

# AI-Driven Material Discovery for Next-Generation Directional Dark Matter Detectors

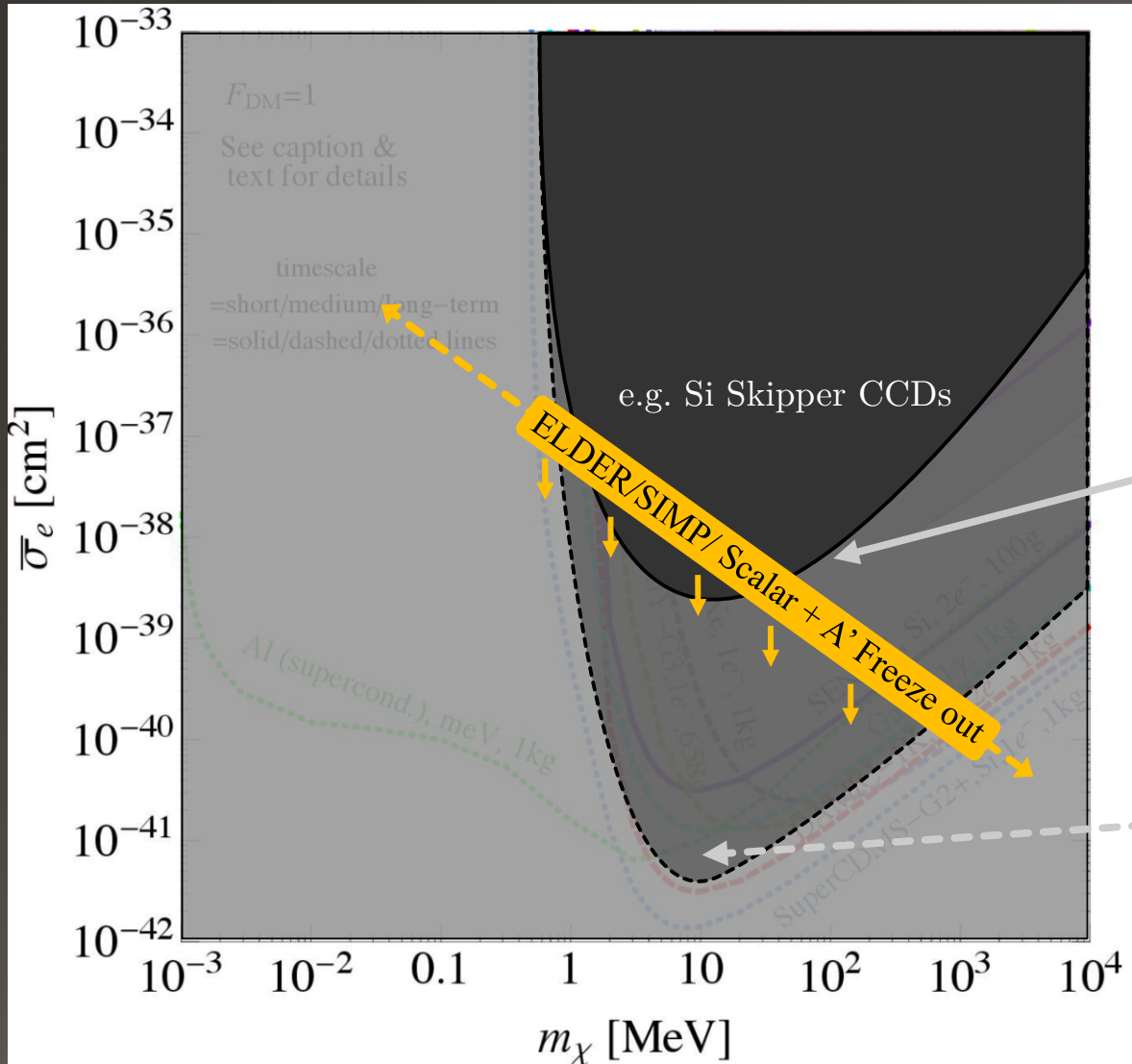
CARLOS BLANCO



Stockholm  
University



# Outlook and Potential Reach



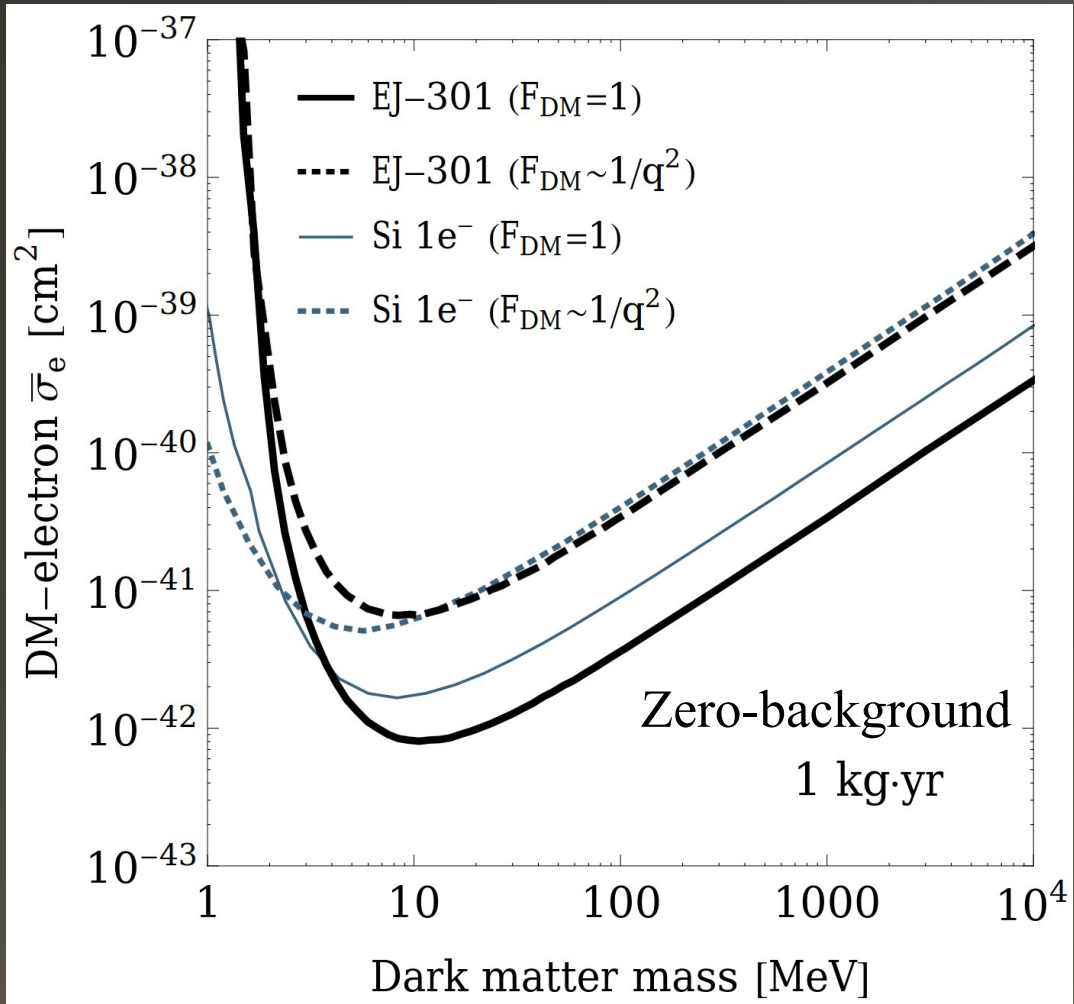
Present day:

Experimental results

Zero-background projection

# Outlook and Potential Reach

[CB, Collar, Kahn, Lillard: 1912.02822]



