

#### IHEP EPD Seminar (Jan 9th, 2020)

# Long-lived Particles in Standard Model and Beyond

<sup>1</sup> Argonne National Laboratory

OF Office of Science

Argonne Argonational Laboratory



# LHC and ATLAS and CMS



CMS

# **Geneva Airport** CERN

# Physics Programs in ATLAS and CMS

ATLAS Exotics Searches* - 95% CL Upper Exclusion Limits Status: May 2019									
	Model	<i>ℓ</i> , γ	<b>Jets</b> †	$E_{T}^{miss}$	∫£ dt[fb	<sup>-1</sup> ] Limit	<i>j2.ut</i> = (c	5.2 - 139/10	Reference
Extra dimensions	ADD $G_{KK} + g/q$ ADD non-resonant $\gamma\gamma$ ADD QBH ADD BH high $\sum p_T$ ADD BH multijet RS1 $G_{KK} \rightarrow \gamma\gamma$ Bulk RS $G_{KK} \rightarrow WW/ZZ$ Bulk RS $G_{KK} \rightarrow WW \rightarrow qqqq$ Bulk RS $g_{KK} \rightarrow tt$ 2UED / RPP	$\begin{array}{c} 0 \ e, \mu \\ 2 \ \gamma \\ - \\ \geq 1 \ e, \mu \\ - \\ 2 \ \gamma \\ multi-channe \\ 0 \ e, \mu \\ 1 \ e, \mu \\ 1 \ e, \mu \end{array}$	$1 - 4 j$ $-$ $2 j$ $\geq 2 j$ $\geq 3 j$ $-$ el $2 J$ $\geq 1 b, \geq 1 J,$ $\geq 2 b, \geq 3$	Yes - - - - - /2j Yes j Yes	36.1 36.7 37.0 3.2 3.6 36.7 36.1 139 36.1 36.1	M <sub>D</sub> M           Ms         M           Mth         M           Mth         M           Mth         M           Mth         M           GKK mass         4.1 TeV           GKK mass         2.3 TeV           GKK mass         1.6 TeV           gKK mass         3.8 TeV           KK mass         1.8 TeV	7.7 TeV 8.6 TeV 8.9 TeV 8.2 TeV 9.55 TeV	$\begin{split} n &= 2 \\ n &= 3 \text{ HLZ NLO} \\ n &= 6 \\ n &= 6, M_D = 3 \text{ TeV, rot BH} \\ n &= 6, M_D = 3 \text{ TeV, rot BH} \\ k/\overline{M}_{Pl} &= 0.1 \\ k/\overline{M}_{Pl} &= 1.0 \\ k/\overline{M}_{Pl} &= 1.0 \\ \Gamma/m &= 15\% \\ \text{Tier (1,1), } \mathcal{B}(A^{(1,1)} \rightarrow tt) = 1 \end{split}$	1711.03301 1707.04147 1703.09127 1606.02265 1512.02586 1707.04147 1808.02380 ATLAS-CONF-2019-003 1804.10823 1803.09678
Gauge bosons	$\begin{array}{l} \mathrm{SSM}\ Z' \to \ell\ell \\ \mathrm{SSM}\ Z' \to \tau\tau \\ \mathrm{Leptophobic}\ Z' \to bb \\ \mathrm{Leptophobic}\ Z' \to tt \\ \mathrm{SSM}\ W' \to \ell\nu \\ \mathrm{SSM}\ W' \to \tau\nu \\ \mathrm{HVT}\ V' \to WZ \to qqqq \ \mathrm{model}\ \mathrm{B} \\ \mathrm{HVT}\ V' \to WH/ZH \ \mathrm{model}\ \mathrm{B} \\ \mathrm{LRSM}\ W_R \to tb \\ \mathrm{LRSM}\ W_R \to \mu N_R \end{array}$	$\begin{array}{c} 2 \ e, \mu \\ 2 \ \tau \\ - \\ 1 \ e, \mu \\ 1 \ e, \mu \\ 1 \ \tau \\ 0 \ e, \mu \\ \end{array}$ multi-channe 2 \mu \\ \end{array}	_ 2 b ≥ 1 b, ≥ 1J, _ 2 J el el 1 J	– – Yes Yes –	139 36.1 36.1 139 36.1 139 36.1 36.1 36.1 80	Z' mass       5.1 TeV         Z' mass       2.42 TeV         Z' mass       2.1 TeV         Z' mass       3.0 TeV         W' mass       6.0 T         W' mass       3.7 TeV         V' mass       3.6 TeV         V' mass       3.6 TeV         V' mass       3.6 TeV         V' mass       3.6 TeV         V' mass       5.0 TeV	ſ ſeV	$\Gamma/m = 1\%$ $g_V = 3$ $g_V = 3$ $m(N_R) = 0.5 \text{ TeV}, g_L = g_R$	1903.06248 1709.07242 1805.09299 1804.10823 CERN-EP-2019-100 1801.06992 ATLAS-CONF-2019-003 1712.06518 1807.10473 1904.12679
C	Cl qqqq Cl ℓℓqq Cl tttt	_ 2 e,μ ≥1 e,μ	2 j 	– – Yes	37.0 36.1 36.1	Λ Λ Λ 2.57 TeV		<b>21.8 TeV</b> $\eta_{LL}^-$ <b>40.0 TeV</b> $\eta_{LL}^-$ $ C_{4t}  = 4\pi$	1703.09127 1707.02424 1811.02305
