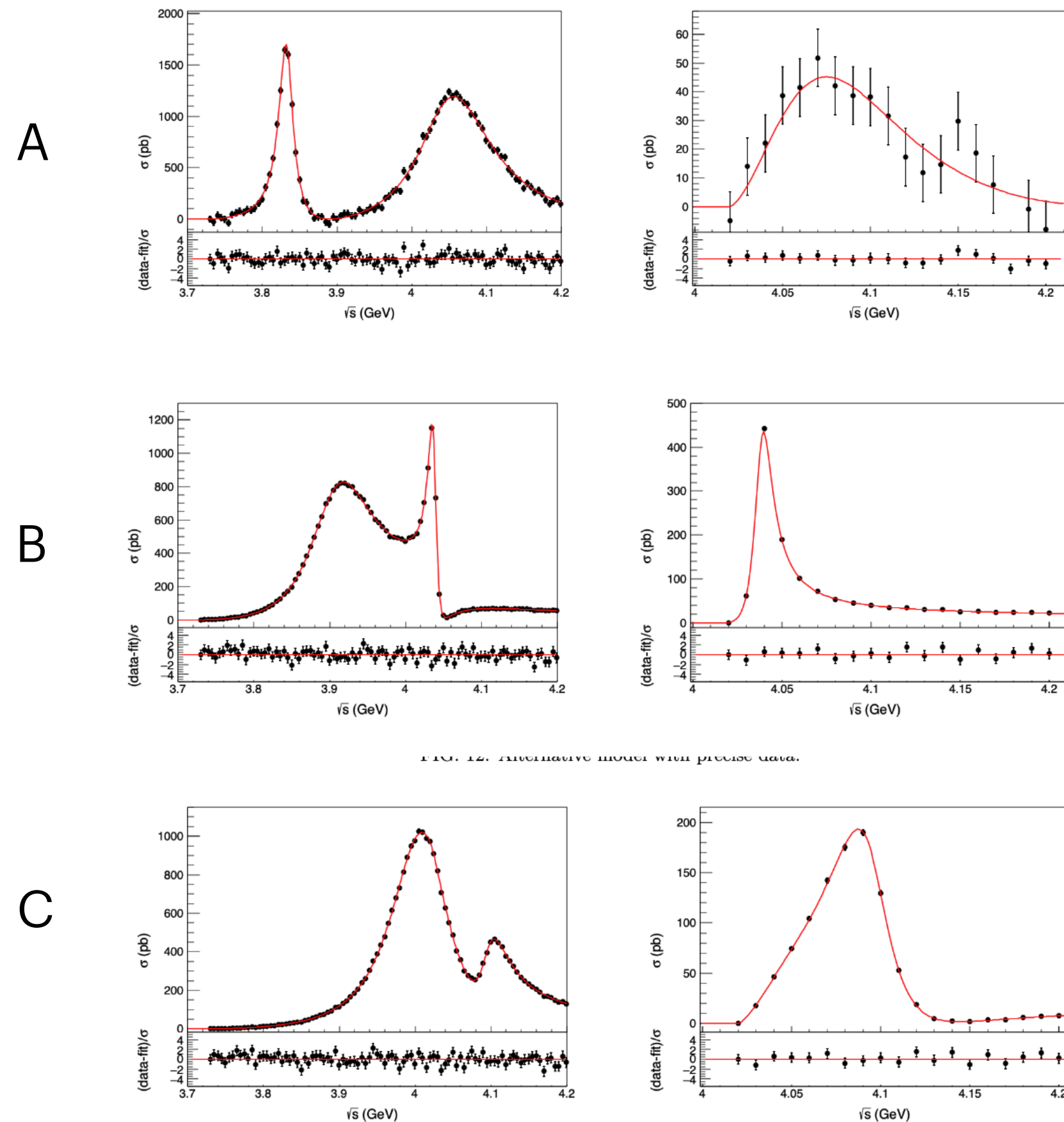
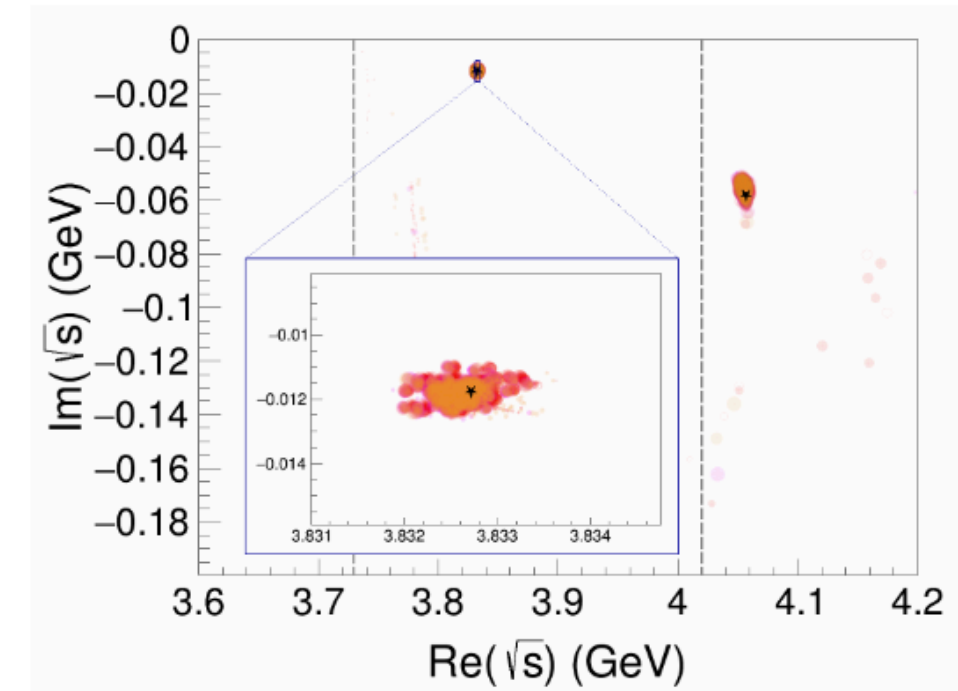
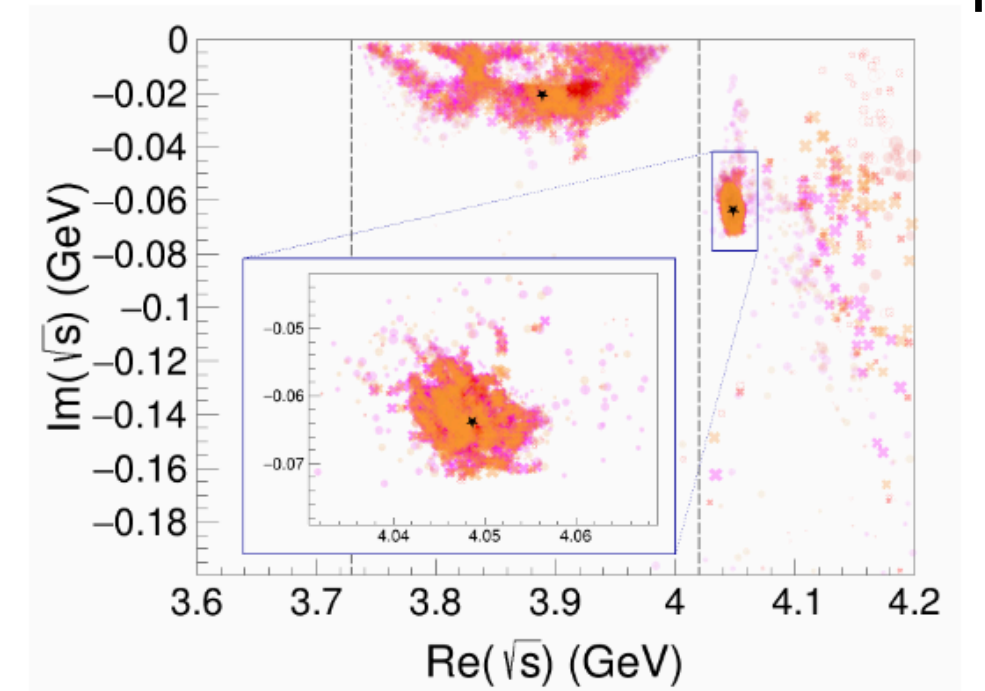


FIG. 12. Alternative model with precise data.

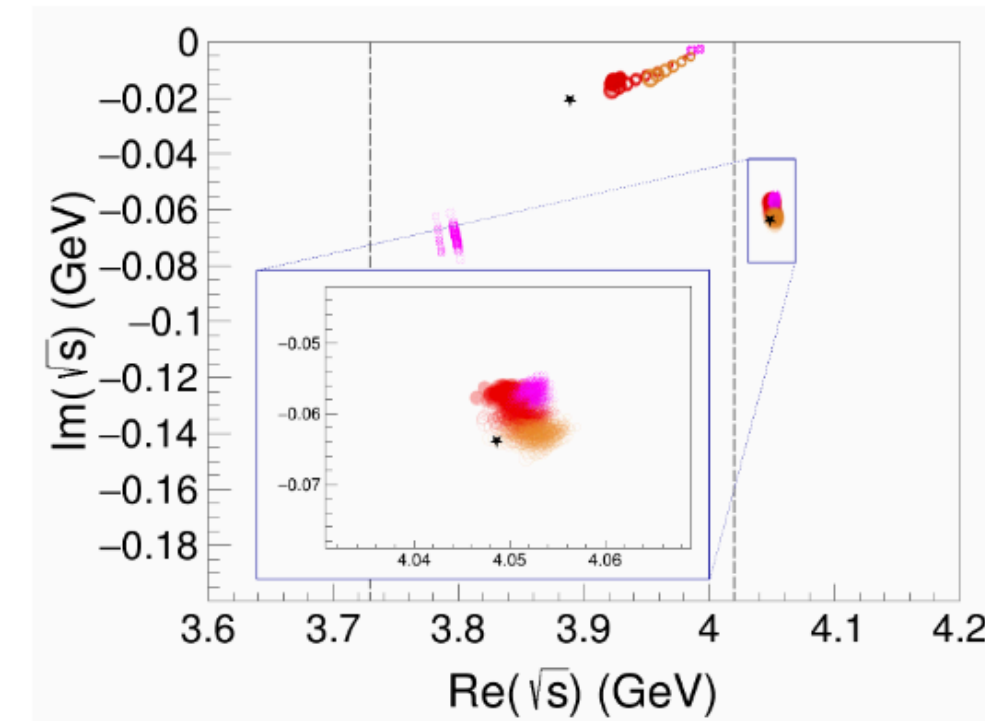
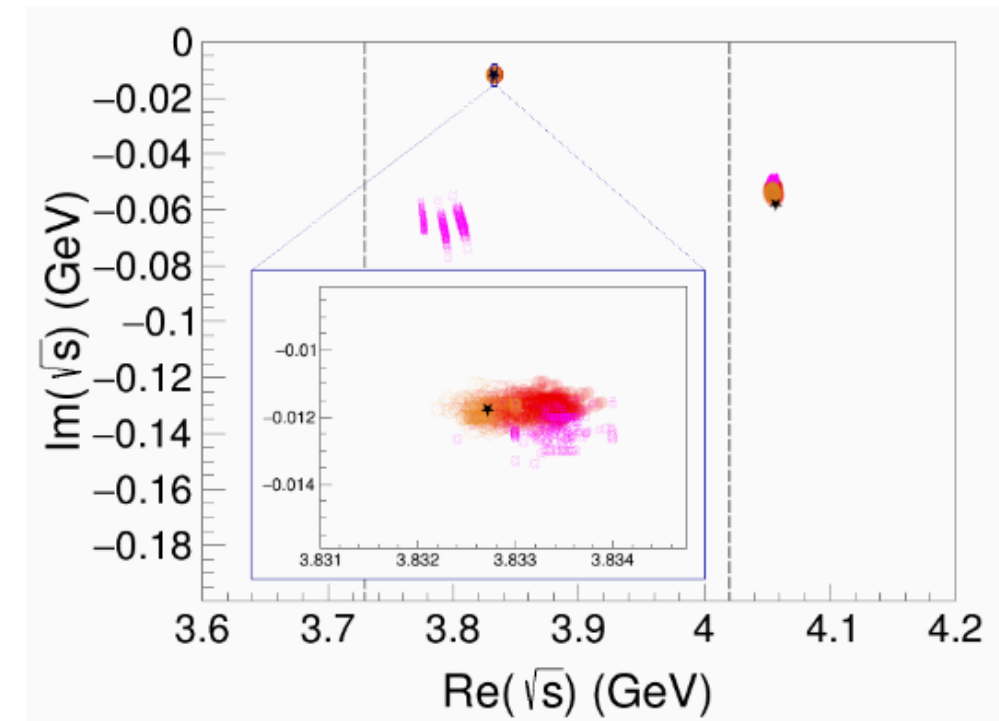
sheet II



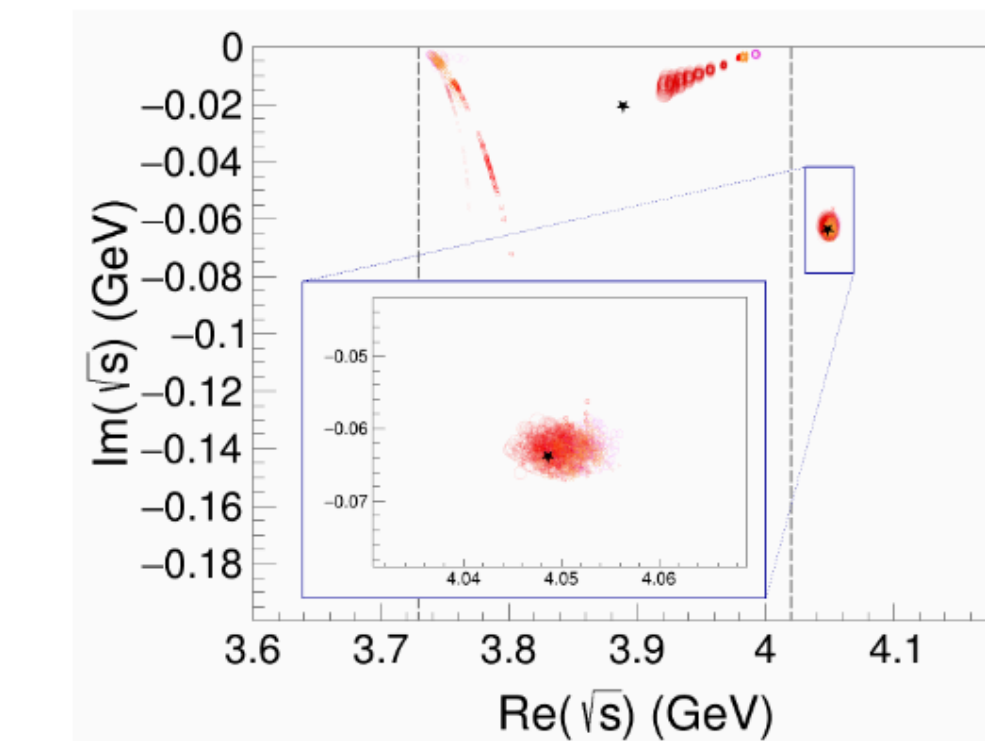
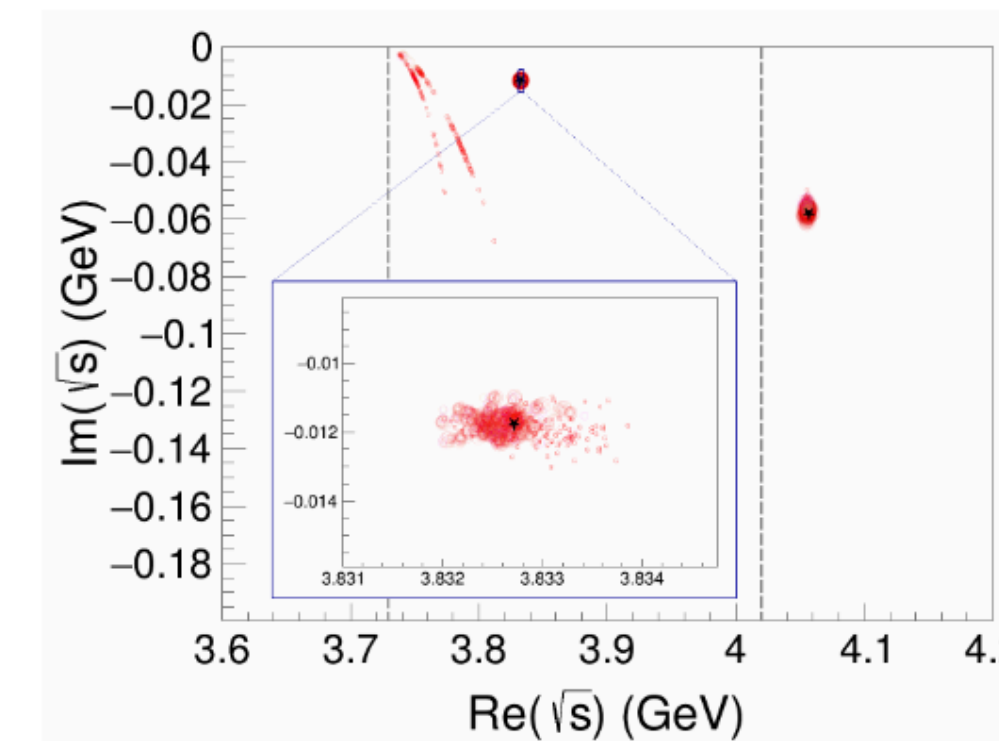
sheet III
2 bare poles