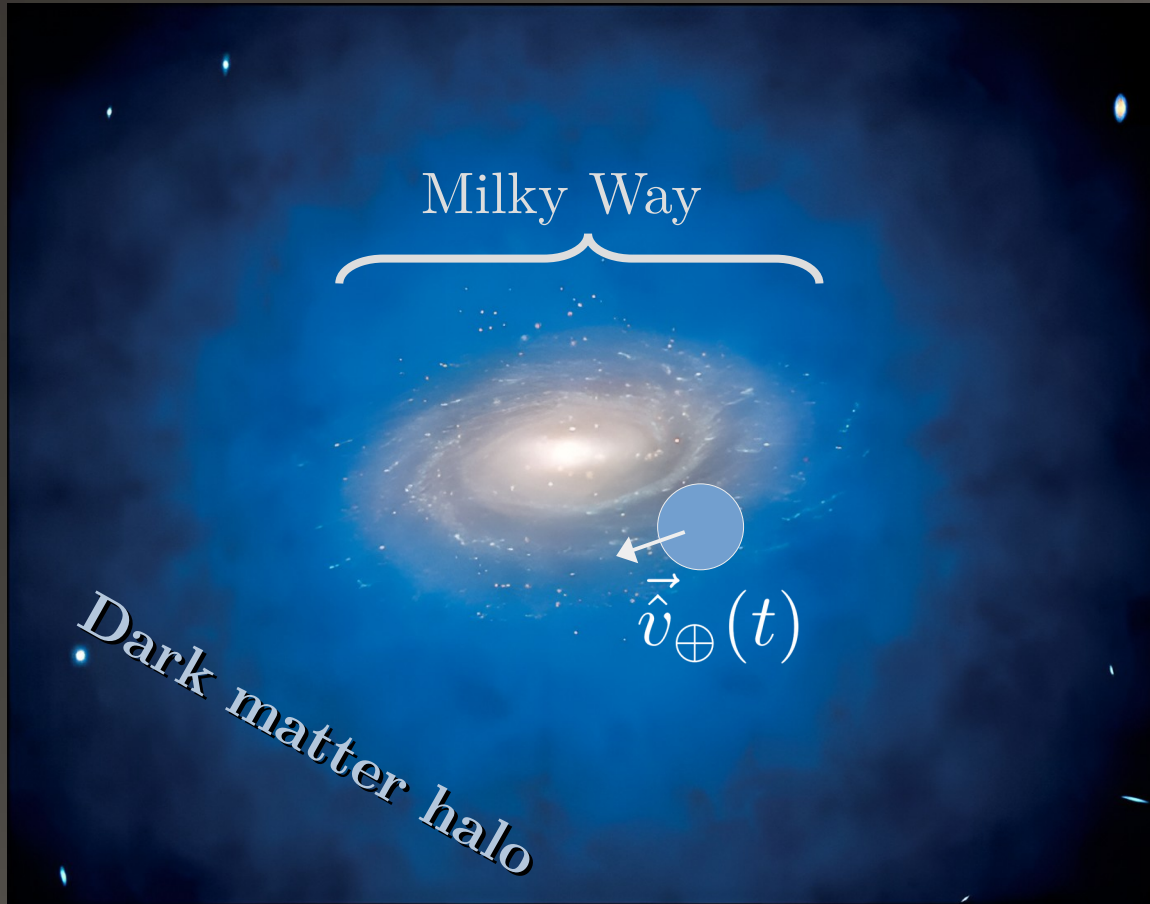
The obstacle

# Backgrounds

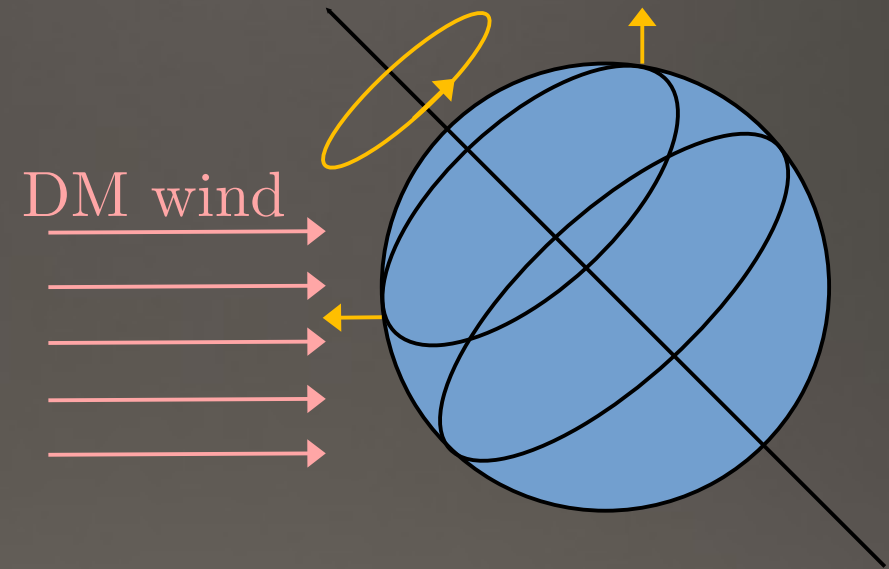
4 orders of magnitude of potential reach awaiting

Pound (kg) for Pound (kg) molecules produce about as much signal as e.g. Si.

# Directional Detection



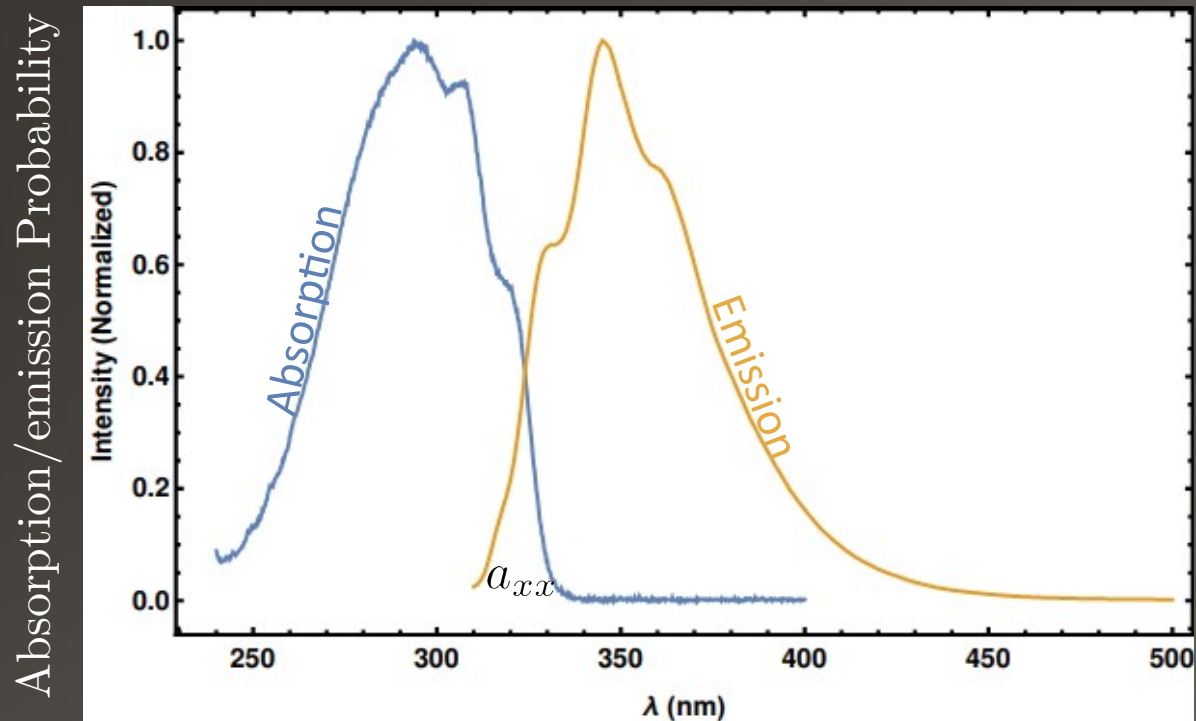
Effective dark matter “wind” from relative motion



Change in relative orientation between detector and dark matter wind leads to *daily modulation*

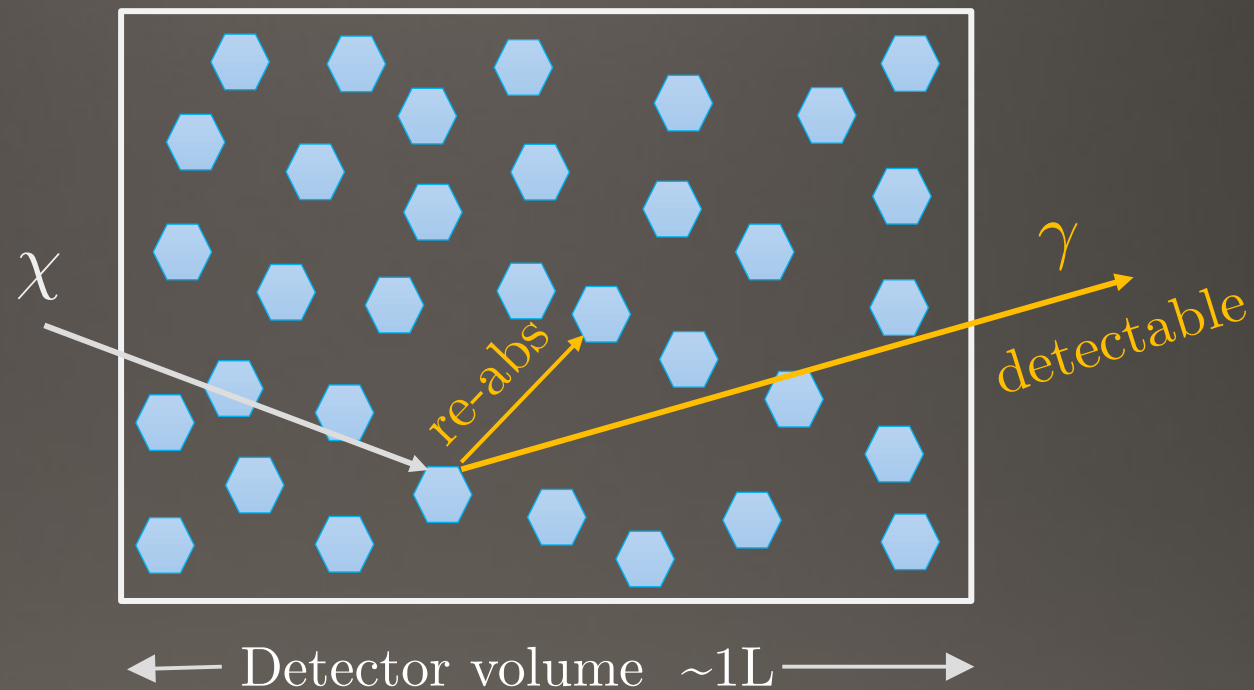
# Fluorescence with DM

fluorescence spectra



Decreasing energy (E) →

⬡: Chromophore:

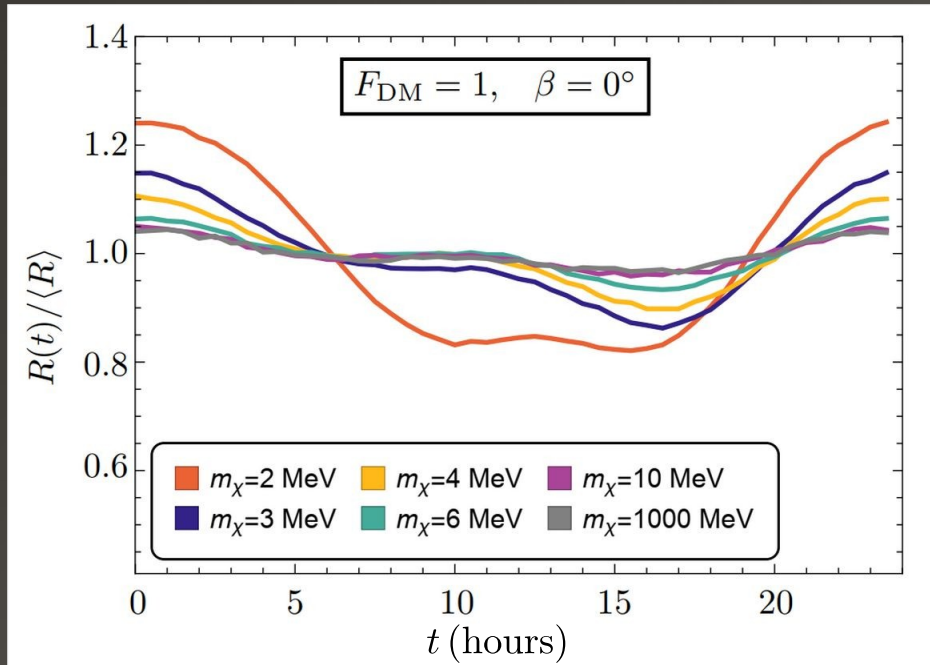


Probability for the photon to free stream

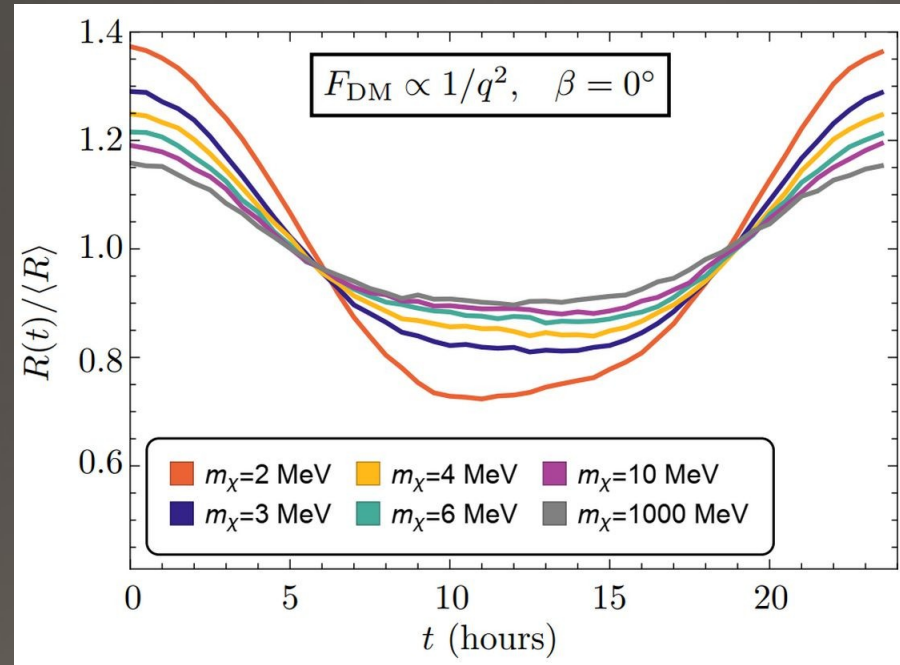
$$\Phi_{FB} \sim (1 - a_{xx}) \quad \text{e.g. molecular crystals: } \Phi_{FB} \approx 65\%$$

# Daily Modulation

(Contact interaction)



(Long-range interaction)



Predicted rate changes by up to 70% throughout the day.

That's a verifiable signal!

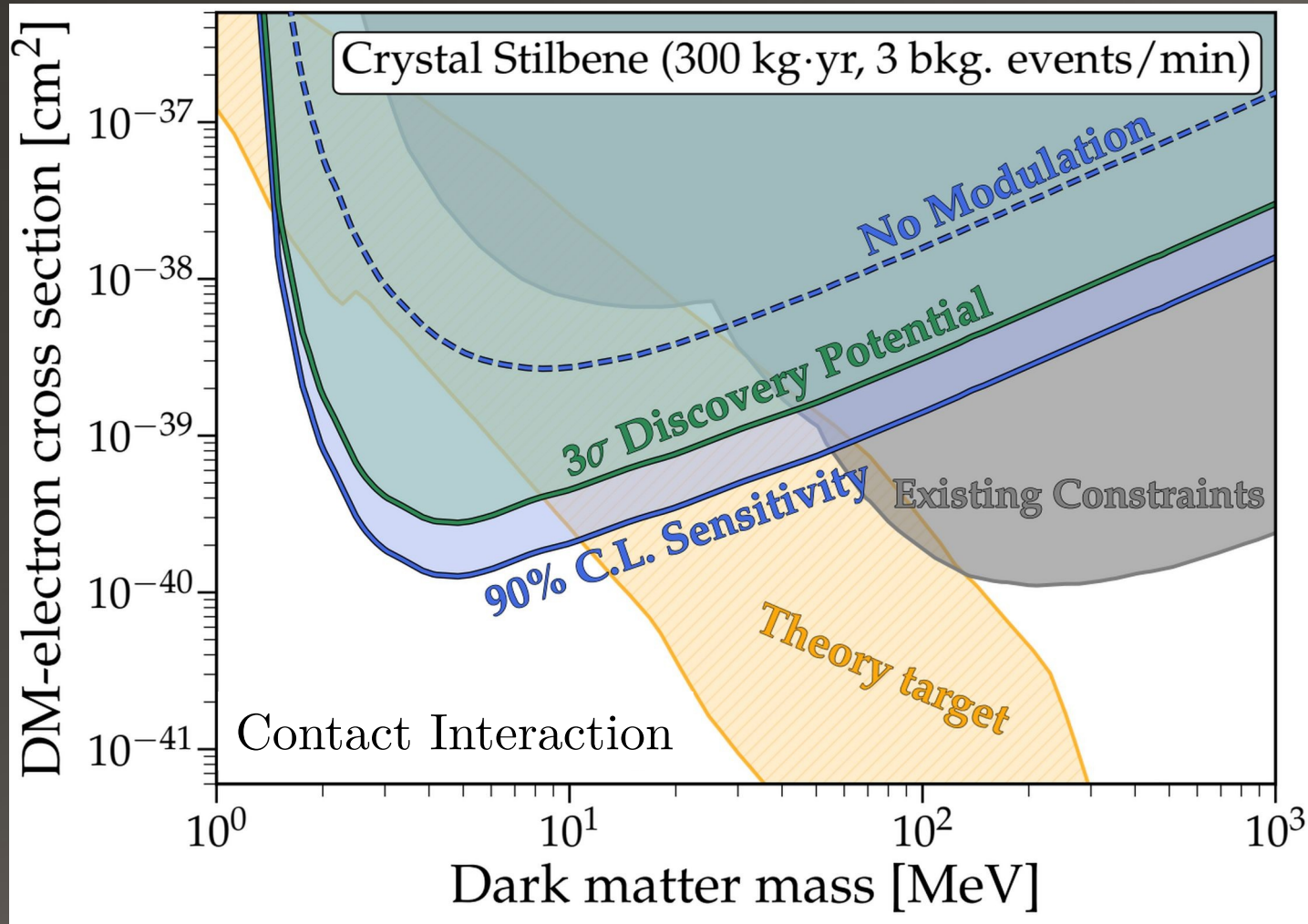
[CB, Kahn, Lillard, McDermott: 2103.08601]

*Modulation amplitude* remains as high as 10% even at the highest masses. This is due to the *fundamental anisotropy of the molecular form factor*.

$$f_{\text{RMS}} = [5\% - 25\%]$$

$$N_\sigma \sim f_{\text{RMS}} \sqrt{N_{\text{events}}}$$

# Sensitivity & Reach



Assuming realistic backgrounds

Exclusion w/o modulation

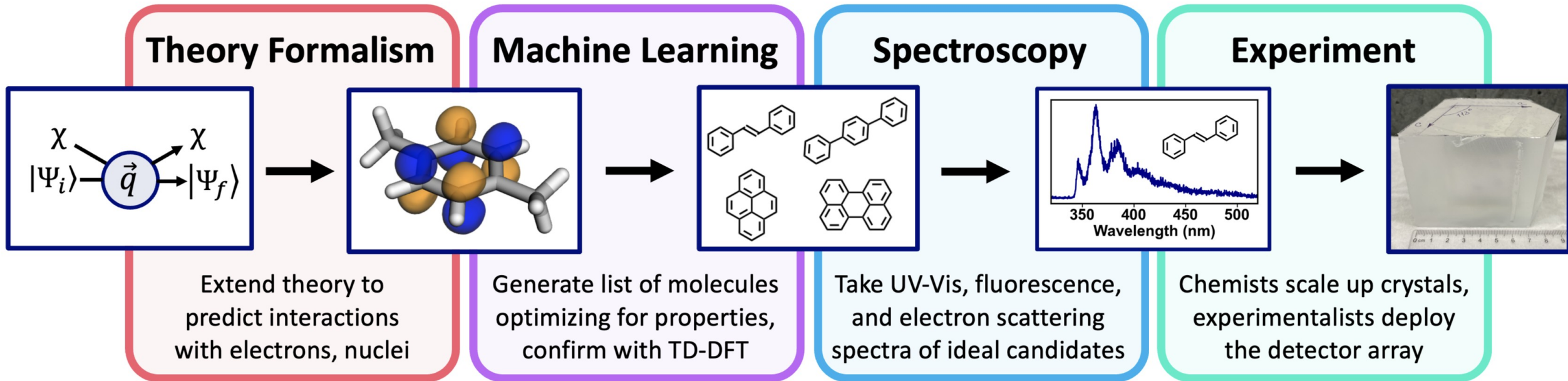
Discovery potential w/ mod.