DM	Axial-vector mediator (Dirac DM) Colored scalar mediator (Dirac DM) $VV_{\chi\chi}$ EFT (Dirac DM) Scalar reson. $\phi \rightarrow t\chi$ (Dirac DM)	0 e,μ Λ) 0 e,μ 0 e,μ 0-1 e,μ	1 – 4 j 1 – 4 j 1 J, ≤ 1 j 1 b, 0-1 J	Yes Yes Yes Yes	36.1 36.1 3.2 36.1	m <sub>med</sub> 1.55 TeV           m <sub>med</sub> 1.67 TeV           M₄         700 GeV           m <sub>φ</sub> 3.4 TeV		$g_q$ =0.25, $g_{\chi}$ =1.0, $m(\chi)$ = 1 GeV $g$ =1.0, $m(\chi)$ = 1 GeV $m(\chi)$ < 150 GeV $\gamma$ = 0.4, $\lambda$ = 0.2, $m(\chi)$ = 10 GeV	1711.03301 1711.03301 1608.02372 1812.09743
ГО	Scalar LQ 1 <sup>st</sup> gen Scalar LQ 2 <sup>nd</sup> gen Scalar LQ 3 <sup>rd</sup> gen Scalar LQ 3 <sup>rd</sup> gen	1,2 <i>e</i> 1,2 μ 2 τ 0-1 <i>e</i> , μ	≥ 2 j ≥ 2 j 2 b 2 b	Yes Yes – Yes	36.1 36.1 36.1 36.1	LQ mass         1.4 TeV           LQ mass         1.56 TeV           LQ <sup>u</sup> mass         1.03 TeV           LQ <sup>d</sup> mass         970 GeV		$egin{aligned} eta &= 1\ eta &= 1\ \mathcal{B}(\mathrm{LQ}_3^u  o b au) &= 1\ \mathcal{B}(\mathrm{LQ}_3^d  o t au) &= 1\ \mathcal{B}(\mathrm{LQ}_3^d  o t au) &= 0 \end{aligned}$	1902.00377 1902.00377 1902.08103 1902.08103
Heavy quarks	$\begin{array}{l} VLQ \ TT \rightarrow Ht/Zt/Wb + X \\ VLQ \ BB \rightarrow Wt/Zb + X \\ VLQ \ T_{5/3} \ T_{5/3}   \ T_{5/3} \rightarrow Wt + X \\ VLQ \ Y \rightarrow Wb + X \\ VLQ \ B \rightarrow Hb + X \\ VLQ \ QQ \rightarrow WqWq \end{array}$	multi-channe multi-channe 2(SS)/≥3 e, 1 e, μ 0 e,μ, 2 γ 1 e, μ	el el $\mu \ge 1$ b, $\ge 1$ j $\ge 1$ b, $\ge 1$ $\ge 1$ b, $\ge 1$ $\ge 4$ j	i Yes j Yes j Yes Yes	36.1 36.1 36.1 36.1 79.8 20.3	T mass       1.37 TeV         B mass       1.34 TeV         T <sub>5/3</sub> mass       1.64 TeV         Y mass       1.85 TeV         B mass       1.21 TeV         Q mass       690 GeV		SU(2) doublet SU(2) doublet $\mathcal{B}(T_{5/3} \rightarrow Wt) = 1, c(T_{5/3}Wt) = 1$ $\mathcal{B}(Y \rightarrow Wb) = 1, c_R(Wb) = 1$ $\kappa_B = 0.5$	1808.02343 1808.02343 1807.11883 1812.07343 ATLAS-CONF-2018-024 1509.04261
Excited fermions	Excited quark $q^* \rightarrow qg$ Excited quark $q^* \rightarrow q\gamma$ Excited quark $b^* \rightarrow bg$ Excited lepton $\ell^*$ Excited lepton $\nu^*$	_ 1 γ - 3 e, μ 3 e, μ, τ	2 j 1 j 1 b, 1 j –	- - - -	139 36.7 36.1 20.3 20.3	q* mass       6.7         q* mass       5.3 TeV         p* mass       2.6 TeV         b* mass       2.6 TeV         ℓ* mass       3.0 TeV         v* mass       1.6 TeV	7 TeV V	only $u^*$ and $d^*$ , $\Lambda = m(q^*)$ only $u^*$ and $d^*$ , $\Lambda = m(q^*)$ $\Lambda = 3.0 \text{ TeV}$ $\Lambda = 1.6 \text{ TeV}$	ATLAS-CONF-2019-007 1709.10440 1805.09299 1411.2921 1411.2921
Other	Type III Seesaw LRSM Majorana $\nu$ Higgs triplet $H^{\pm\pm} \rightarrow \ell \ell$ 22 Higgs triplet $H^{\pm\pm} \rightarrow \ell \tau$ Multi-charged particles Magnetic monopoles $\sqrt{s} = 8 \text{ TeV}$	$1 e, \mu  2 \mu  2,3,4 e, \mu (St  3 e, \mu, \tau  -  -  = 13 TeV  tial data$	≥ 2 j 2 j S) - - - - √s = 1	Yes   - - 3 TeV	79.8 36.1 36.1 20.3 36.1 34.4	$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	<u></u> ] 1/	$m(W_R) = 4.1$ TeV, $g_L = g_R$ DY production DY production, $\mathcal{B}(H_L^{\pm\pm} \rightarrow \ell\tau) = 1$ DY production, $ q  = 5e$ DY production, $ g  = 1g_D$ , spin 1/2	ATLAS-CONF-2018-020 1809.11105 1710.09748 1411.2921 1812.03673 1905.10130
	par			ald			I V	Mass scale [TeV]	