1 bare pole



0 bare poles

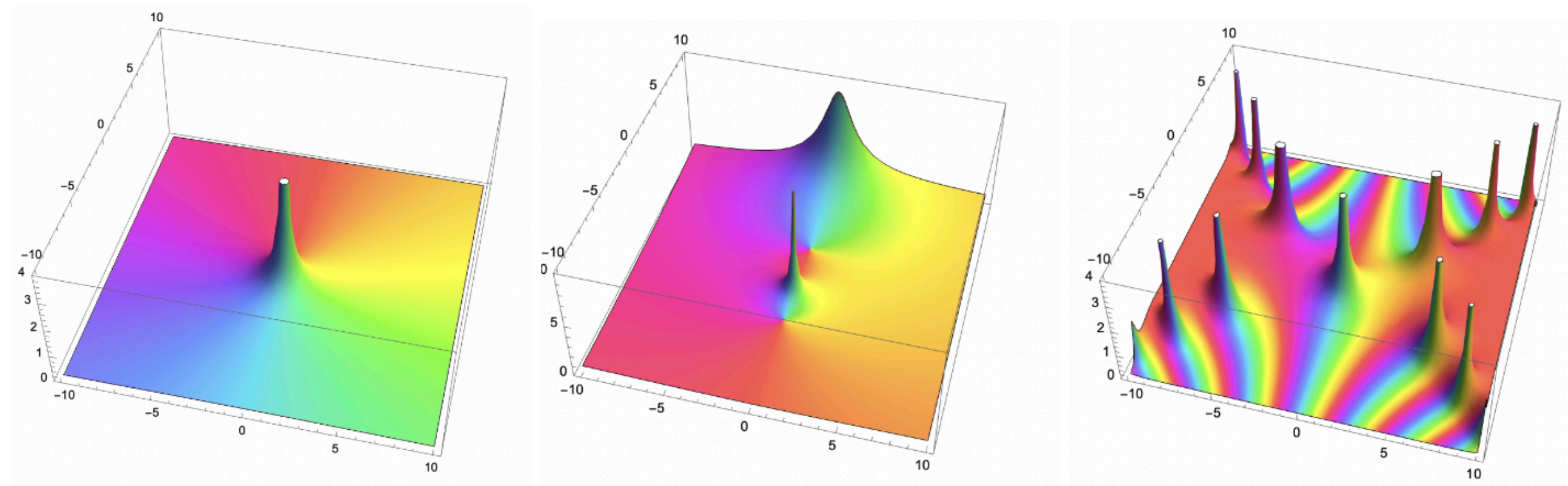


additional concerns

vi. model parameters are (approximately?) meaningless

vii. we are not interested in the model, but rather the analytic structure of the model

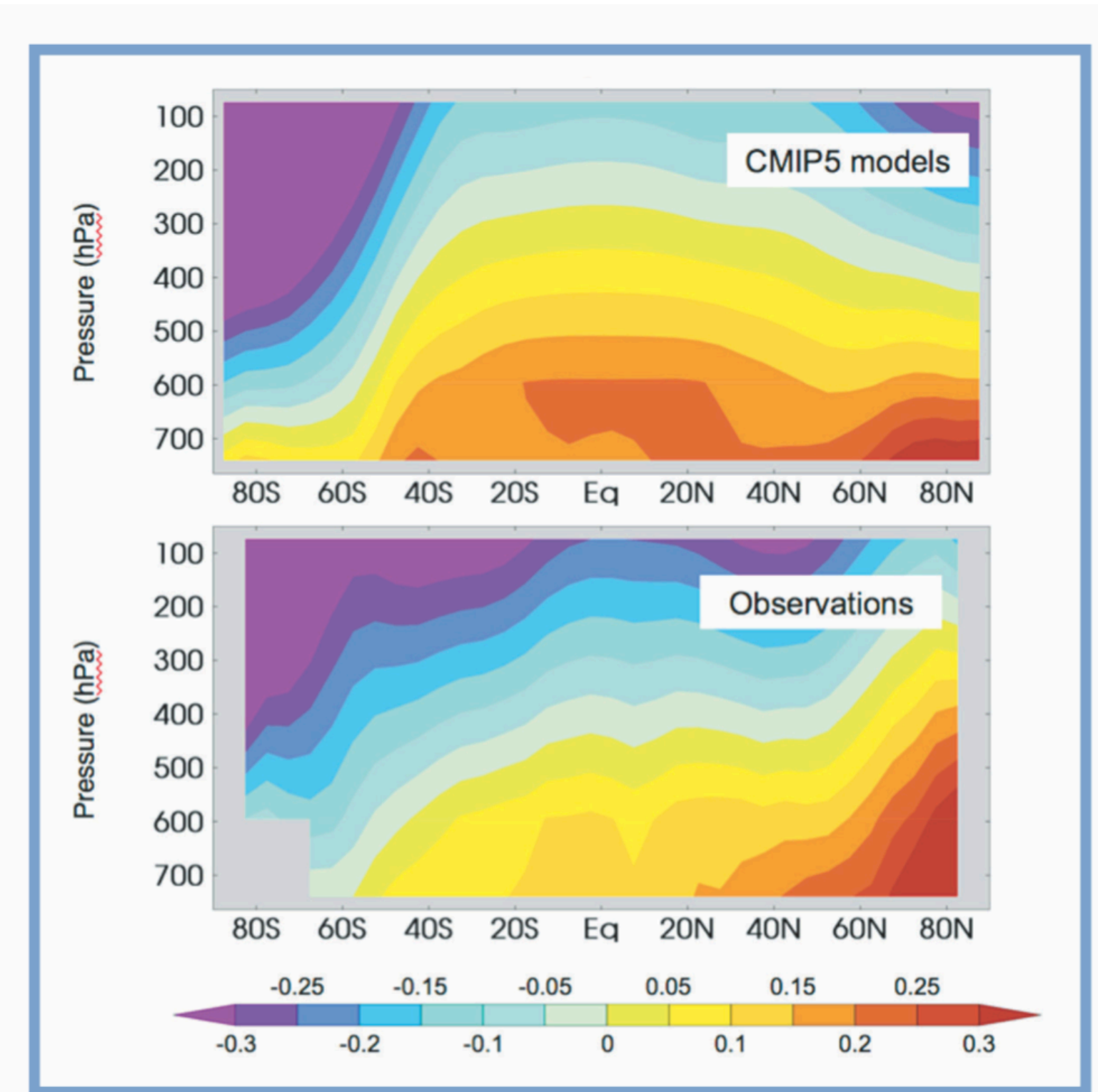
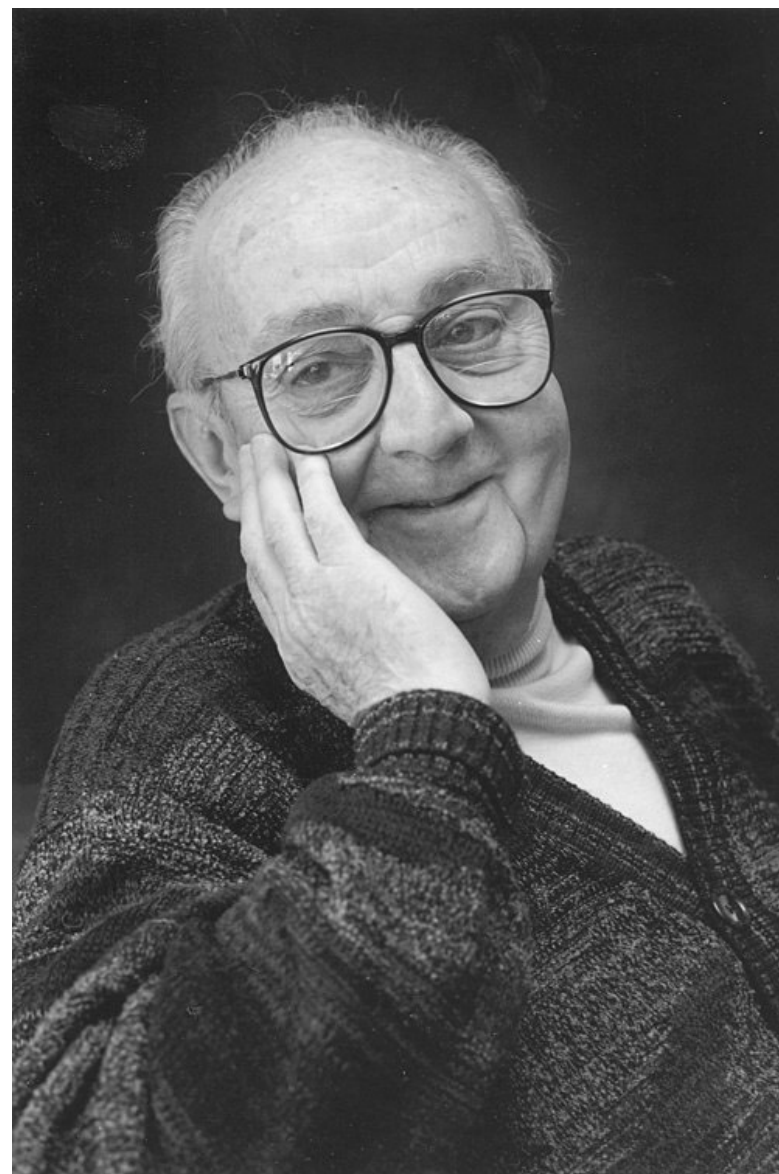
(or, at least, should be!)



bigger problems

viii. what does it mean to model?

ix. what does it mean to minimize?



"Fingerprinting" with changes in the vertical structure of atmospheric temperature: The average of eight CMIP-5 models with anthropogenic forcing (upper panel) and satellite observations from Remote Sensing Systems (lower panel) both show coherent warming of the troposphere and cooling of the stratosphere. (Source: Santer et al., Proceeding of the National Academy of Sciences, submitted)

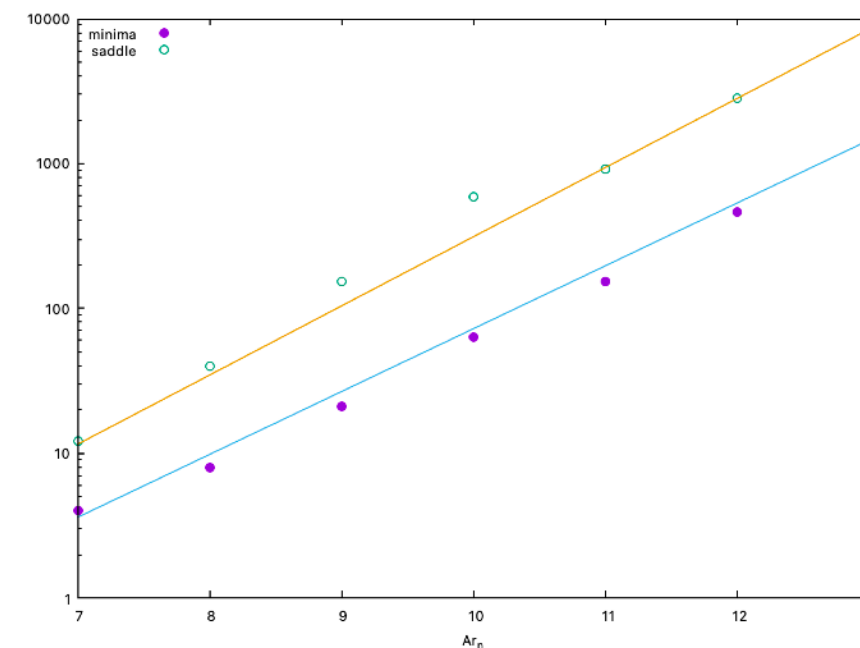
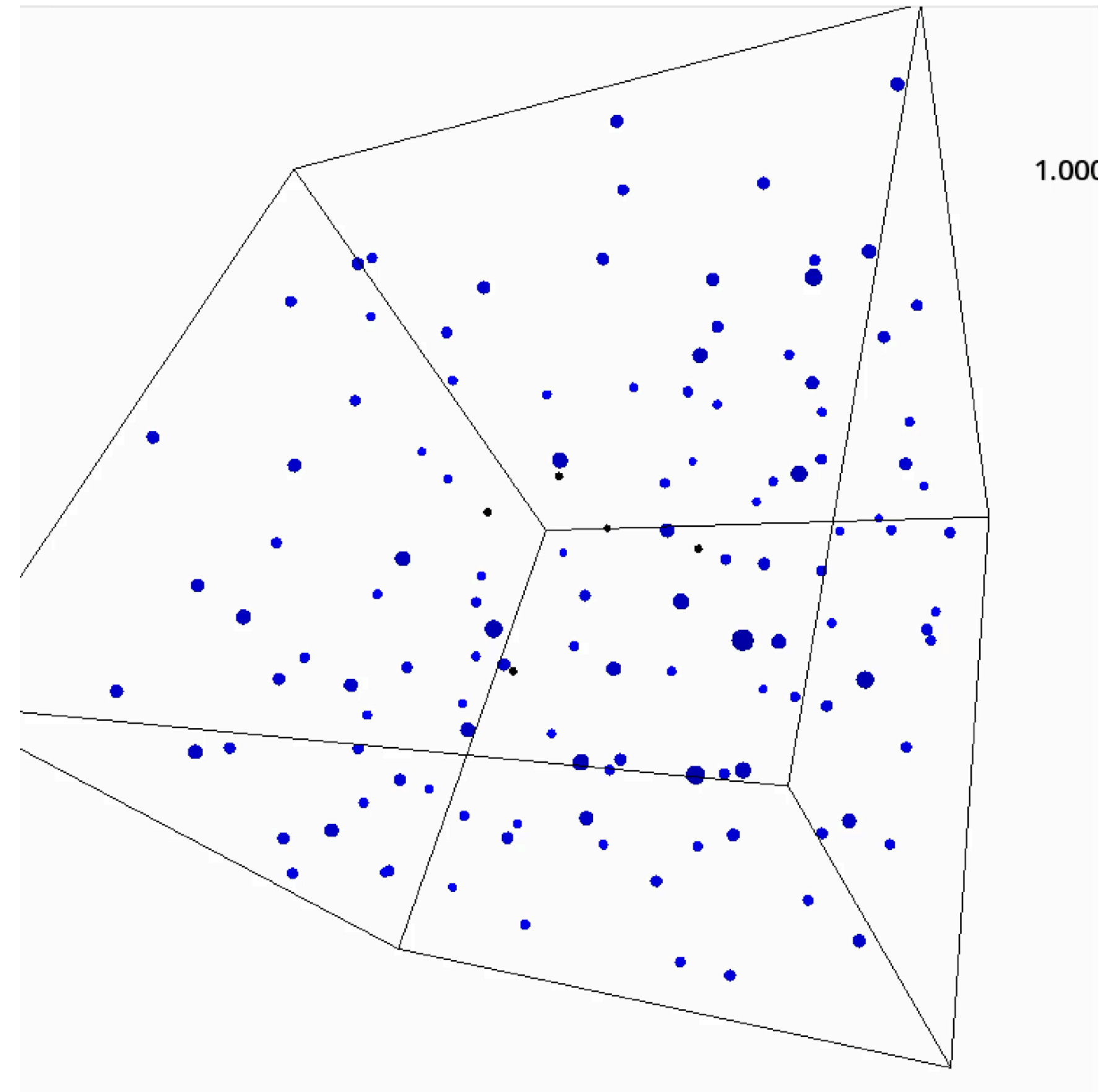
"Statisticians Physicists, like artists, have the bad habit of falling in love with their models."

bigger problems

viii. what does it mean to model?

ix. what does it mean to minimize?