Exclusion w/ modulation

100 kg t-stilbene crystals exposed for 3 years w/ background rate of 3 events/min dominated by radioactivity of <sup>14</sup>C, Th, & U.

Consistent w/ Borexino <sup>14</sup>C/<sup>12</sup>C  $\sim 10^{-18}$  (0.01/min/kg)

# Detector Development Pipeline



# Inventing Optimal Targets

Problem: Chemical space is unreasonably large

How many molecules possible with  
C, O, N, F, H?

< 9 atoms: 100s of Thousands (DFT Computable)

< 30 atoms: 100s of Billions (Intractable)

...toluene has 15, xylene has 18, t-stilbene has 26

## Method

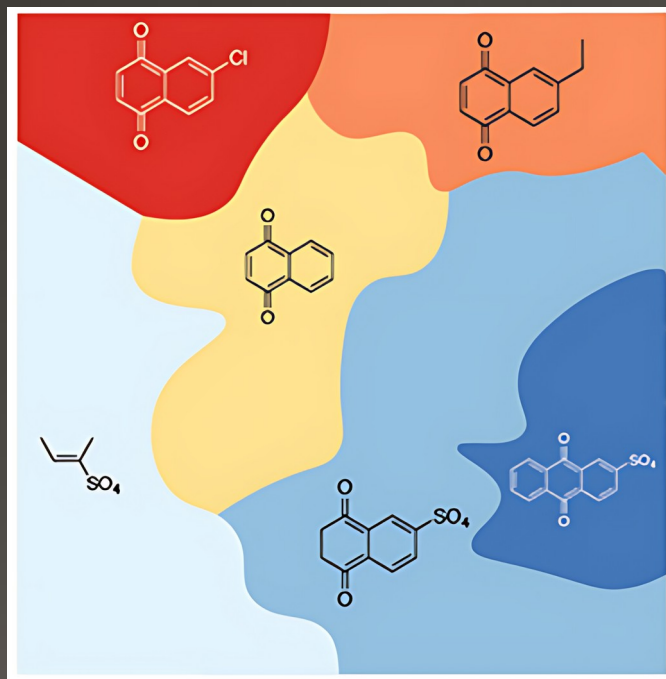
1. Look for known favorable properties - *cheminformatics*
2. Extra(intra)polate onto new molecules – *machine learning*

# ML for DM Direct Detection

Property prediction    Molecular Generation

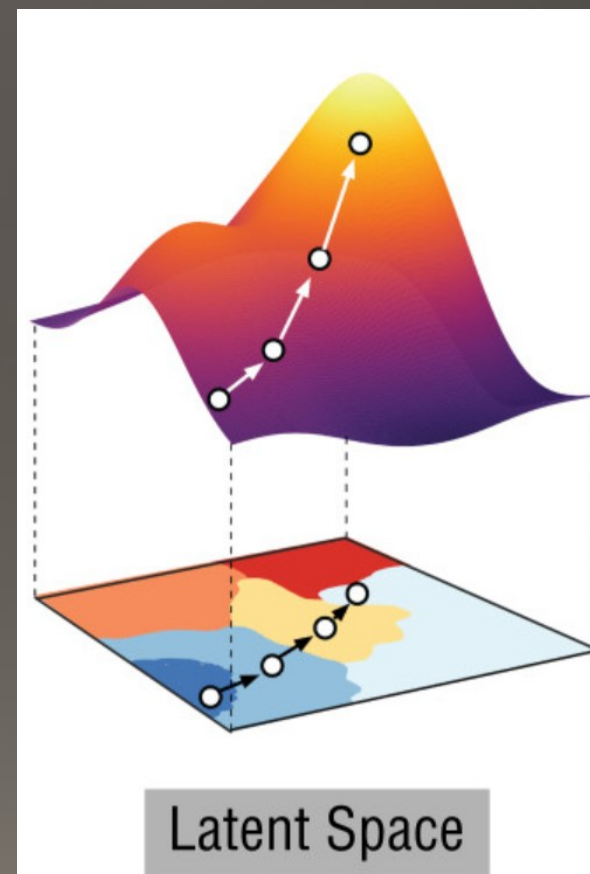
Energies & Matrix elements    Sample latent space  $\rightarrow$  new molecules

Map space of molecules into a smooth “latent space”



[Gomez-Bombarelli, ACS central science 4.2 (2018)]