\*Only a selection of the available mass limits on new states or phenomena is shown.

†Small-radius (large-radius) jets are denoted by the letter j (J).

- Over the past decade, both CMS and ATLAS have produced a large number of interesting results
- This rich table comes from just one physics group (EXOTICS) in one collaboration (ATLAS)!
- What have I done?







#### **Research** Timeline









# Exotic Search with B-tagging...



### **Resonance Search with b-tagged Jets**

- flagship analysis in hadron colliders for decades
  - The most inclusive search
- A natural extension is to apply b-tagging on the jets
  - couplings to bottom quarks



Search for heavy resonances in the final states with two jets has been the

# More sensitive to new physics where the heavy resonance has larger



### Analysis Strategy

- In five steps:
  - Apply a jet trigger with the lowest threshold
  - Select two good jets
  - Apply signal enhancement selections:  $|y^*| < 0.8$  and b-tagging
  - Fit the data
  - Interpretation



#### Sounds really simple!

- A very complex spectrum to fit
- B-tagging applied on a mixture of jets with different flavors (Cartoons only show the light and bottom)





# **Background Modeling Method**

validate it using control samples



*ith* bin

To fit the background, we apply a Sliding Window Fit (SWiFt) method and

- The canonical global fit can not describe the background given the complexities
- SWiFt is a numerical approximation where each bin is estimated by performing a global fit within a subset (window) of the full spectrum
- It works but it is a rather complex model with many free parameters



# **Background Modeling Method**

- We have to validate the background modeling method
  - Need background-only control samples
  - Major background, multi-jet, does not have reliable simulations • Flavor fractions are not well modeled
- - Last differential bb measurement was done in Run1