A 38 element Lennard-Jones system has $\sim 10^{14}$ local minima (!!)

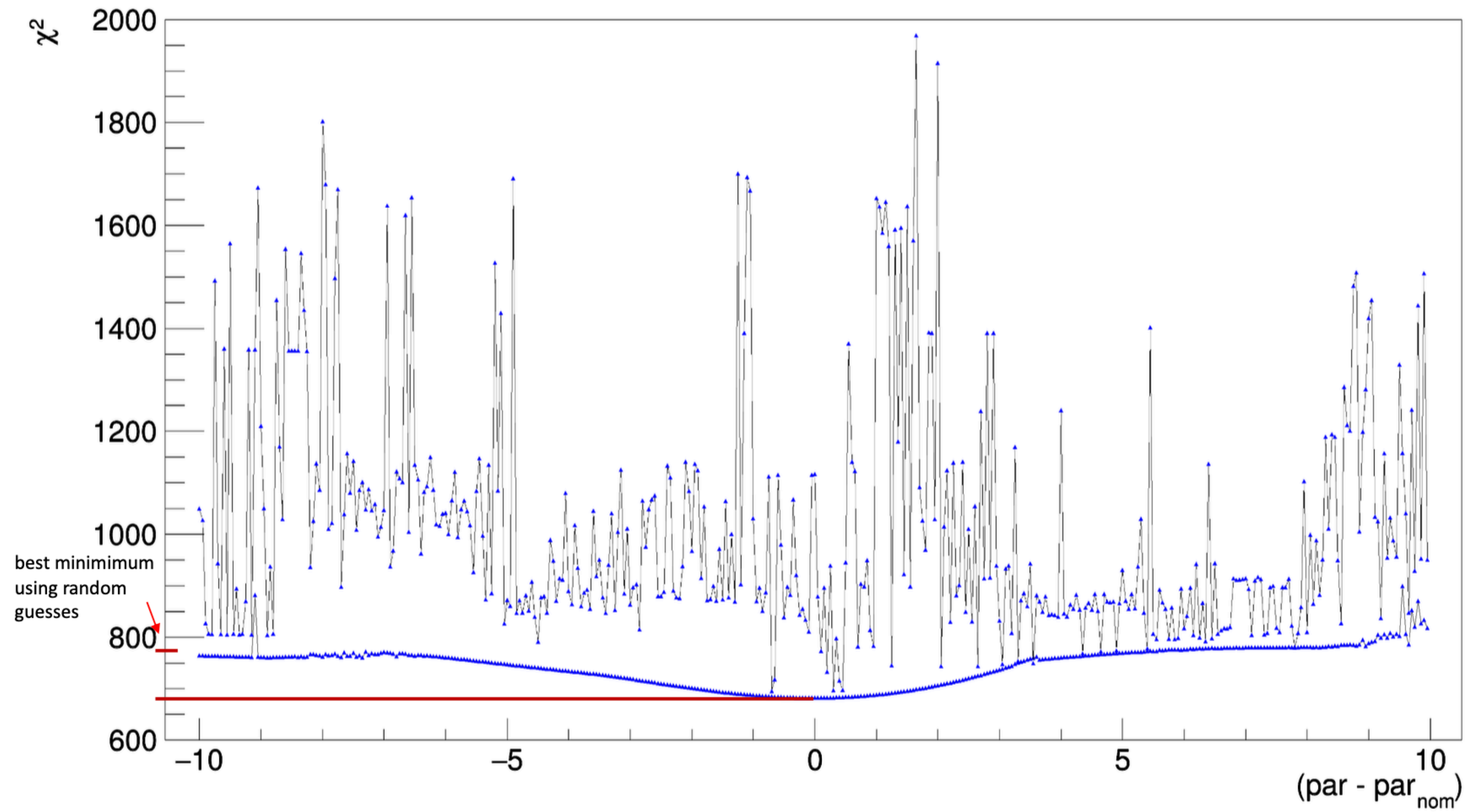


C.J. Tsai and K.D. Jordan, J. Phys Chem, **97** 227 (1993).

bigger problems

viii what does it mean to model?

ix. what does it mean to minimize?



a way forward

(i) seat of the pants method

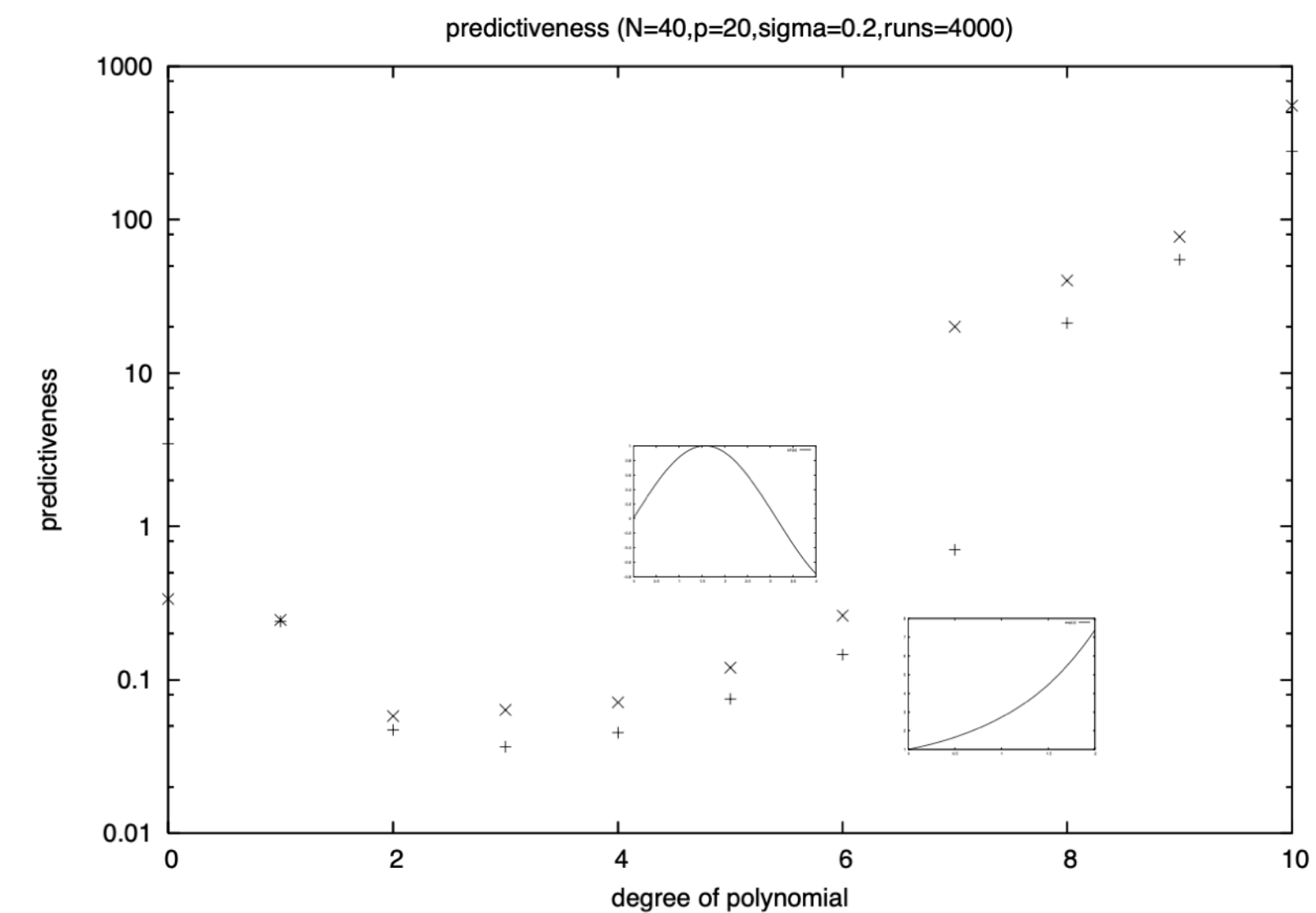
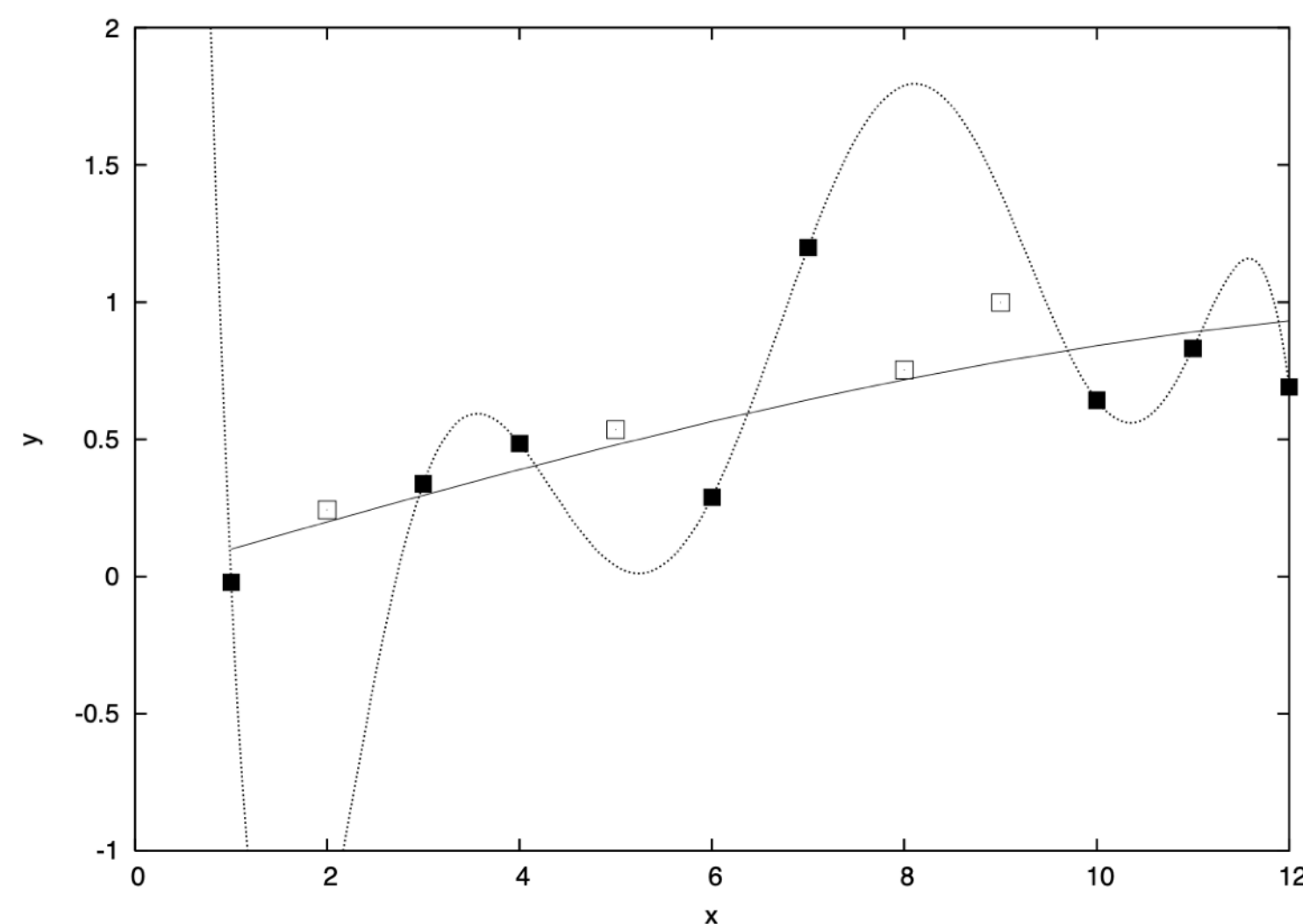
a way forward:

- One way in which one can claim the discovery of a state is if that knowledge permits better, or *predictive*, statements to be made about future experiments.

"How well does my model fit the data?" → "How well can I predict the outcomes of future experiments?"

- We can use cross-validation to avoid overfitting.
- Combine these ideas by splitting the data set into training (ante) and validation (post) sets.