Property prediction  
Energies & Matrix elements

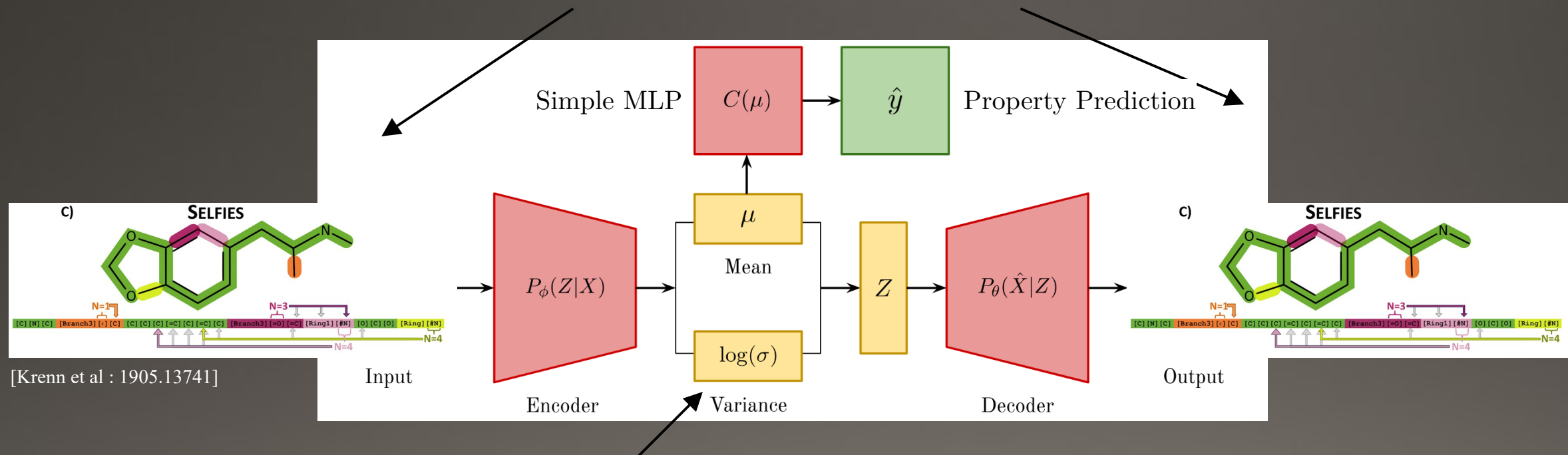


# ML for DM Direct Detection

Molecular Space: Variational Autoencoders

Energies & Matrix elements Sample latent space  $\rightarrow$  new molecules

Discrete and sparse data space



Continuous latent space

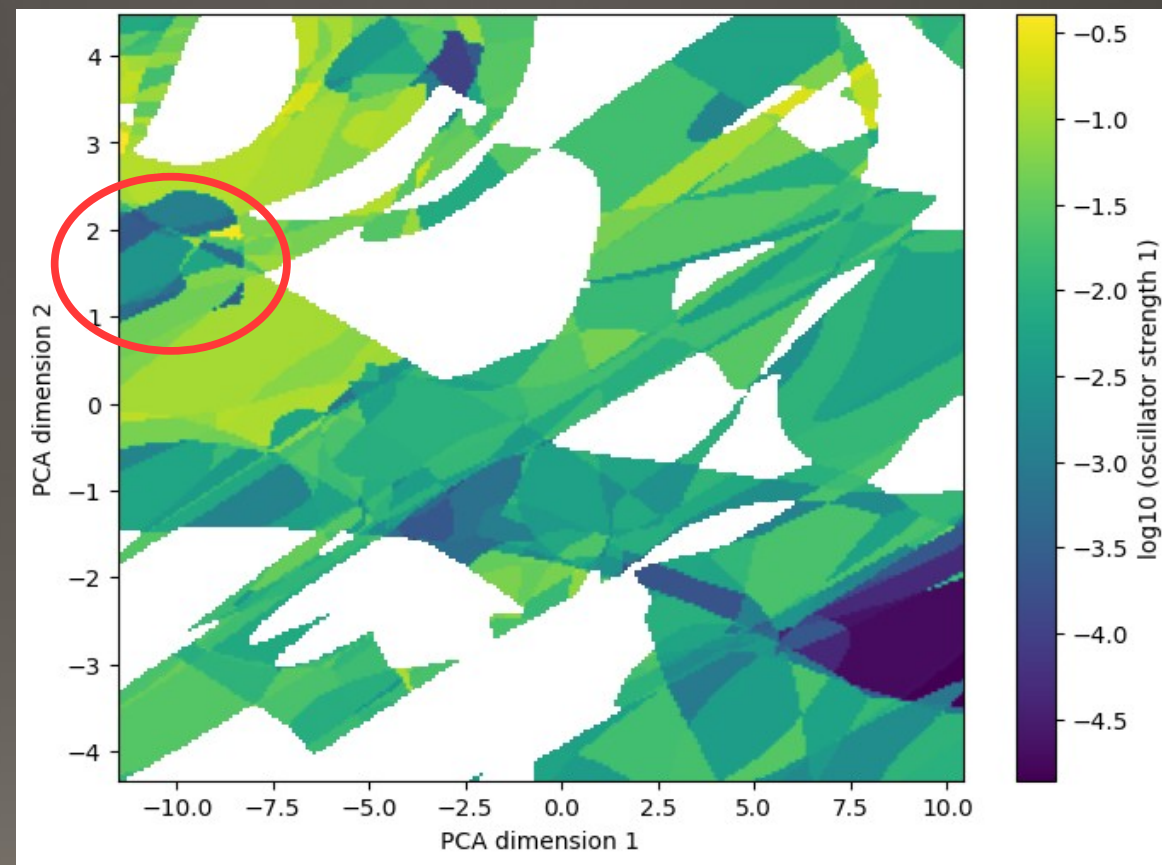
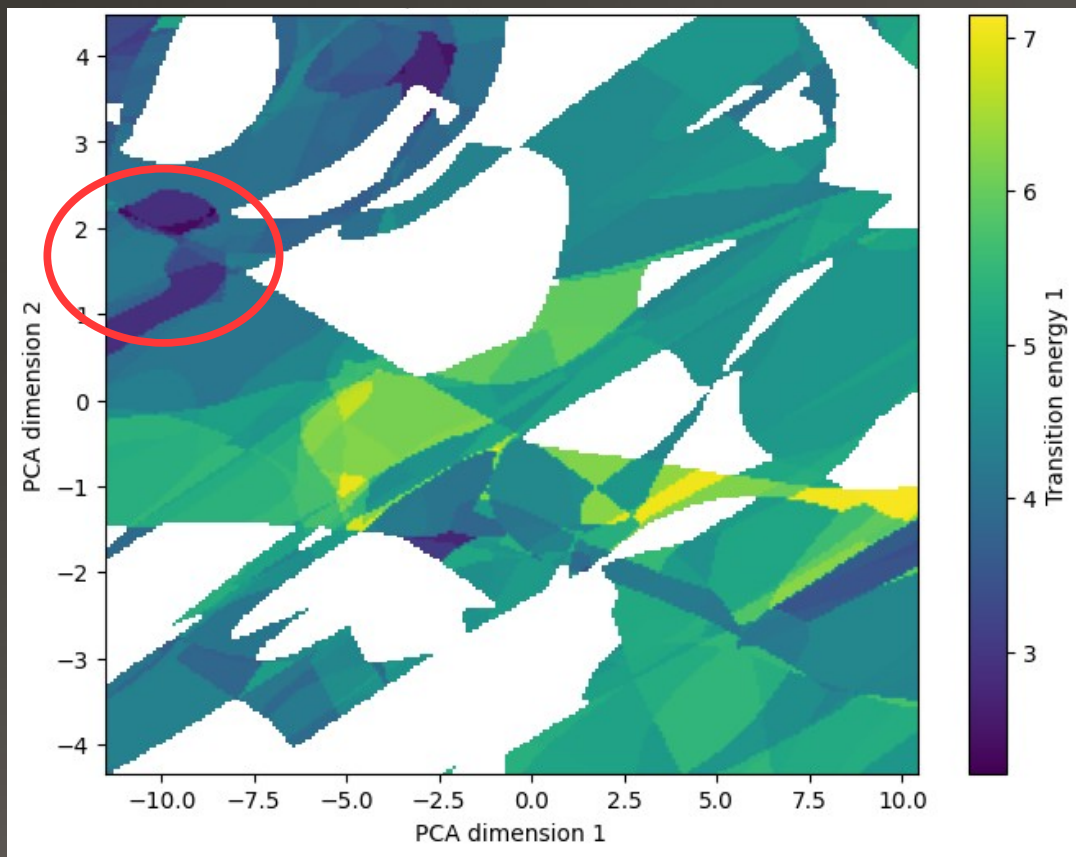
# ML for DM Direct Detection

Property prediction: Molecular latent space

Energies & Matrix elements Sample latent space  $\rightarrow$  new molecules

$$\Delta E$$

$$\langle r_{01} \rangle \approx |f_{0,1}|/q$$



# Clustering: Results

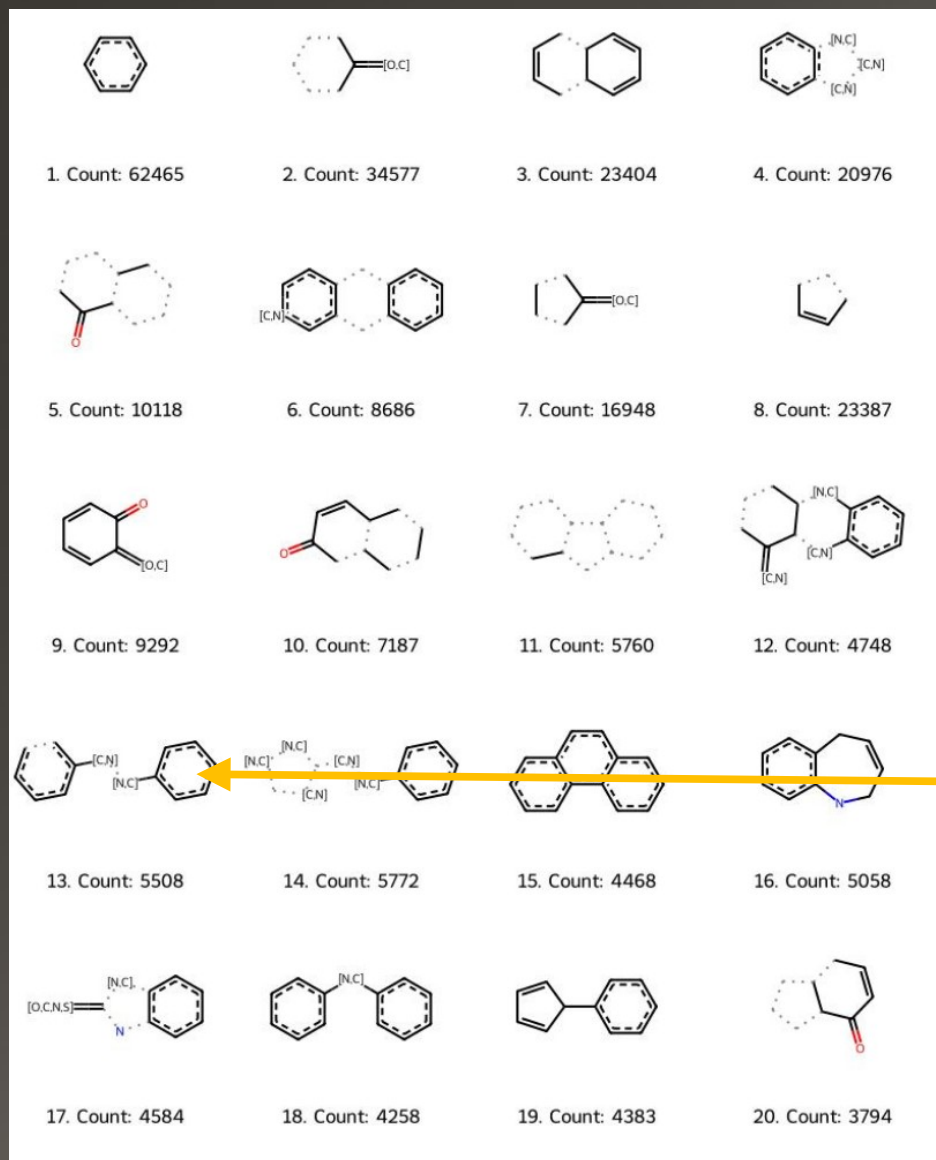
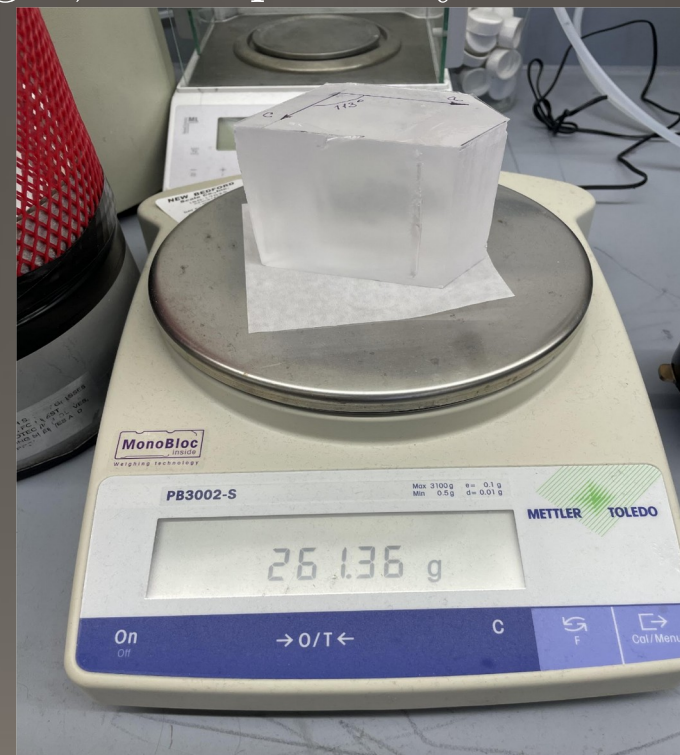