# **Background Modeling Method**

- We have to validate the background modeling method
  - Need background only control samples
  - Major background, multi-jet, does not have reliable simulations
    - b-tagging/mis-tag rate are not calibrated at high  $p_T$





# A New "ABCD" Approach

- and flavor compositions in multi-jet events
  - Extract this information from data

Two sidebands data scaling met						
A tagged ly*l > 0.8	B tagged ly*l < 0.8	<ul> <li>Obtain per j efficiencies regions (A a in p<sub>T</sub> and η</li> <li>Calculate ev</li> </ul>				
C untagged ly*l > 0.8	D untagged ly*l < 0.8	<ul> <li>tagging effic</li> <li>Scale untagg to tagged reg</li> </ul>				

# The major challenge comes from unknown tagging efficiency/mis-tag rate,

#### hod

et tagging n |y\*| inverted nd C), binned

ent level iencies ged (D) region gion (B)

- Inverting |y\*| suppresses the signal • contaminations, but also alters  $m_{ii}$
- Instead we calculate per-jet tagging efficiency
  - Since the flavor fraction of leading jet is correlated with the sub-leading jet's flavor. Both absolute efficiency and conditional efficiency are calculated:
    - $P(j_1)$ ,  $P(j_2)$  and  $P(j_2|j_1)$
- Event level efficiency is calculated by
  - >= 1 b-tag:

 $P(j_1) + P(j_2) - P(j_1)P(j_2|j_1)$ 

• ==2 b-tag:

 $P(j_1)P(j_2|j_1)$ 



#### Results

#### • Both 2015 + 2016 di-b-jet search and full Run 2 di-(b)-jet search are public!





### **Resonance Search with More b-tagged Jets**

fermions?



- quarks
  - Multi-b-jet final state
  - Only considering leading four jets

What if the new heavy resonance is exclusively coupled to third generation



The new heavy resonance has to be produced in association with two b-



# **New Ideas Deployed I: Efficiency Scaling Method**

- are considering NLO multi-jet processes
- I developed a new approach



- smear the influence from potential signal
  - The impact from signal is found to be small in stress tests
- Produce bkg-only pseudo-data samples to validate the background modeling strategy

• The "ABCD" approach does not work any more in this final state as now we

• By fitting the signal region selection efficiency (simple polynomial fit), we



### **New Ideas Deployed II: Functional Decomposition**

• We have been using empirical functions to fit the di-jet invariant mass spectra in history

#### New Method:

- Functional Decomposition
  - Using a truncated series to describe the spectrum
  - Analogous to Fourier Analysis



arXiv:1805.04536

### The Future Of "Simple" Search

- LHC has accumulated ~140  $fb^{-1}$  data • It will take LHC a while to double the integrated luminosity
- - Differential measurements are in their way
- - corresponding calibrations

 Search strategies need to be thoroughly re-evaluated and improved Search with tagging techniques can benefit significantly from CP development

• Di-b-jet search is limited by the b-tagging performance at high  $p_T$  and the





# So, Flavor Tagging...



### **B-hadron Properties**

- b-hadrons have:
  - Majority of the energy from hadronization (~80%)
  - Relatively large mass (~5 GeV)
  - Significant lifetime (~1.5 ps)
  - High decay multiplicity (~5 charged particles)
  - Relatively large  $b \rightarrow \mu + X$ branching ratio (~20%)
- b-tagging algorithms are constructed based on the above properties

- **Primary Vertex**
- **b-hadron Decay Vertex**
- Tracks from b-hadron
- c-hadron Decay Vertex
- Tracks from c-hadron
- Tracks from PV
- Lepton from b-hadron



# Low Level Taggers

- Experimental signatures:
  - Secondary Vertex
  - Heavier Vertex Mass
  - Tertiary Vertex
  - Displaced Tracks
  - Larger Track Multiplicity
  - Muon
- Low level taggers:
  - Track based:
    - IP2D, IP3D and RNNIP
  - Secondary vertex based: SV1
  - Decay topology based:
    - JetFitter
  - Muon based: SMT

- Secondary Decay Vertex **Tertiary Decay Vertex Displaced Tracks**
- → Muon



### Low Level Taggers



#### **ATL-PHYS-PUB-2017-013** ATL-PHYS-PUB-2017-003 **ATL-PHYS-PUB-2018-025**

#### IP2D and IP3D

 Consider the IP parameters of individual track

#### RNNIP

- Explore the correlations between tracks
- SV1
  - Explore properties of the secondary vertex such as vertex mass