A simple example



Predictiveness

- abandon the idea that we "know" the model -- work in "super-model space" (which, of course, is a model; but now we seek a degree of agnosticism).
- stochastically explore model space. [One could model average by averaging over the trajectory. This is not our primary goal here.]
- explore the continuous portion of model space with Markov chain Monte Carlo. Metropolis-Hastings update according to

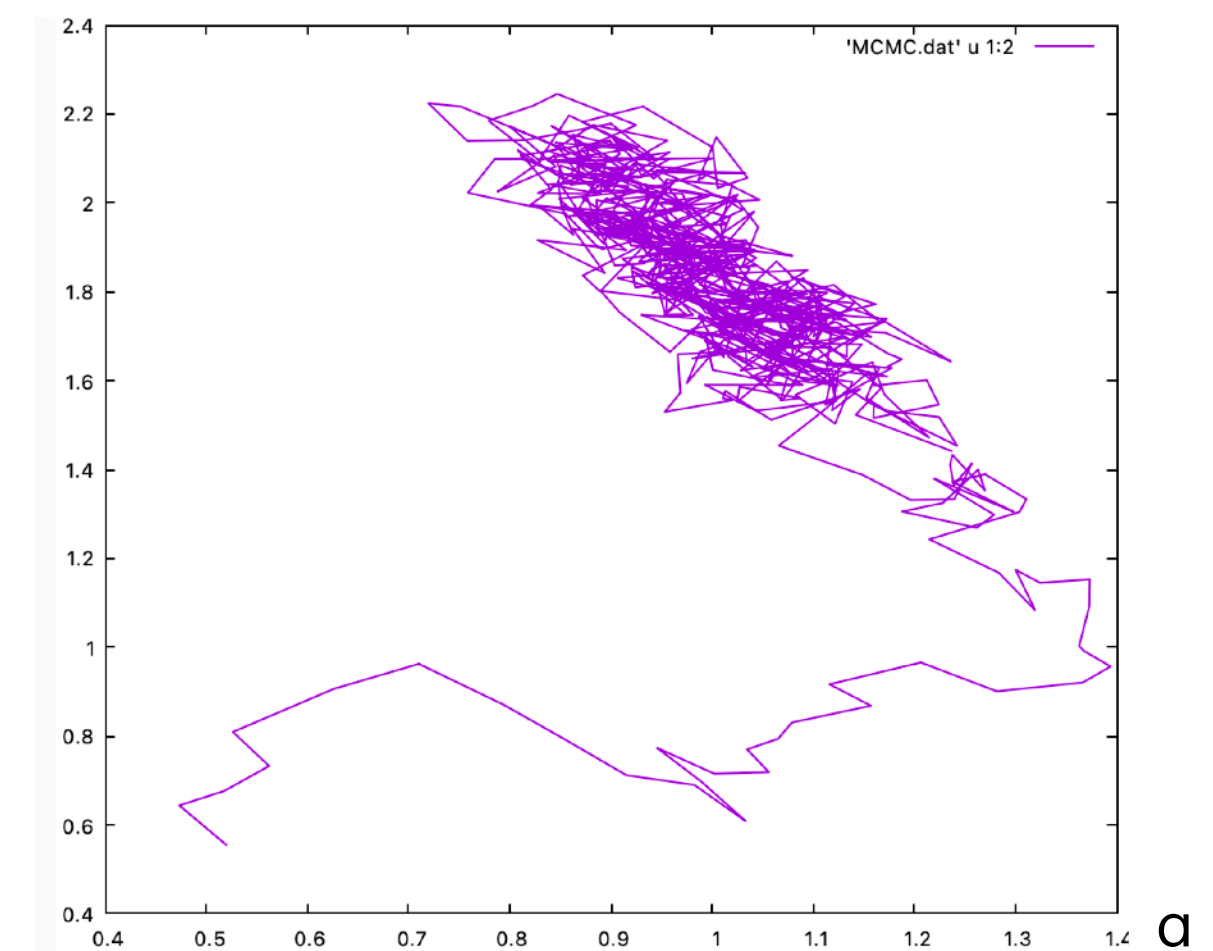
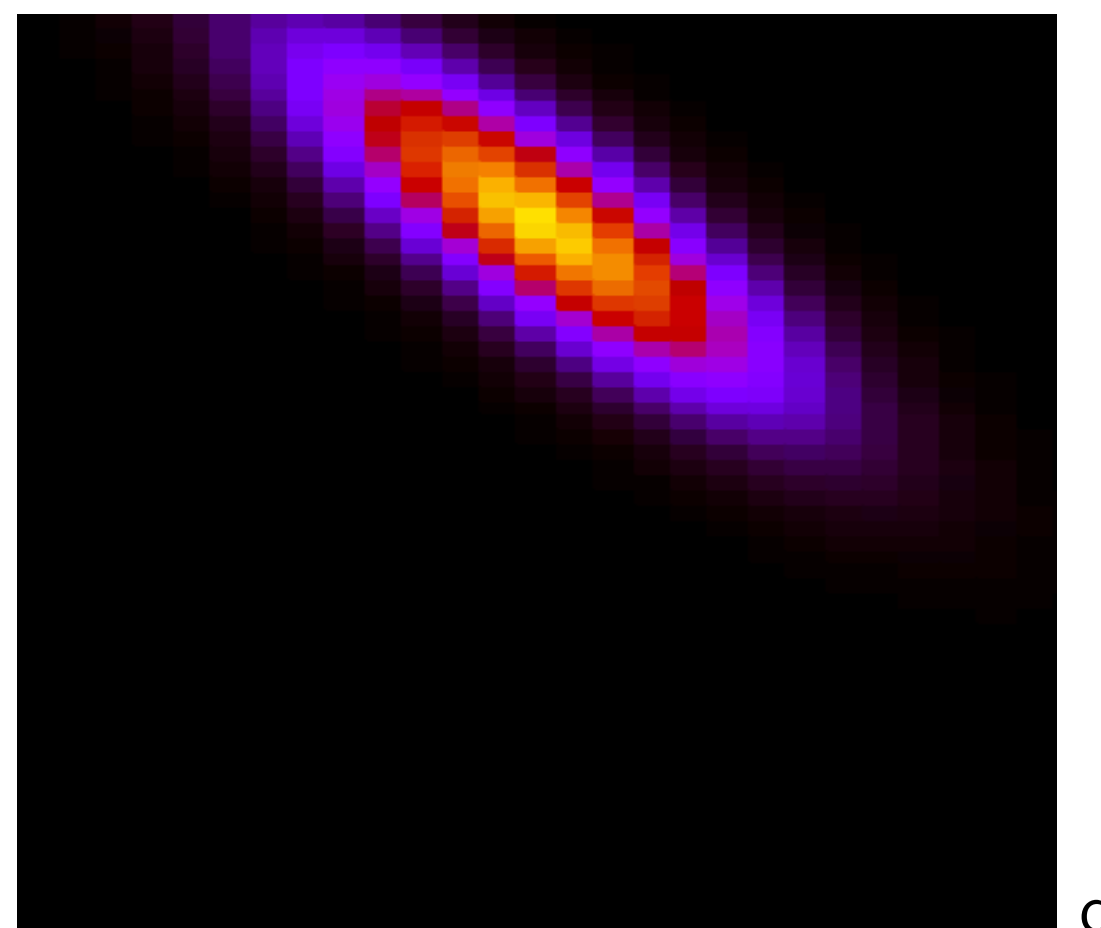
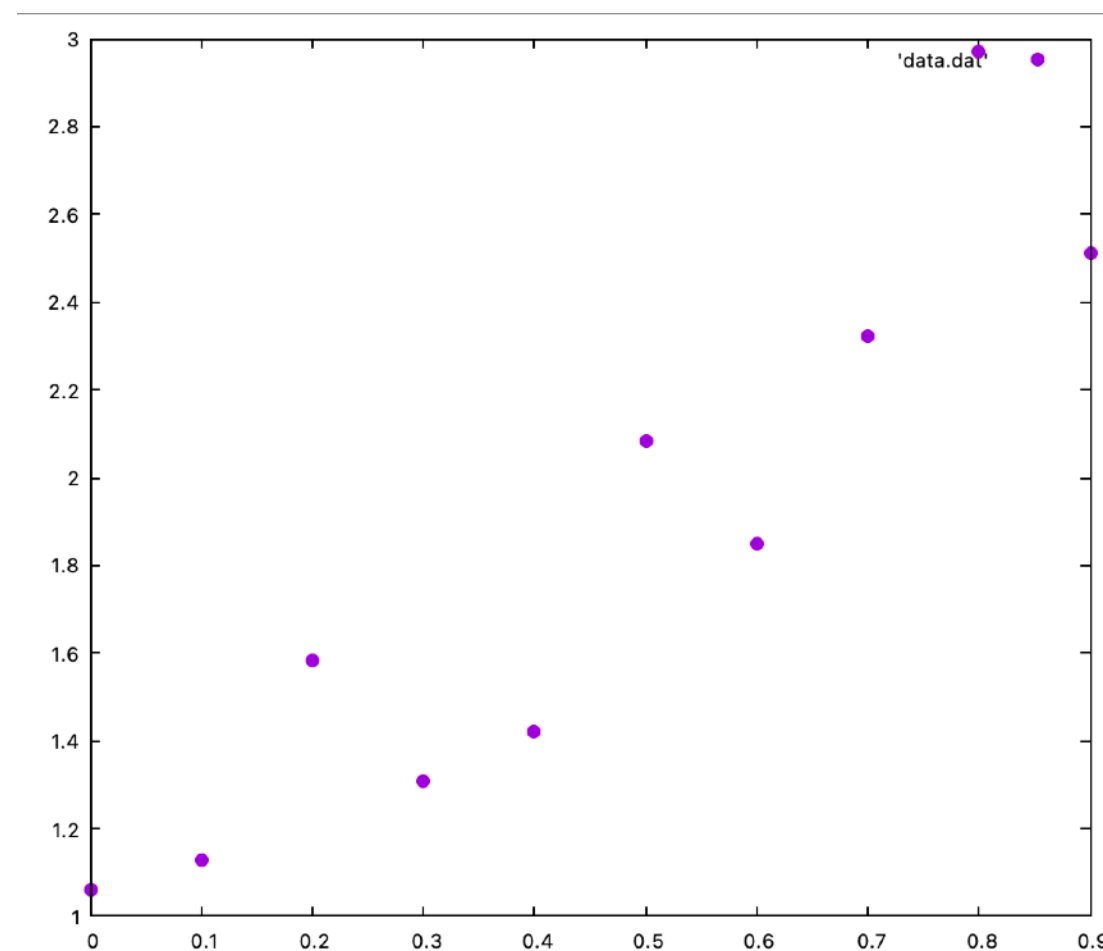
$$p(\theta \rightarrow \theta') = \min \left(1, \frac{p(\theta' | \mathcal{D}) f_{\theta, \theta'}}{p(\theta | \mathcal{D}) f_{\theta', \theta}} \right) \quad p(\theta | \mathcal{D}) = \frac{p(\mathcal{D} | \theta) p(\theta)}{p(\mathcal{D})}$$

1000 MCMC steps starting at (0.5,0.5); uniform step of size $\in [-0.1,0.1]$

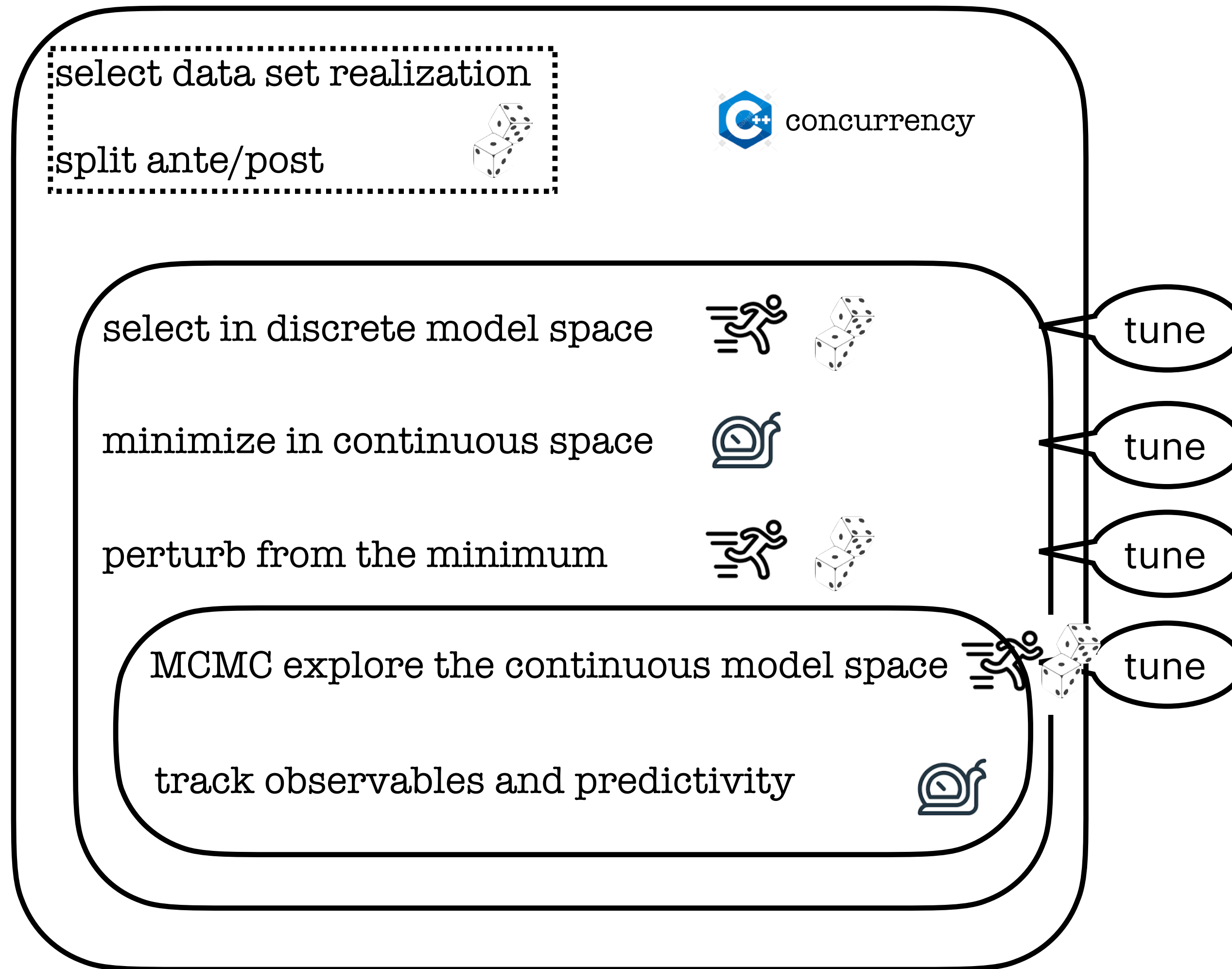
Every point in this space is a model, some are just better than others.