FIG. 7. Top 20 molecular motifs ranked by score. Note that certain atoms can be substituted by [C,N,O,S] as long as the structure remains isoelectronic. Dotted lines indicate that the bond could be a double or single bond.

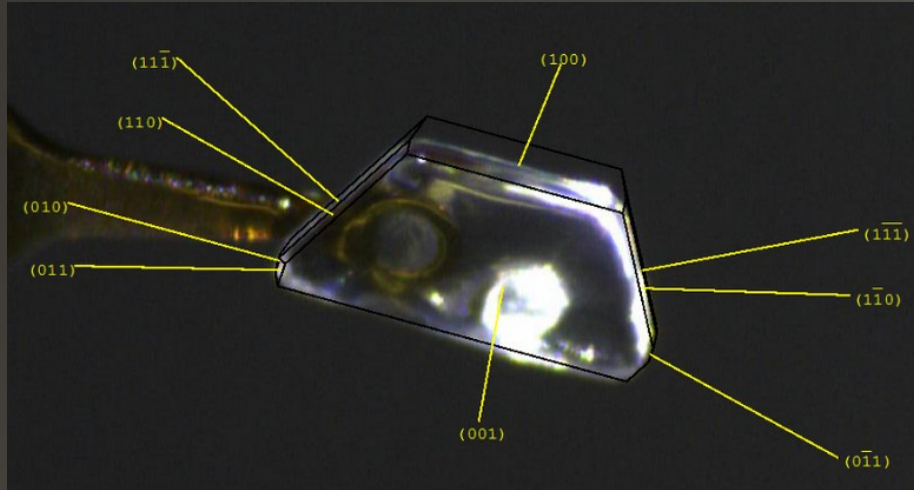
[C. Cook\*, CB, J. Smirnov 2501.00091]

In hindsight, we're probably on the right track.



# Characterizing t-stilbene

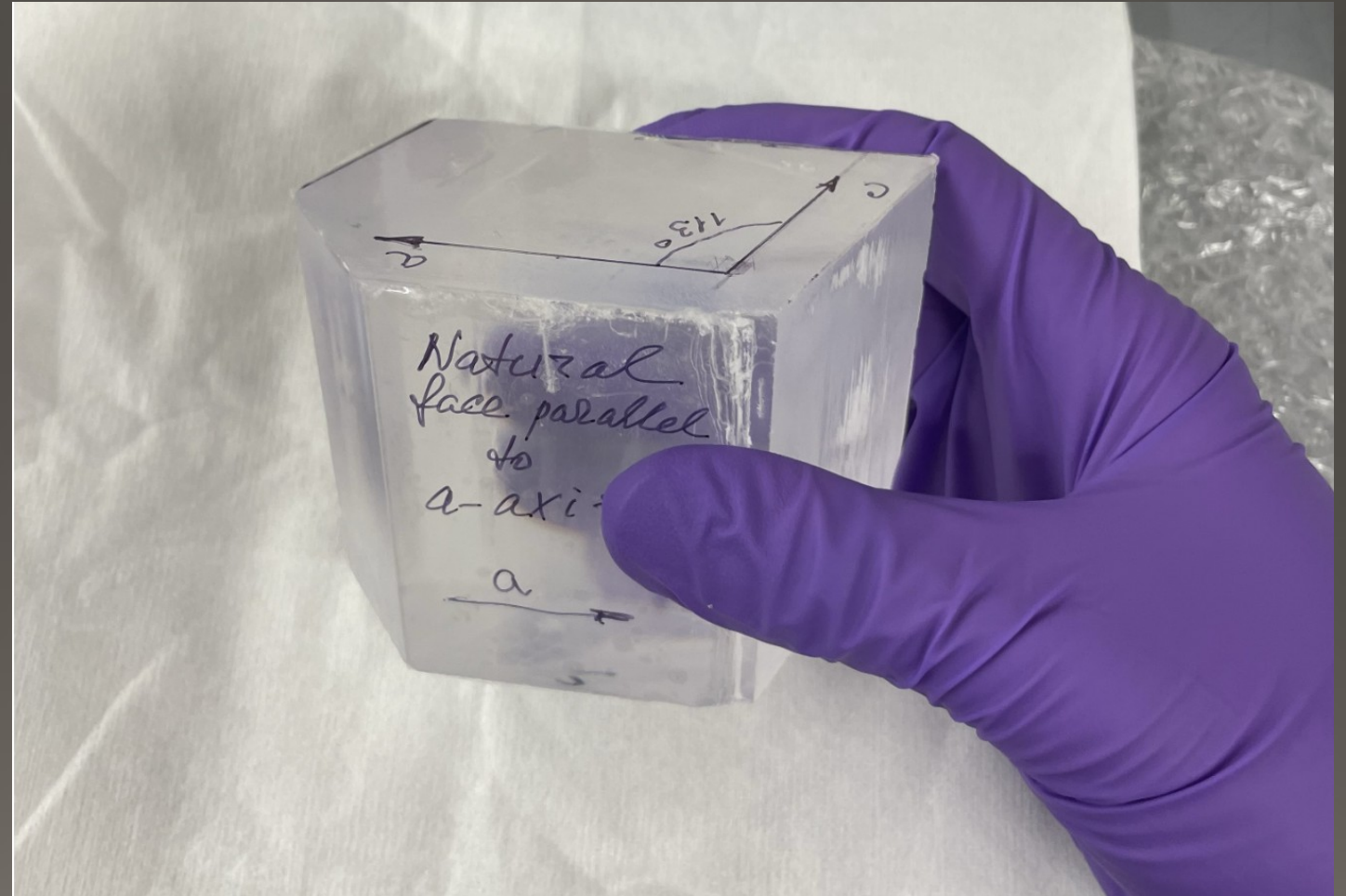
Calibrate at mg-scale



Credit: Dane Johnson  
Freedman Group (MIT)

Natalia Zaitseva (LLNL)

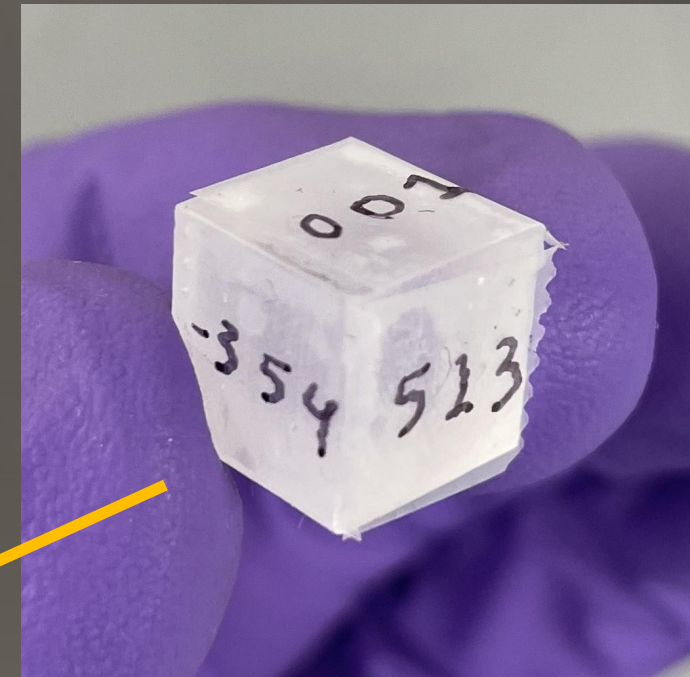
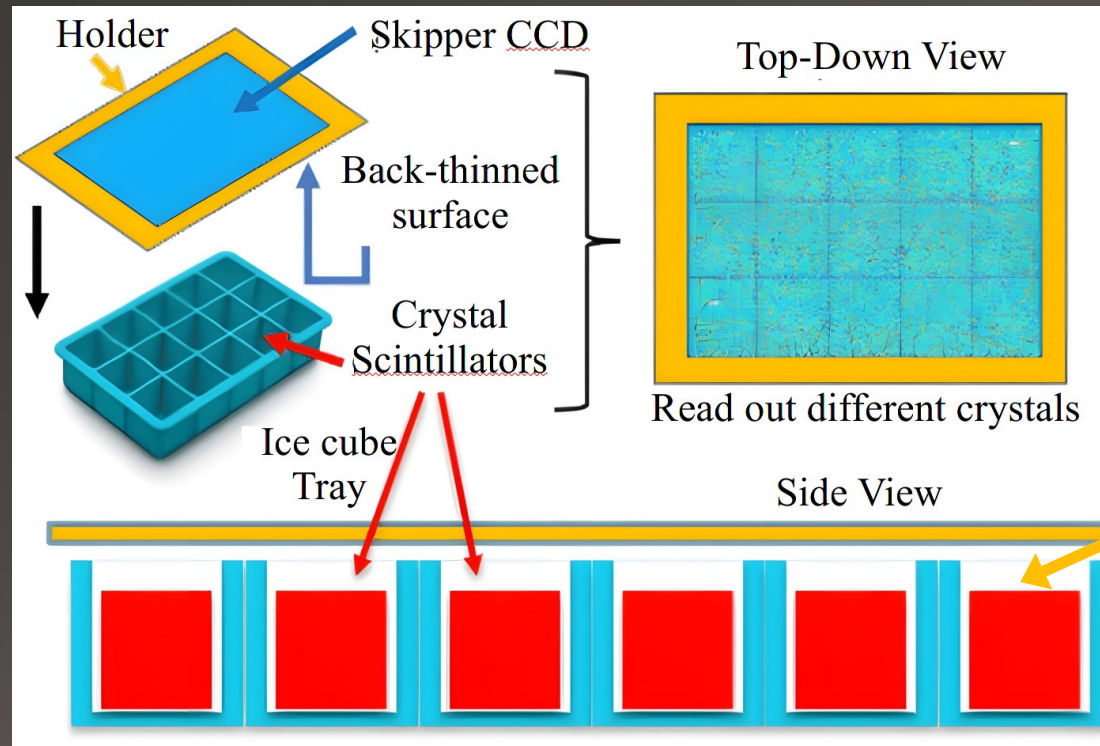
Deploy up to kg-scale