#### JetFitter

• Explore decay properties such as energy ratio





# High Level Taggers

#### Construction **Inclusive Secondary** Trak IP:

IP2D/IP3D

# Training

Trained on multiple jet collections:

AK4EMTopo Jets (Phasing out)

Vertex:SV1

AK4EMPFlow Jets

- Variable Radius Track Jets





Using Hybrid Sample  $t\bar{t}$  and Z'

- Both MV2 and DL1 have various versions
  - MV2r and DL1r (Recommended)
    - Including RNNIP
  - MV2 and DL1 (Backup)
  - MV2mu and DL1mu (R&D)
    - Including SMT
  - MV2rmu and DL1rmu (R&D)
    - Including RNNIP and SMT



#### Performance

#### Recent training campaign gives us the best performance so far



#### FTAG-2019-005

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# Working Points (WPs)



#### FTAG-2019-005





# Working Points (WPs)



#### FTAG-2019-005





# Working Points (WPs)



#### FTAG-2019-005





# Flavor Tagging Calibration

- Mis-modeling of the input variables used by the algorithms
- The performance of b-tagging in MC is different from that in data Need to correct the performance in MC
  - Ideally one can correct all input variables
  - Practically we correct the resulting tagging performance
- Basic idea
  - $p_T$ , and correct the performance in MC to match the data
- Measure the b-tagging efficiency/mis-tag rate in data and MC binned in jet • The mainstream calibrations are done in  $t\bar{t}$  and Z + jets events The expanding physics programs now demand more versatile calibrations To cover different kinematic regions (Di-b-jet resonance) • To avoid circular dependence (Top decay branching ratio measurement)





# Mainstream Calibrations



#### **B-tagging Efficiency Calibration** Eur. Phys. J. C 79 (2019) 970

distribution and perform a combined log-likelihood fit



# • Method: Select di-lepton $t\bar{t}$ events, construct CR/SR based on the $m_{i,l}$



# **B-tagging Efficiency Calibration**

Latest results for VRTrack Jets and EMPFlow Jets



#### FTAG-2019-003 FTAG-2019-004



# **C-mistag Rate Calibration**

 Method: Select semi-leptonic tī events, associated jets with W boson via a kinematic fit and perform a likelihood fit to extract the mis-tag rate



_					
j jet p <sub>T</sub> [GeV	[65,140]	26.6 ± 0.1	24.2 ± 0.1	27.3 ± 0.1	
leading	[40,65]	21.8 ± 0.1	20.8 ± 0.1		
	[25,40]	20.0 ± 0.1	ATLAS Simulation Preliminary		
		[25,40]	[40,65]	[65,140]	
			sublead	ling jet p <sub>T</sub> [GeV]	

#### ATLAS-CONF-2018-001





### **C-mistag Rate Calibration**

Latest results for VRTrack Jets and EMPFlow Jets



FTAG-2019-003 FTAG-2019-004



### L-mistag Rate Calibration

likelihood fit to extract light jet mistag rate



- Light-flavor jets have a ~symmetric signed  $d_0$  distribution
- Charm- and bottom-flavor jets have much longer positive tails

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#### ATLAS-CONF-2018-006

• Method: Enrich the light jet fraction via "flipped" taggers, select Z + jets events, fit the secondary vertex mass to obtain flavor fractions and perform a

#### Flipped Tagger

- Negate the sign of track IP parameters before b-tagging
- bottom or charm are shifted towards lower values
- regions







### L-mistag Rate Calibration

Latest results for VRTrack Jets and EMPFlow Jets



FTAG-2019-003 FTAG-2019-004

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# **Calibrations In Multi-jet**



# **Calibrations Using Multi-jet Events**

- Multi-jet events provide an abundance source of b-jets
  - A large number of b-jets populated in a broader kinematic region
- More challenging
  - There is also more background
  - Simulation is not sufficient
- The calibrations apply template fits
  - Two discriminant variables can be used
    - $S_{d_0}$  and  $p_T^{rel}$



$$S_{d_0} = \left| \frac{d_0}{\sigma_{d_0}} \right| \cdot s_j$$
$$s_j = \text{sign} \left[ \sin \left( \arctan(\frac{p_y(j)}{p_x(j)} - \phi(t)) \right) \right]$$

longer lifetimes



# $p_T^{rel}$ Calibration: Discriminant Variable

• Muons from direct bottom decays have harder  $p_T$  in the rest frame of b-hadron (p\*) than that from charm or cascade bottom decays