$\mathcal{D}(1 + 2x + \hat{\eta}(0.2))$

scan of the posterior



Predictivity - Algorithm



focus on predictiveness

de-emphasize fitting and fit quality

explore a large model space to enhance the reliability of the conclusions

use data realizations to enhance the reliability of the conclusions

be as agnostic wrt priors and models as possible

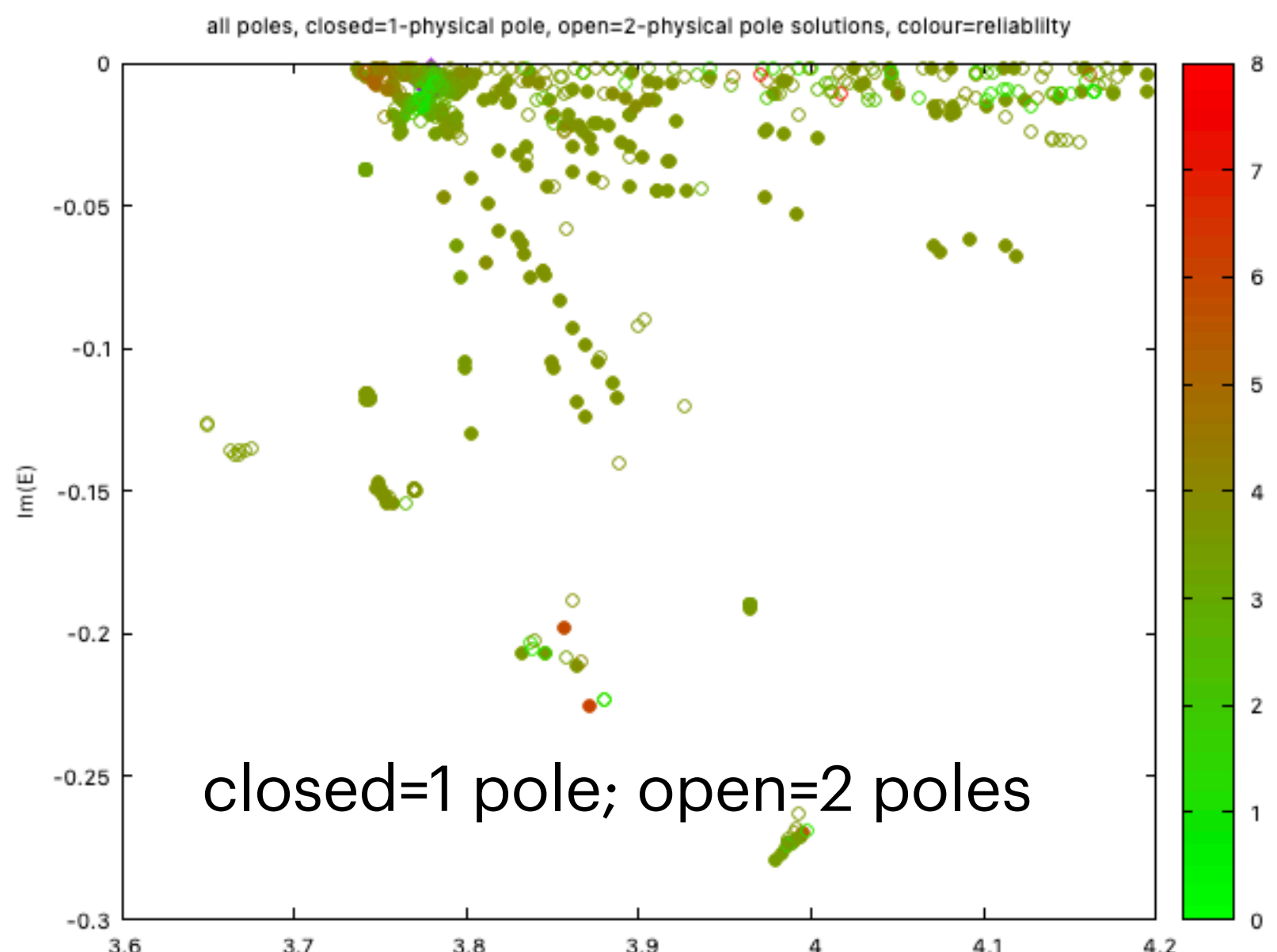
model parameters are not physical

analyze complete set of observables

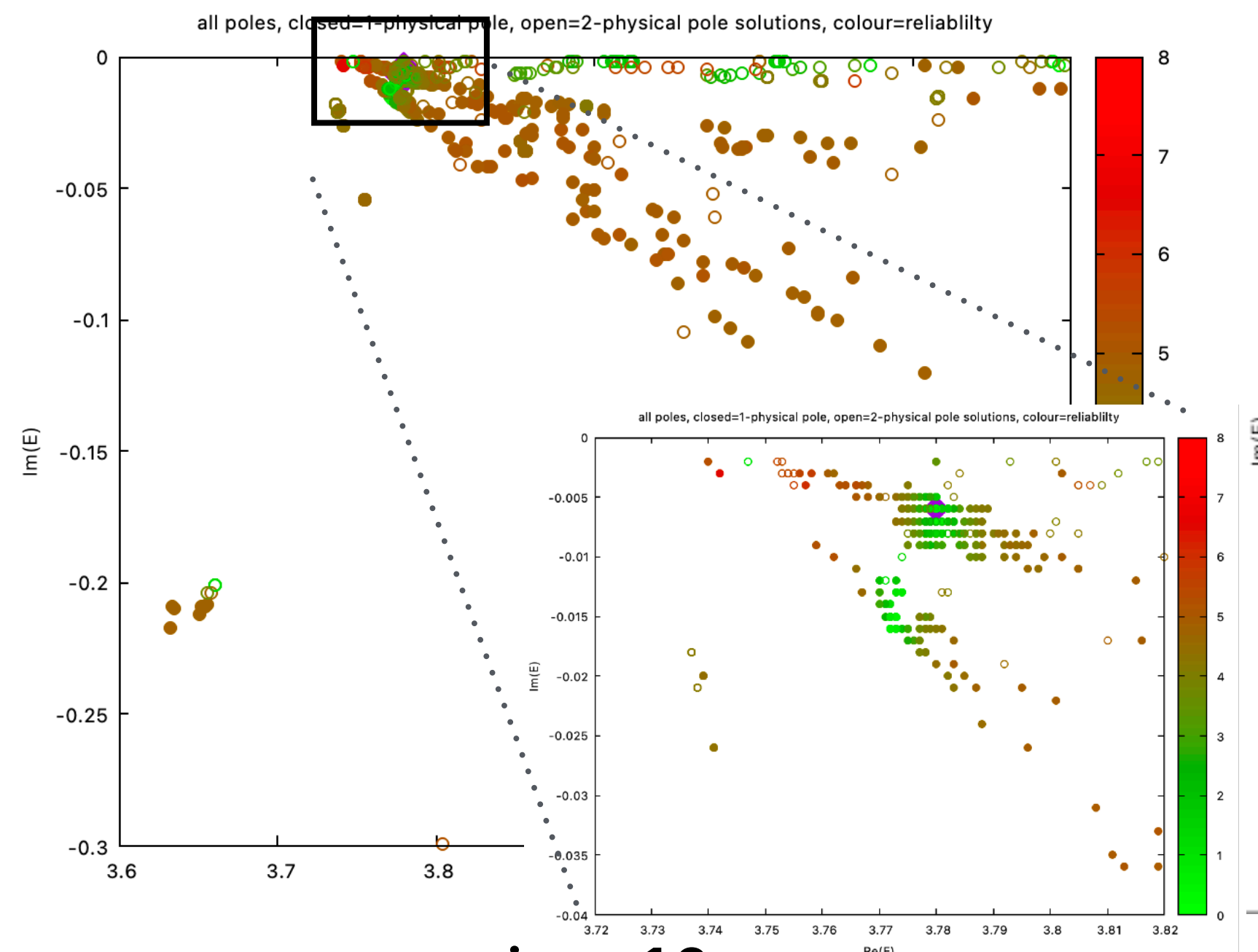
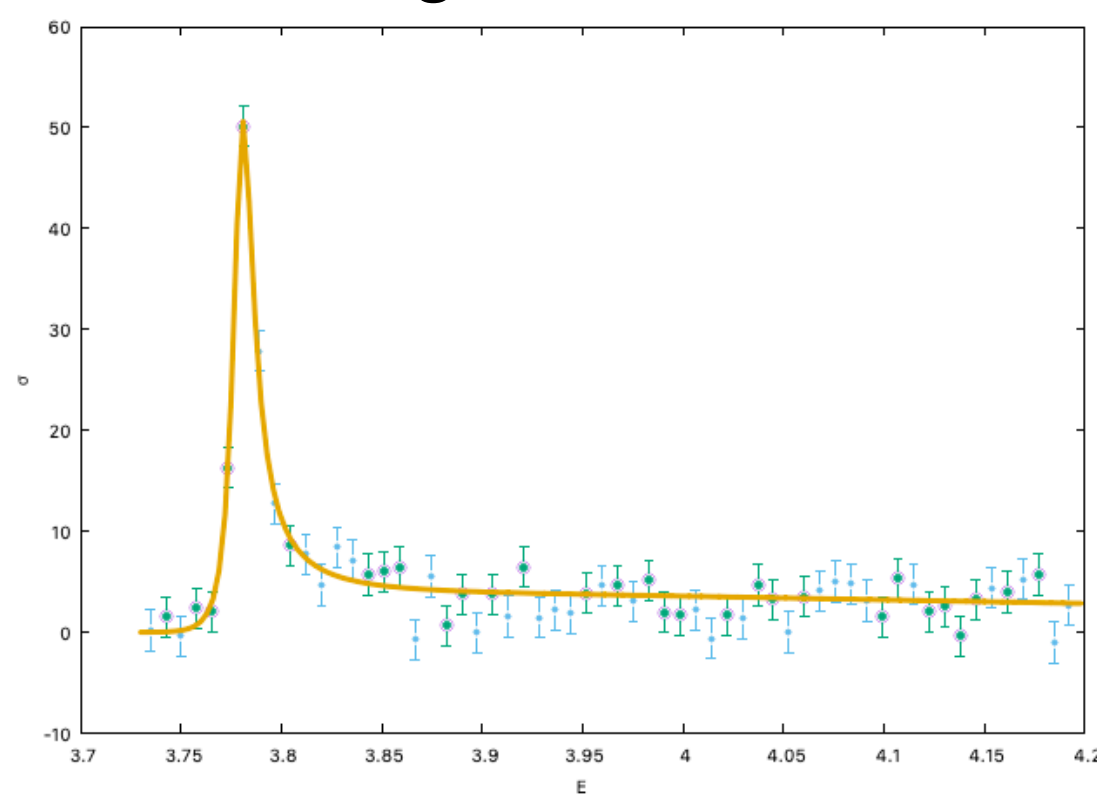
Pole Positions

Model space excludes the generating model.

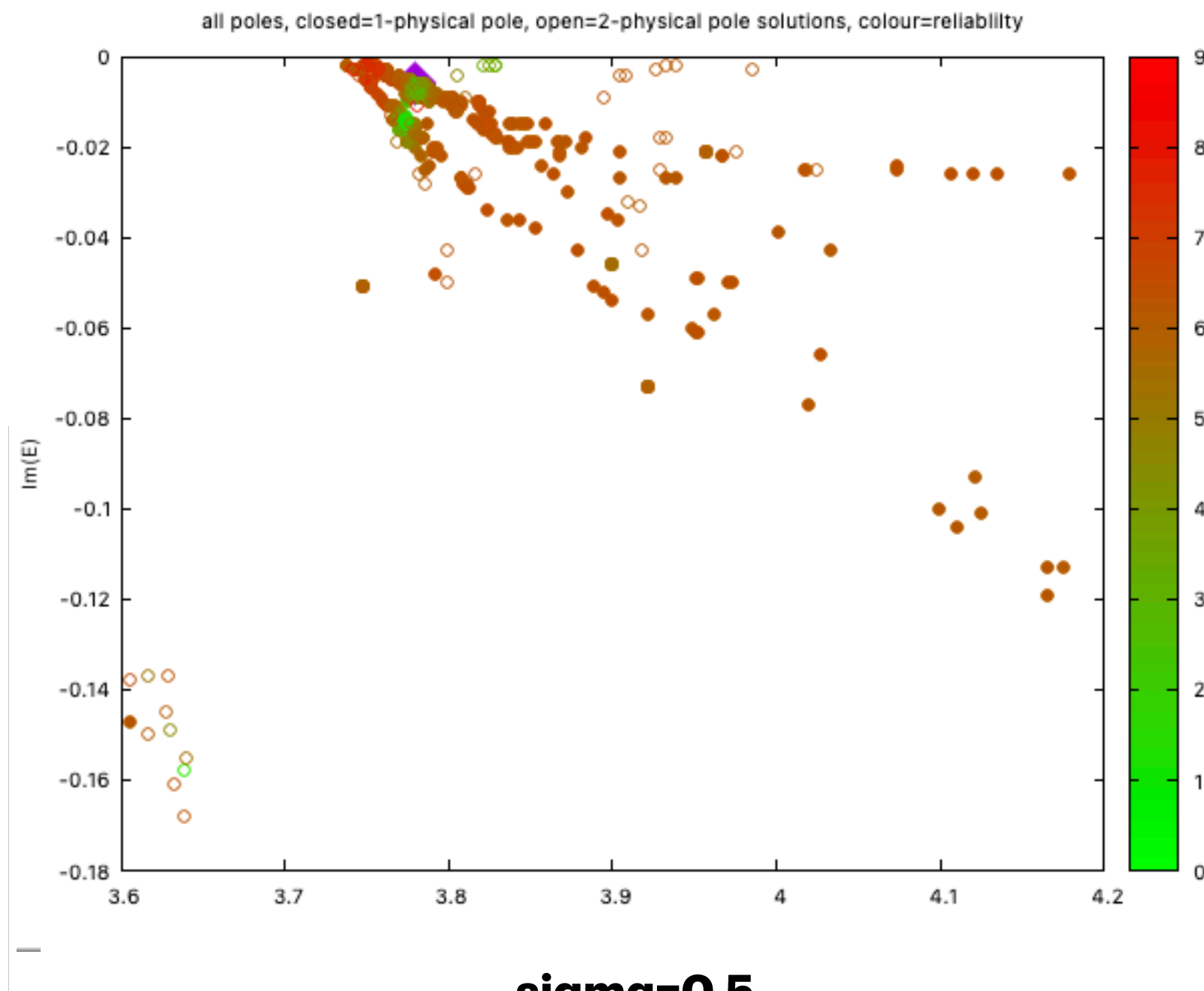
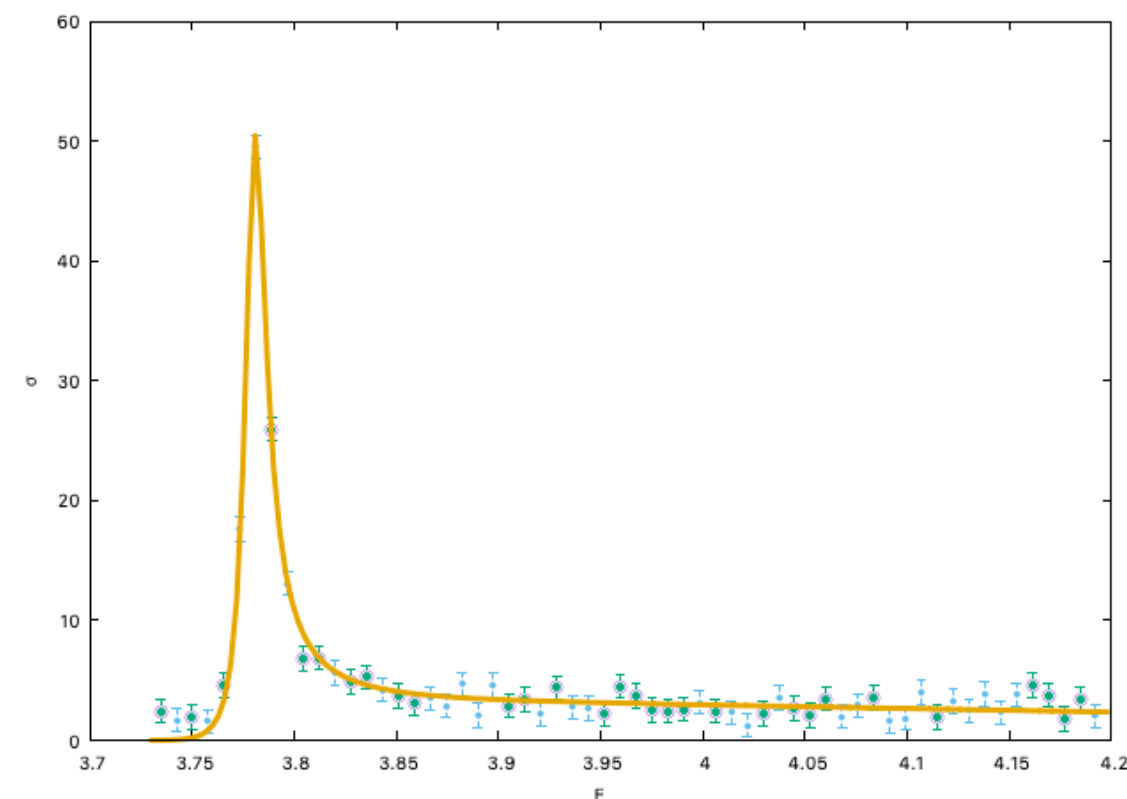
colour = "reliability" := $\text{post}(\text{ante}) \cdot \text{post}(\text{post}) = \exp(-1/2 \text{LP}(a) - 1.2 \text{LP}(p))$. This is $\log(1/2 \text{LP}(a) + 1/2 \text{LP}(p))$