# Experimental Deployment

## DIANA\*

Daily modulation in an Intrinsically Anisotropic Array



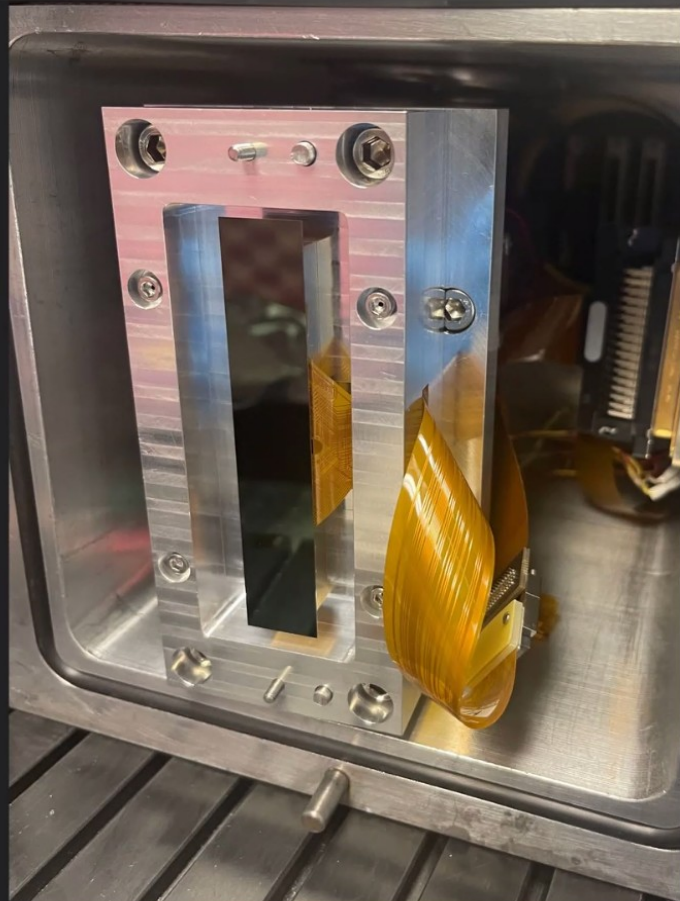
Many crystal samples can be read out by a single skipper CCD.

Collaboration: FermiLab, U. Toronto, MIT, UIUC, U. Oregon, Penn State

# Astro-Skipper CCDs as Readout

Taking pictures of crystals with a *very* sensitive CCD

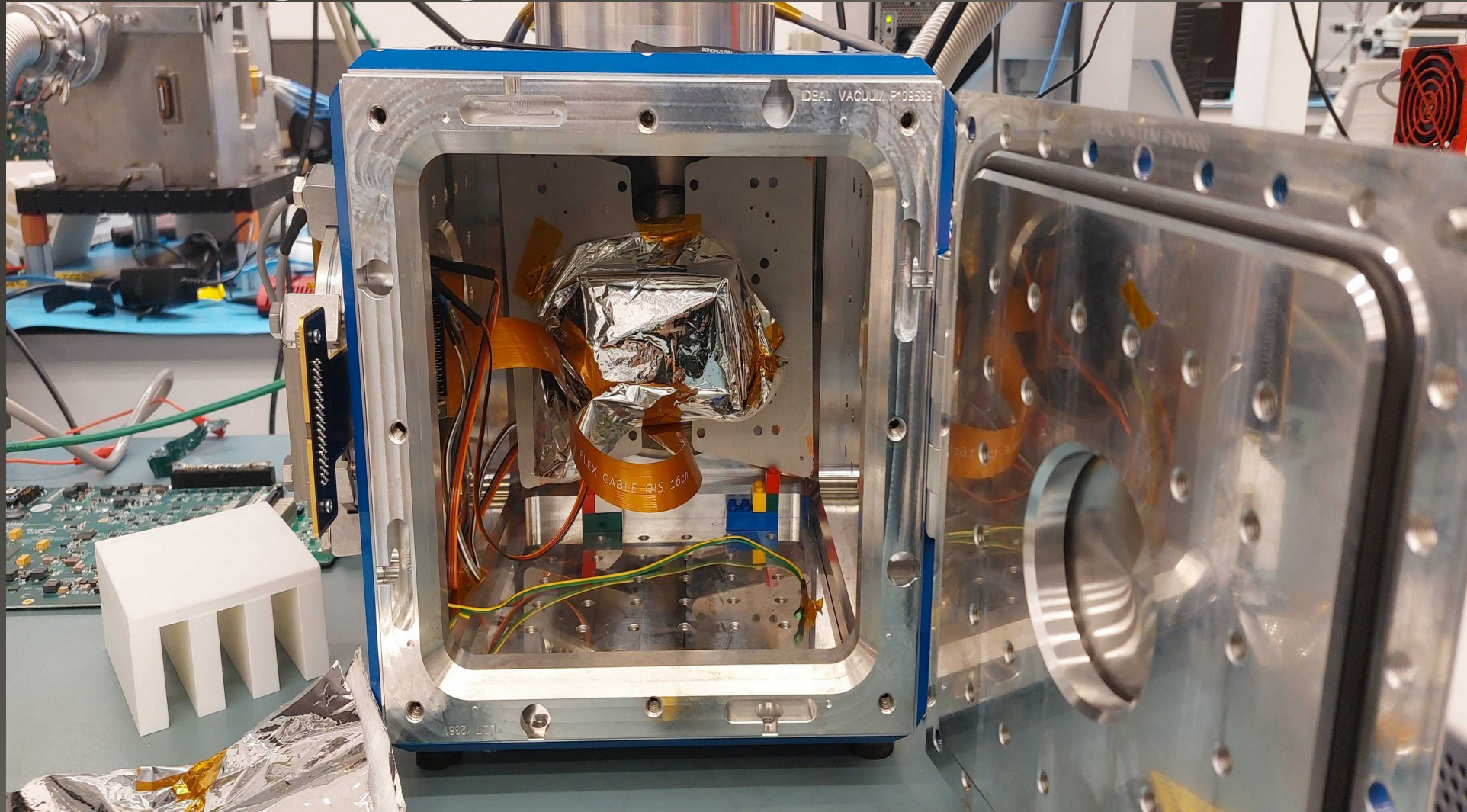
Skipper CCD Measurements of trans-Stilbene