#### direct b decays c decays

Consider a massless twoday decay:

 $(P, \overrightarrow{P}) + (P, \overrightarrow{P}) = M_{b/c}$  $P = -M_{b/c}$ 



# $p_T^{rel}$ Calibration: Method

Calibration strategy:

- Define a discriminating variable  $p_T^{rel}$
- Produce  $p_T^{rel}$  templates for jets with different flavors, (b), charm (c) and light (I)
- Apply the b-enhanced selections in data
- Perform a template fit to tagged data and un-tagged data to obtain the fraction of b's

$$\varepsilon_{b}^{data} = \frac{f_{b}^{tagged} N_{data}^{tagged}}{f_{b}^{tagged} N_{data}^{tagged} + f_{b}^{untagged} N_{data}^{untag}}$$

- Extract the b-tagging efficiency using number of events and the fractions
- Compare the efficiencies measured in data with the those in MC to derive corrections

ged





 $p_T^{rel}$  Calibration: Results



Full Run2 calibration is going with much improved strategies

Better precision is expected

#### 2016 JINST 11 P04008

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#### High *p*<sub>T</sub> Calibration: Discriminant Variable • Similar strategy as the $p_T^{rel}$ calibration • No public results are available yet. Show example distributions from g ightarrow bbmeasurement









# Even Longer Lifetimes...



# **B-tagging and LLP Tagging**

- Long-lived particles can have lifetimes similar as b-hadrons Can be b-tagged



#### Standard searches with b-tagging have sensitivities to such LLPs





### Future of b-tagging





• Explore new strategies to improve performance at high  $p_T$ • Optimize the existing low taggers

High Level Taggers

#### Proving ground for new Machine Learning techniques

Training

 More efficient and unbiased training samples?

- Training for charm/ strange tagging
- Training for LLP tagging?





# A Flashback...Displaced Lepton In CMS



# **Displaced Lepton Search**

parameters



#### • R-parity Violating Supersymmetry can yield leptons with large impact

- A search was done by simply requiring a displaced  $e - \mu$  lepton pair
- Large impact parameter  $d_0$  Main backgrounds come from heavy flavor multi-jet and events  $Z \rightarrow \tau \bar{\tau}$
- Estimate multi-jet via a datadriven method and the rest in simulation



# **Displaced Lepton Search**

parameters



#### PhysRevLett.114.061801

#### R-parity Violating Supersymmetry can yield leptons with large impact

- Applying standard algorithms with the analysis strategies optimized for long-lived signatures, top squark mass up to 790 GeV with  $c\tau$  of 2 cm is excluded.
- However it was found sensitivities to longer lifetimes were constrained by standard algorithms



### **Displaced Muon Reconstruction**

#### During LS1 I developed a set of new algorithms for displaced muons



#### **CMS-DP-2015-015**

# **Displaced Lepton Search**

search in CMS using 13 TeV data



#### **CMS-PAS-EXO-16-022**

Applying some of the improvements, I performed the first displaced lepton

 Top squark mass up to 870 GeV with  $c\tau$  of 2 cm is excluded with 2015 data

#### • My thesis!

 Full Run2 results should be on its way





#### **Displaced Lepton Reinterpretation CMS-PAS-EXO-16-007**

A similar re-interpretation work was done



- A search for  $LQ \rightarrow \mu\mu jj$  suing standard techniques can probe shorter lifetimes
- Again it demonstrates the power standard algorithms
- It also shows us where the gap is



### Summary and Outlook

- We have seen that traditional methods such as b-tagging and lepton reconstruction are already sensitive to certain long-lived signatures
- The corner of the parameter space for traditional algorithms is already very long-lived (or in general "Exotics") like
- CP work is crucial and a very comprehensive program can be built during LS2, making us better suited for Run3!

High  $p_T$  B-tagging Calibration : Tracking  $\longrightarrow$  High  $p_T B$ -tagging

LLP Tagging









and a second second