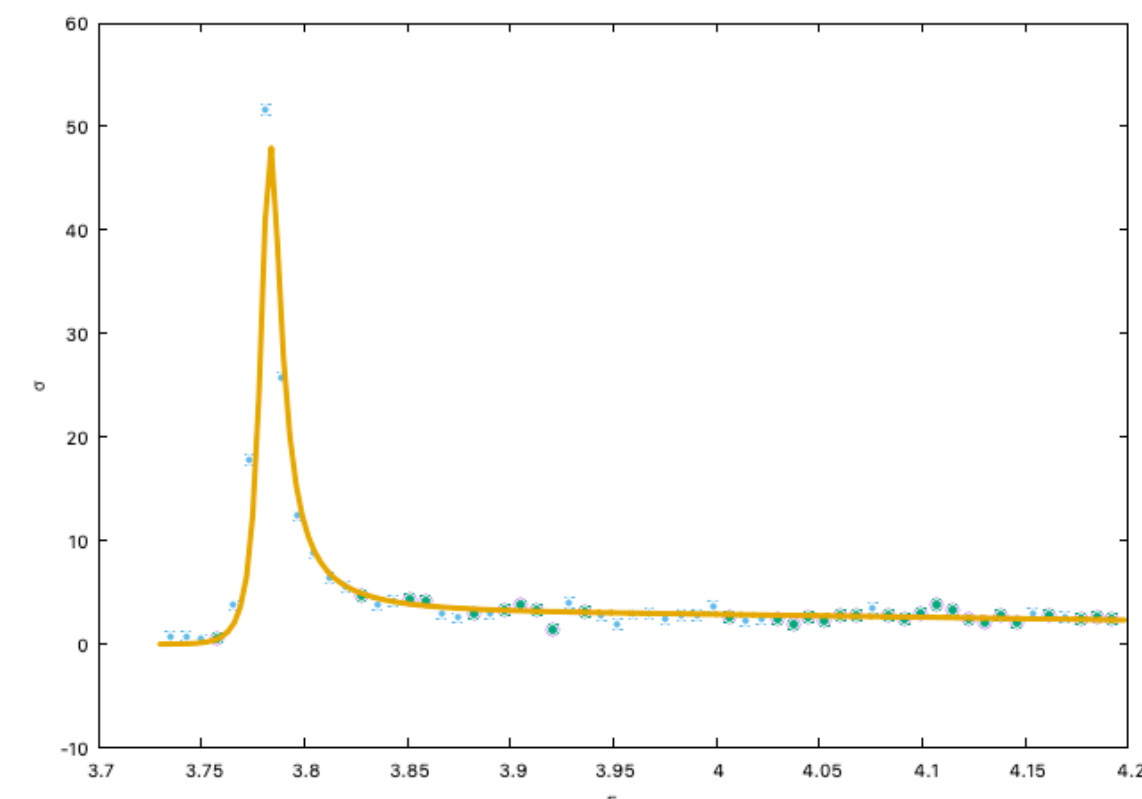
sigma=2.0



sigma=1.0



sigma=0.5



a way forward

(ii) Bayesian inference

Bayesian Inference

$$\begin{aligned}
 p(t|D) &= \sum_M \int d\theta_M \overbrace{p(t|\theta_M, M, D) p(\theta_M|M, D)}^{\text{posterior predictive}} \cdot \int d\theta' \overbrace{p(D|\theta'_M, M) p(\theta'_M|M)}^{\text{likelihood}} \overbrace{p(M)/p(D)}^{\text{evidence}} \\
 &\rightarrow p(D|\hat{\theta}_M, M) p(\hat{\theta}_M|M) \underbrace{(2\pi)^{d/2} \sqrt{\det \Sigma}}_{\text{Occam factor}} p(M)/p(D)
 \end{aligned}$$

Bayes theorem



marginalize, implements Occam's razor



Laplace approximation (steepest descent)

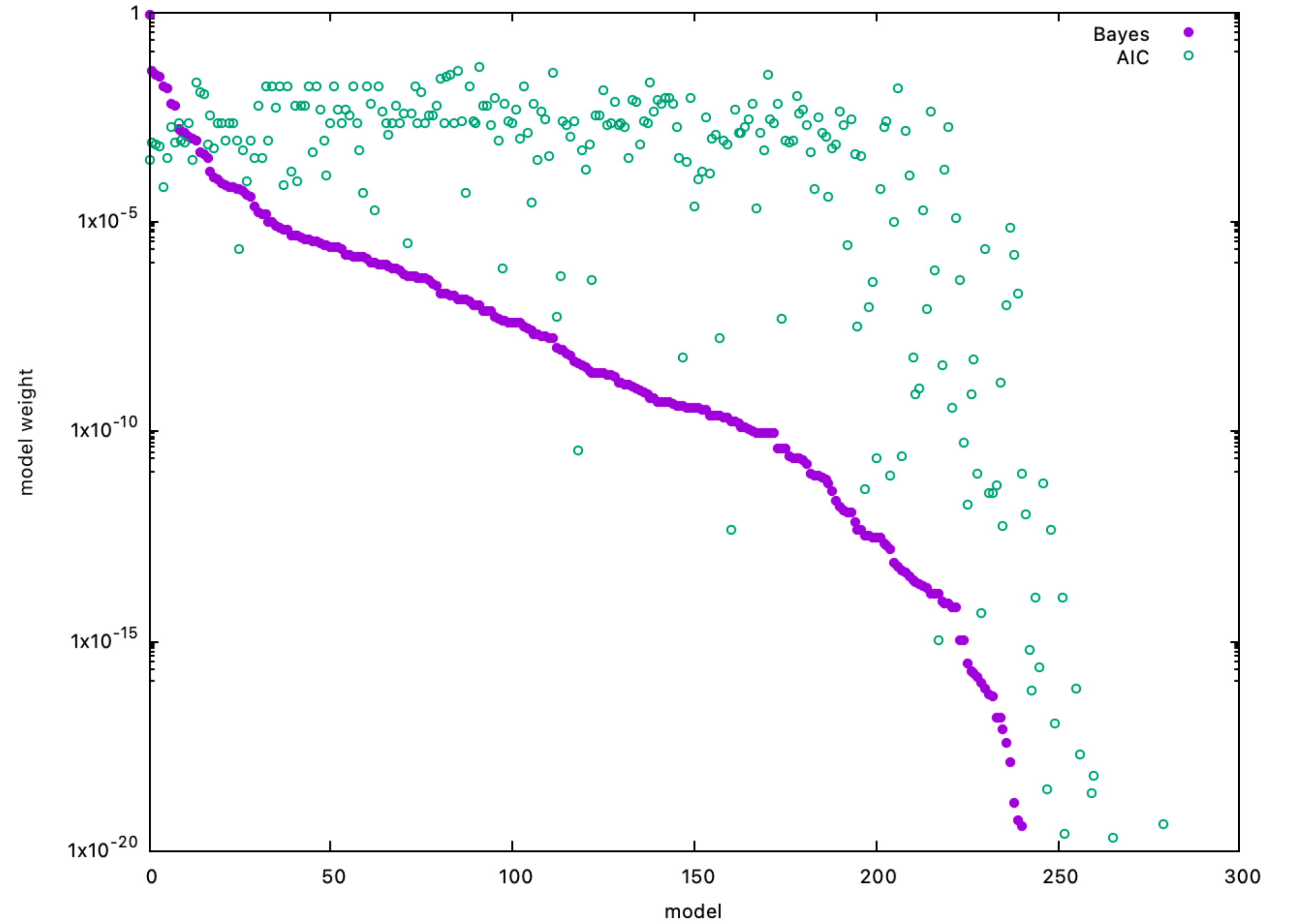


No need to bootstrap data

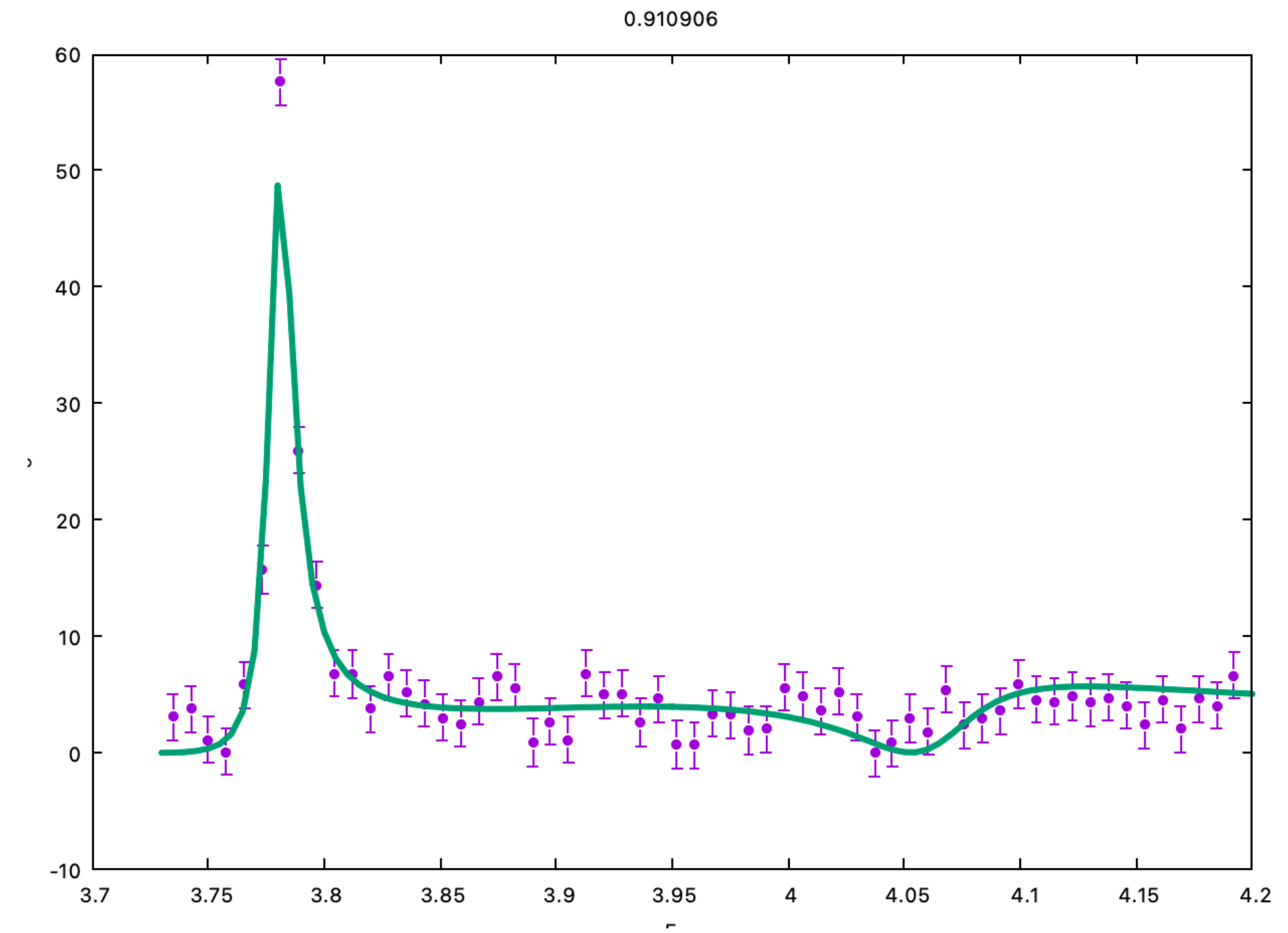
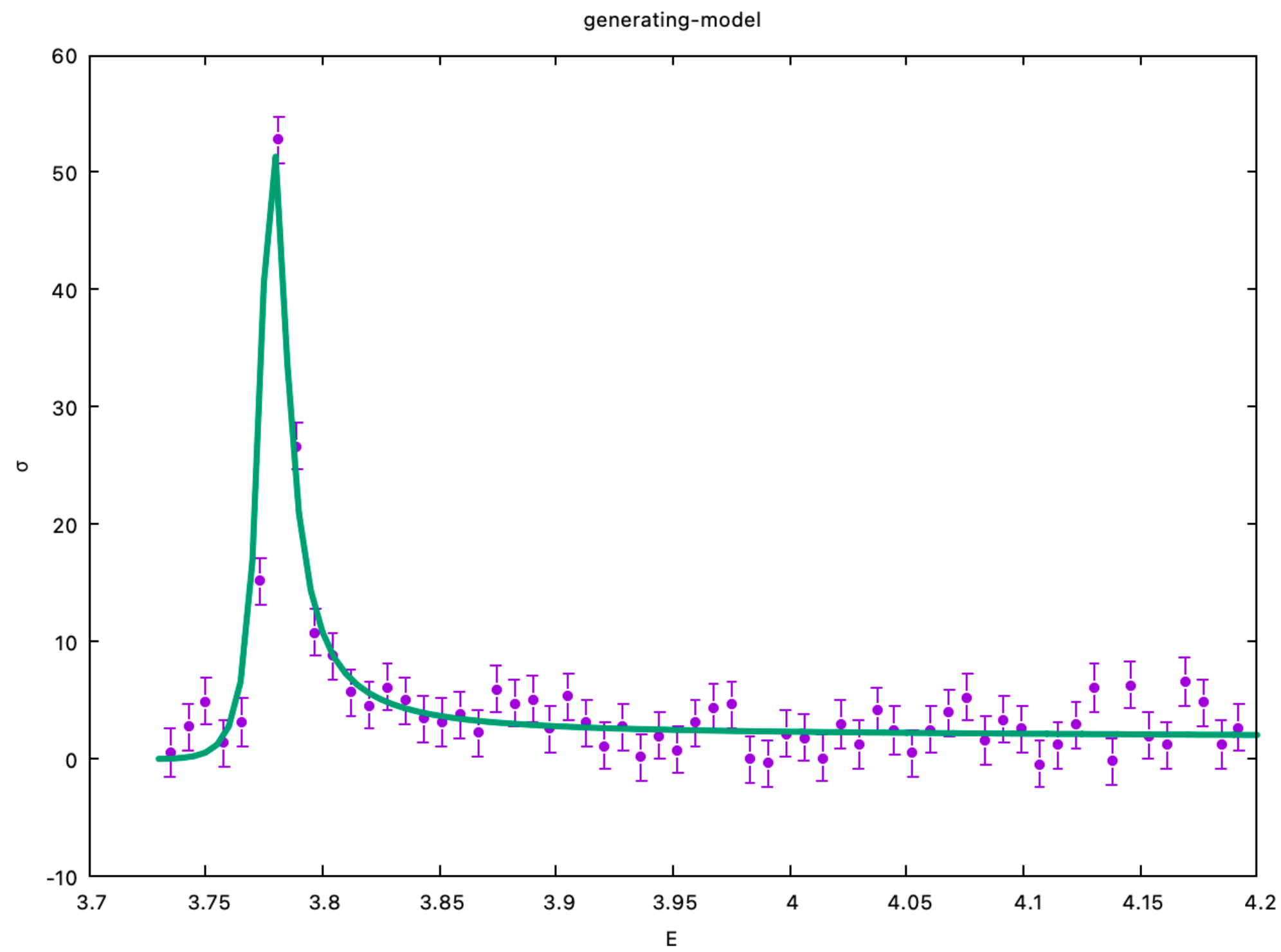
No need to explicitly penalize model complexity

Single Channel K-matrix

compare Bayes and AIC weighting for 1000 models



Single Channel K-matrix



Single Channel K-matrix

generating-model poles:
pole = (3.778, -0.007) residue = (-15.4954, 6.44069)

Bayes pole count probabilities

0 0%
1 99.2358%
2 0.59453%
3 0%
4 0%

Bayes pole count probabilities (no sheet I poles)

0 0%
1 96.1141%
2 0%
3 0%
4 0%

<pole>(1) = (3.778, -0.007) +/- (delR, delI, delRI) 0 0 4.5139e-36
<res>(1) = (-14.6995, 6.29522) +/- (delR, delI, delRI) 1.09651 0.277913 0.0706413

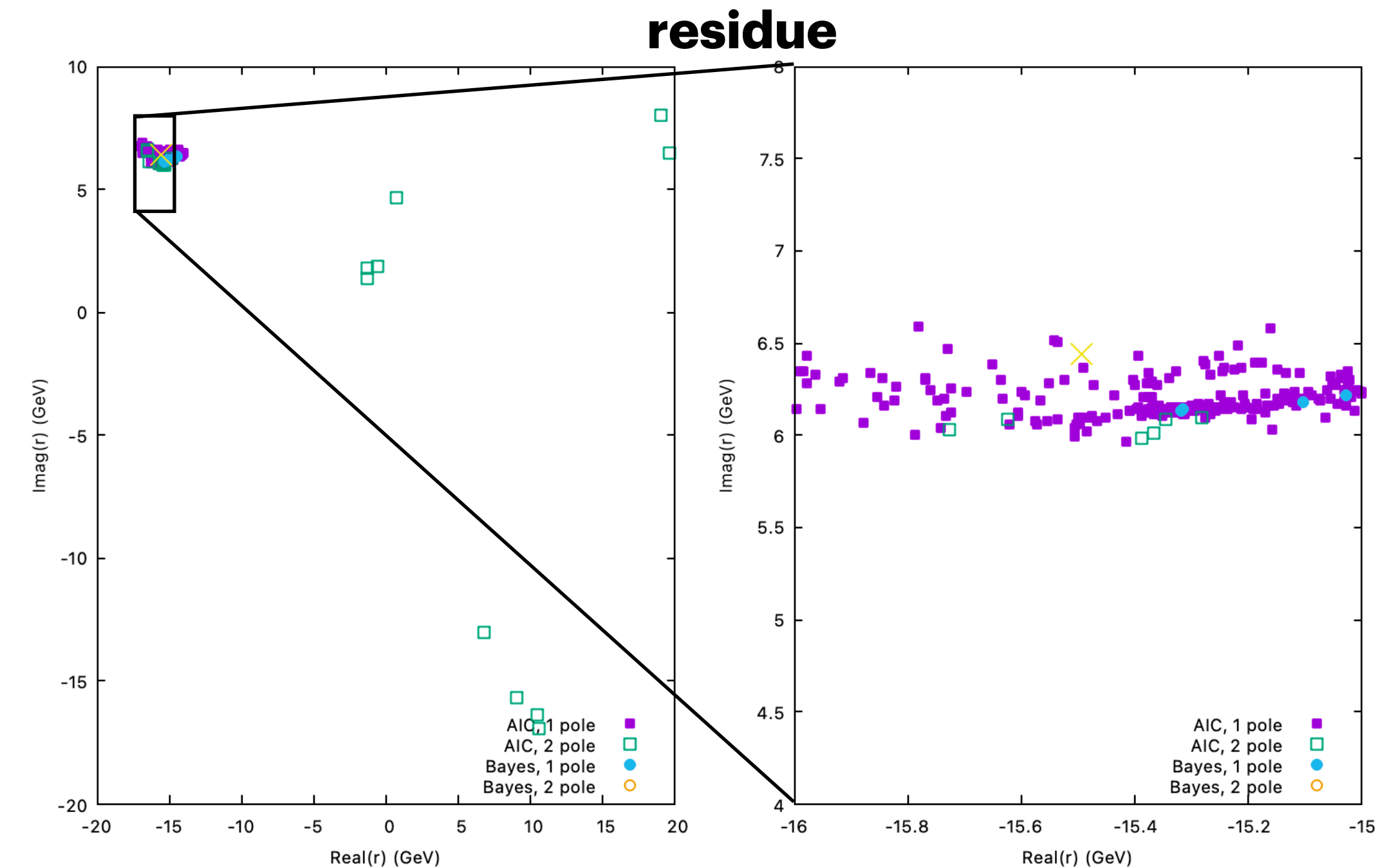
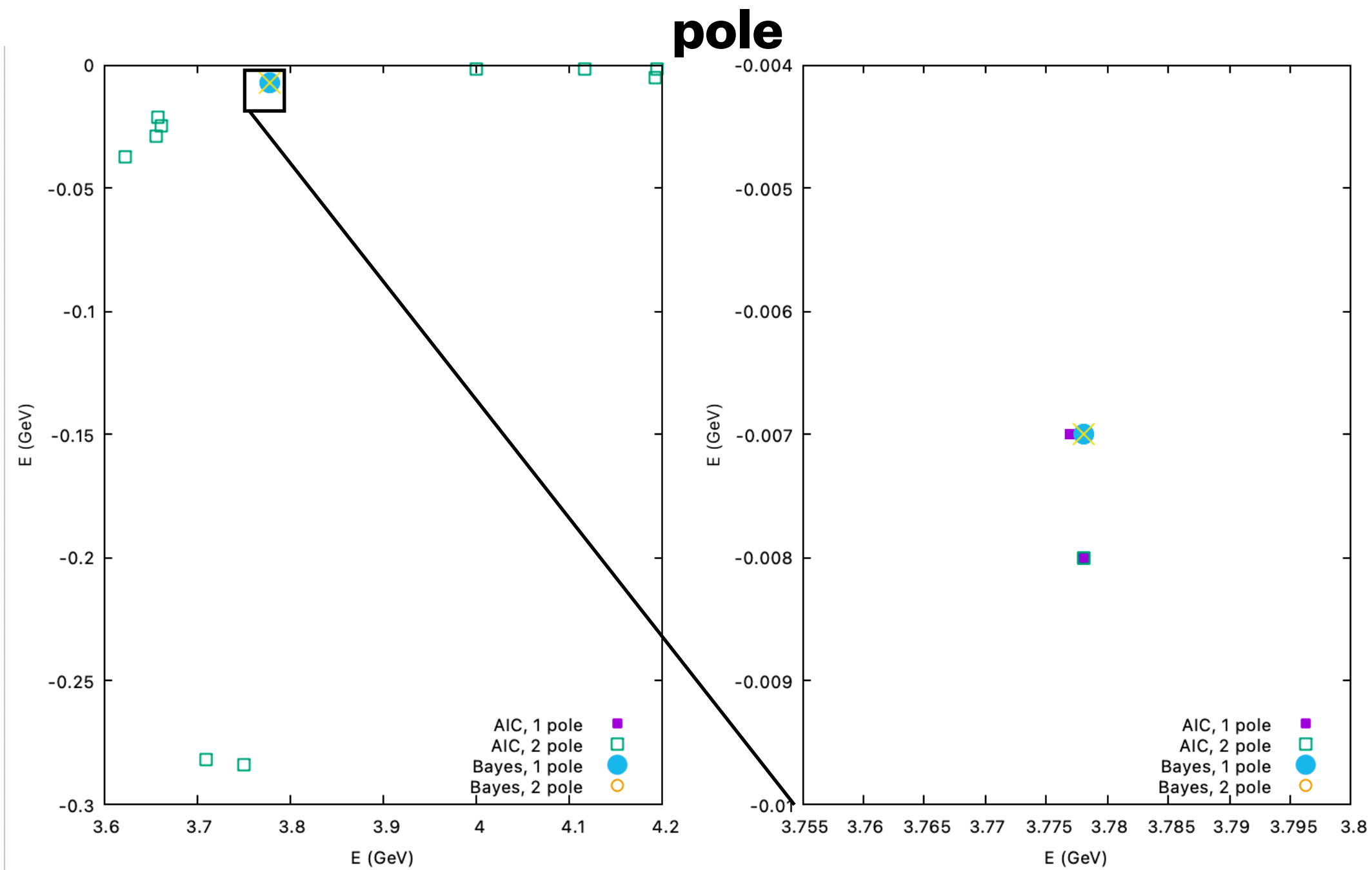
AIC pole count probabilities

0 0%
1 94.3365%
2 2.60253%
3 0.100984%
4 0%

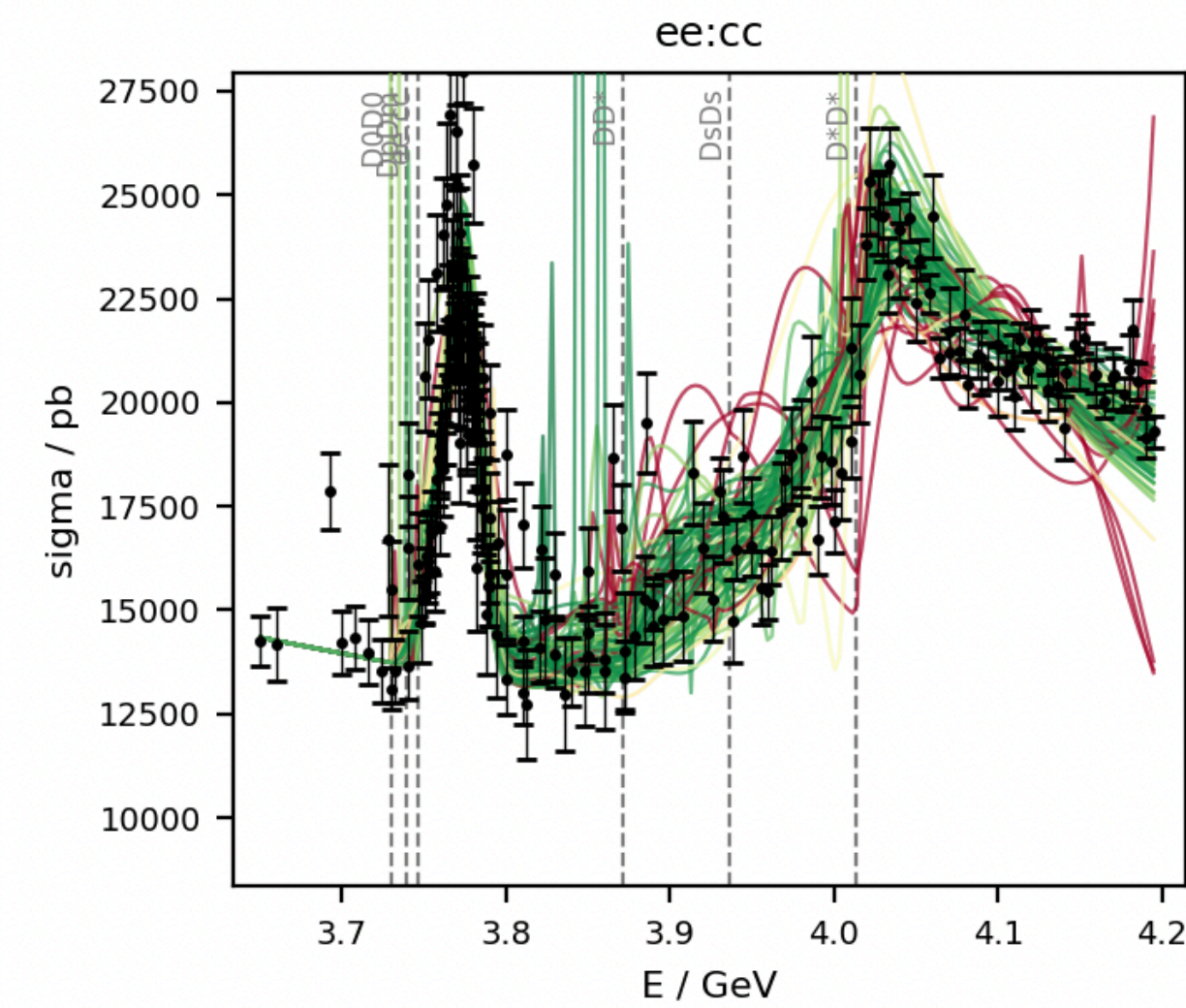
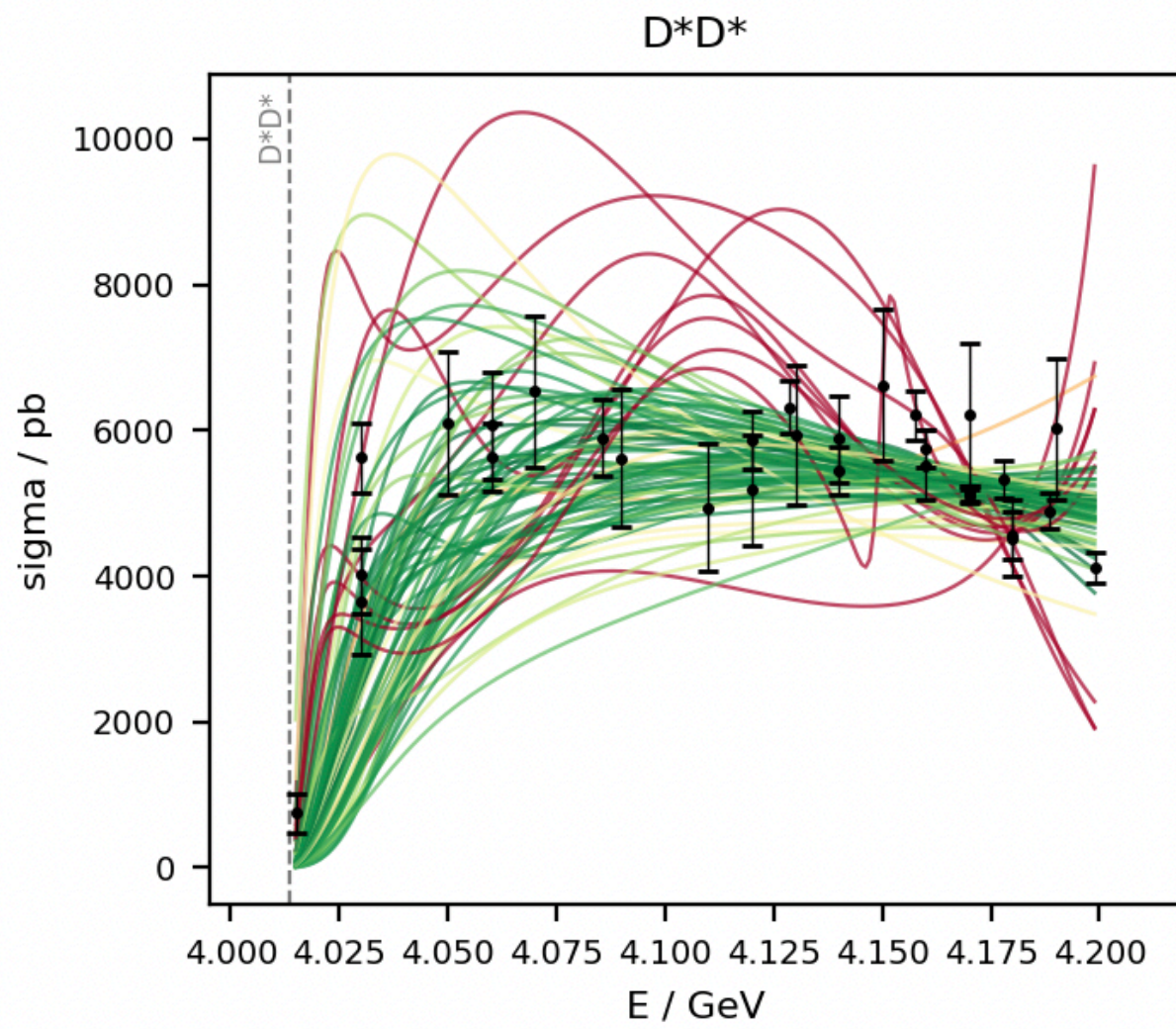
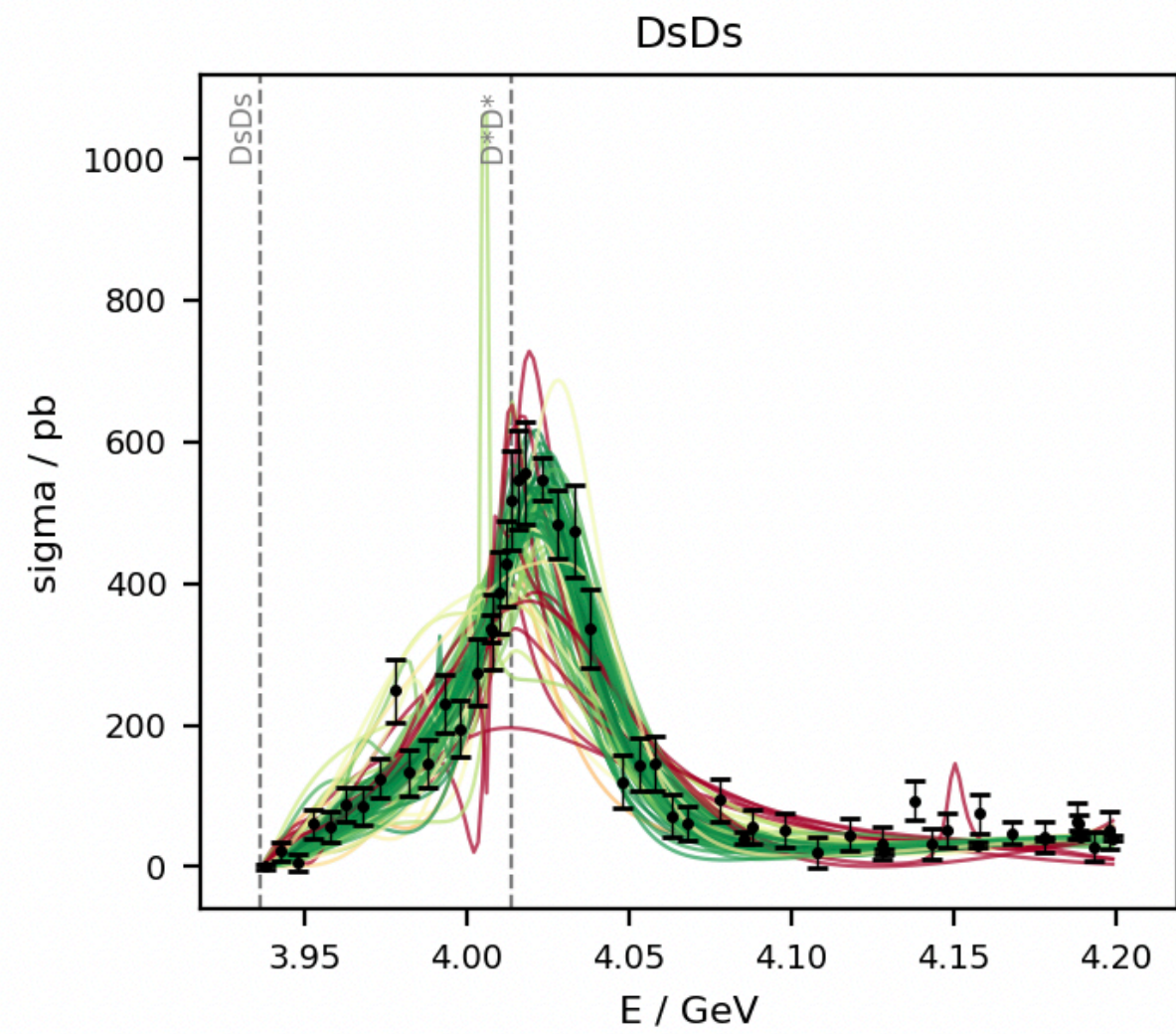
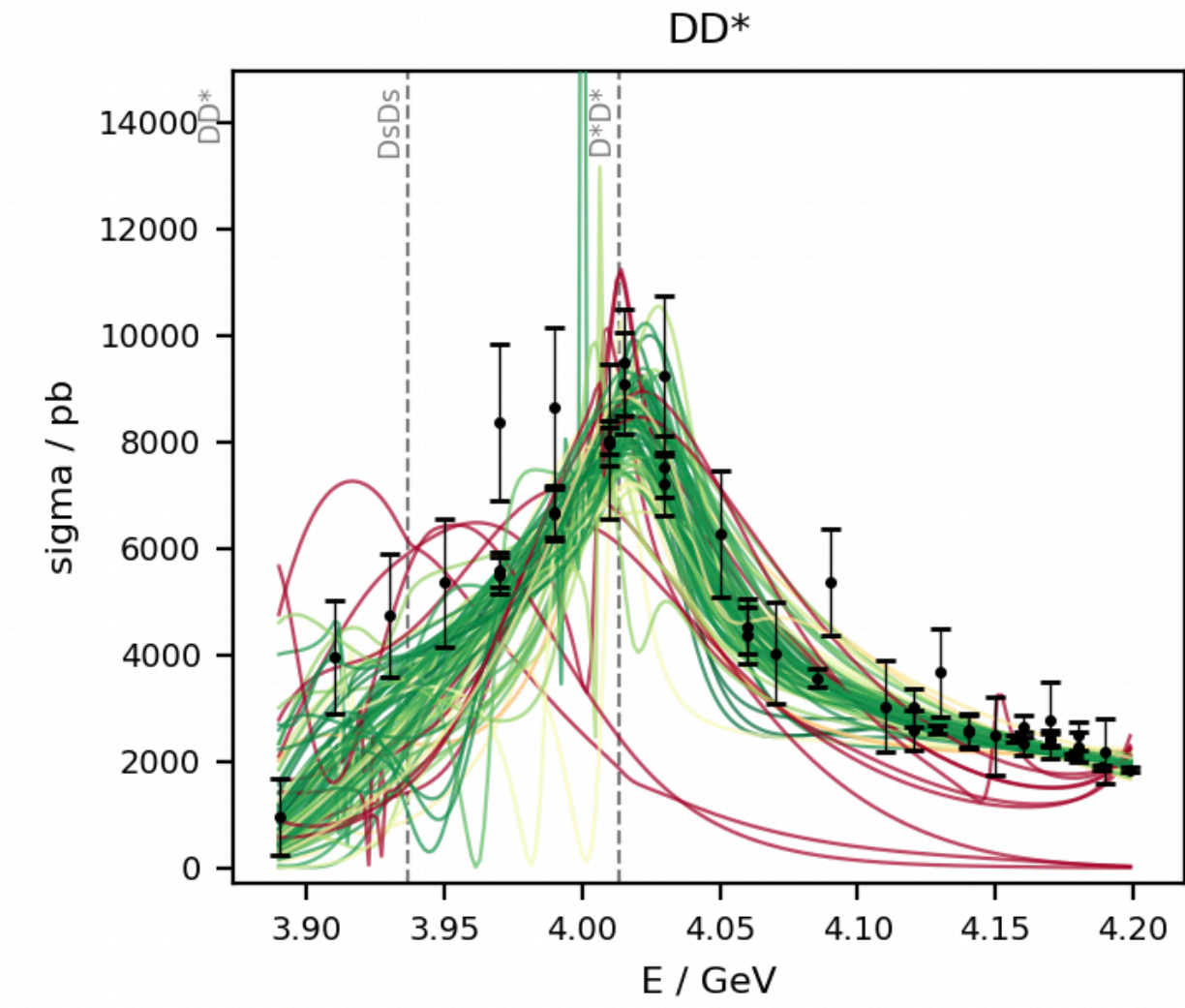
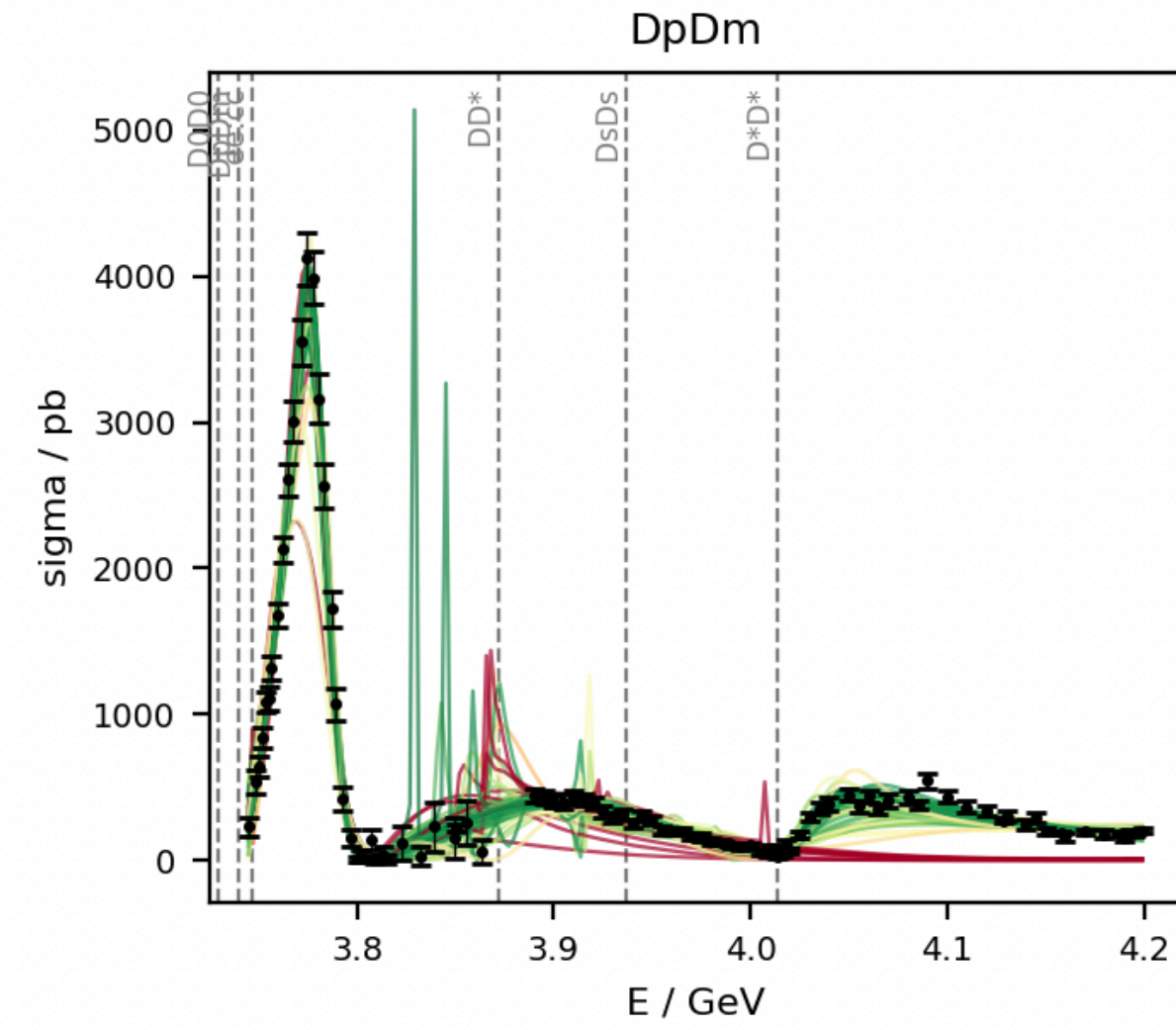
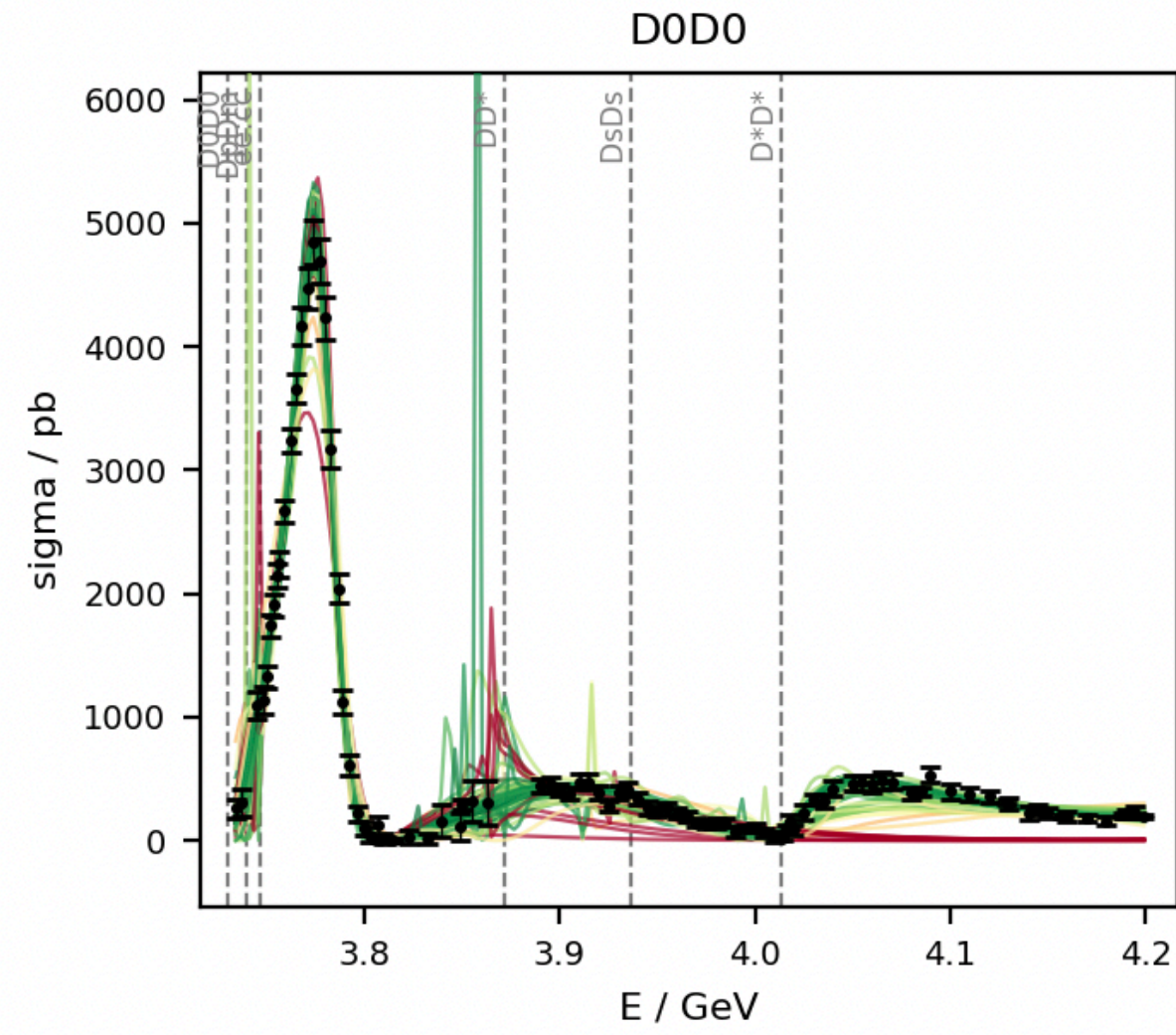
AIC pole count probabilities (no sheet I poles)

0 0%
1 77.6579%
2 0.935699%
3 0.0287494%
4 0%

<pole>(1) = (3.77798, -0.00700587) +/- (delR, delI, delRI) 1.36091e-05 -1.7134e-07 6.92959e-06
<res>(1) = (-15.3122, 6.24651) +/- (delR, delI, delRI) 70.7888 -1.24645 7.54011



PIKE5 v0.2d6 all 81 fits



G(3900)

problems:

fitting is extraordinarily difficult

LASSO is little help

pole identification can be tricky

Laplace approximation is often not good; the Hessian condition number is very large (indicating a nonGaussian minimum)

attempts at Monte Carlo methods are bedeviled by very peaky multimodal space (a random step raises χ^2 by many orders of magnitude)

summary

- focus on predictiveness
- de-emphasize fitting and fit quality
- explore a large model space to enhance the reliability of the conclusions (we need to admit model uncertainty!)
- use data realizations to enhance the reliability of the conclusions
- be as agnostic wrt priors and models as possible
- model parameters are not physical
- problems wrt model optimization and finding global minima are reduced.

application to unbinned data?

make models completely agnostic with NNs? Cf, use a VAE to model the posterior...

+ ÆRIC MEC HEHT GEWYRCAN