Daniel Baxter, Alex Drlica-Wagner,  
Edgar Marrufo, Brandon Roach

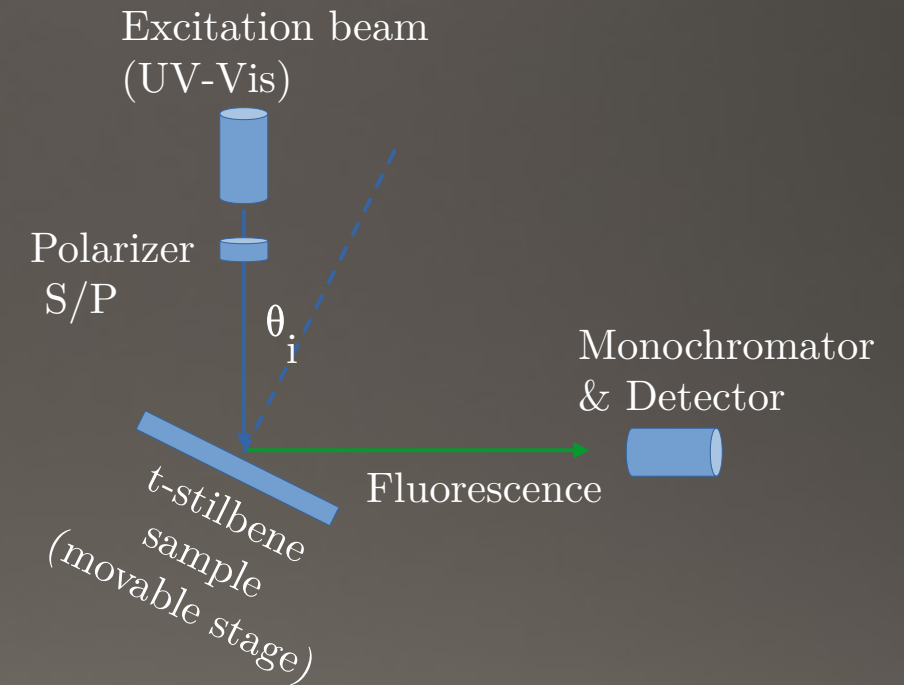
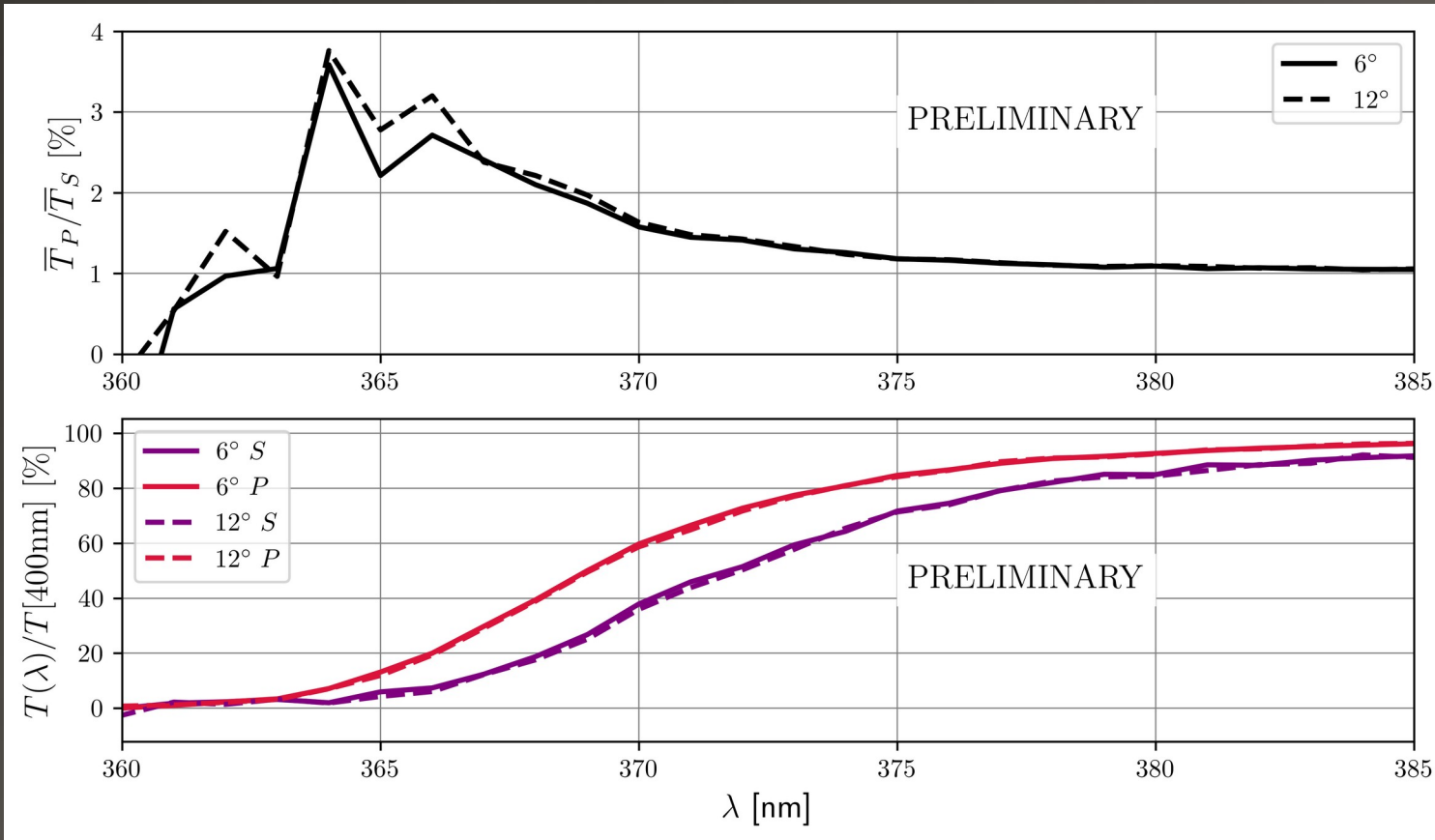
# Astro-Skipper CCDs as Readout

Now taking background measurements at Fermilab




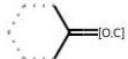
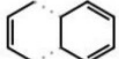
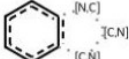
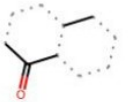

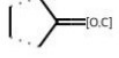

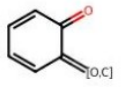
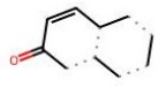

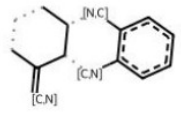
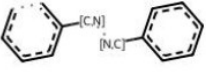
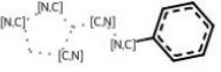

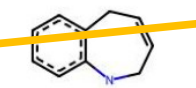
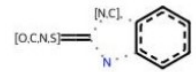
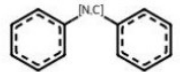
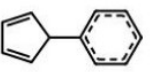
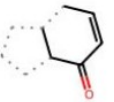
# Calibrating Anisotropic Materials

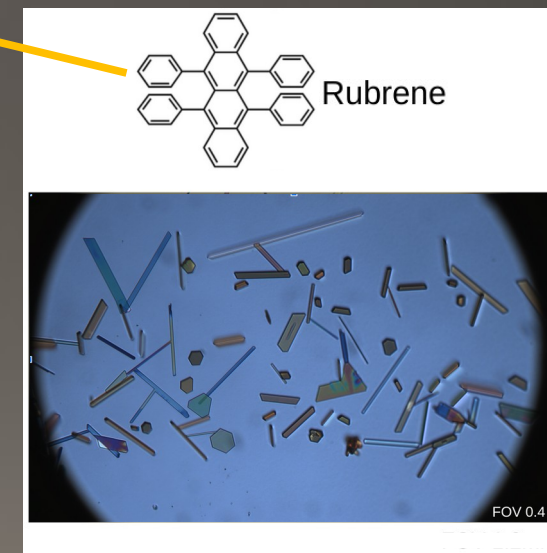
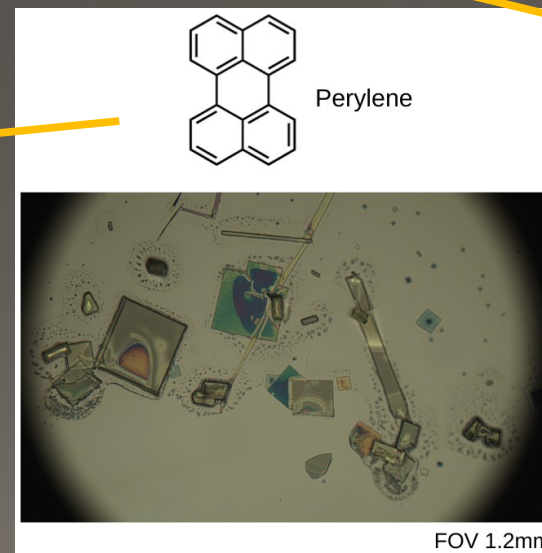
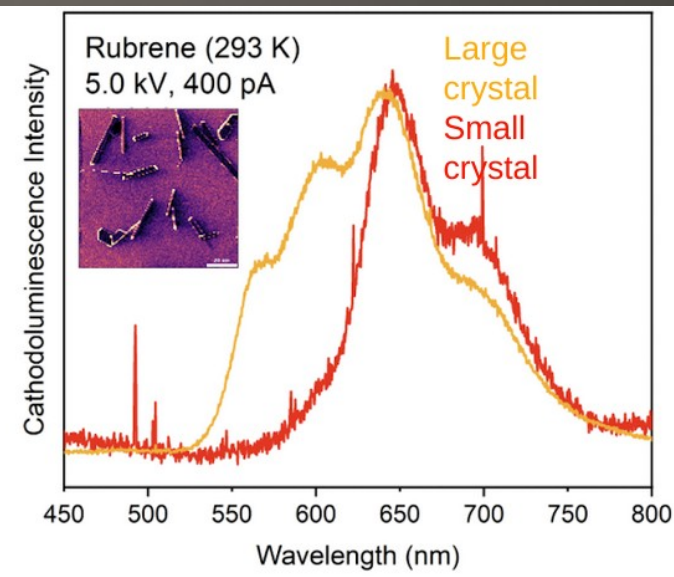
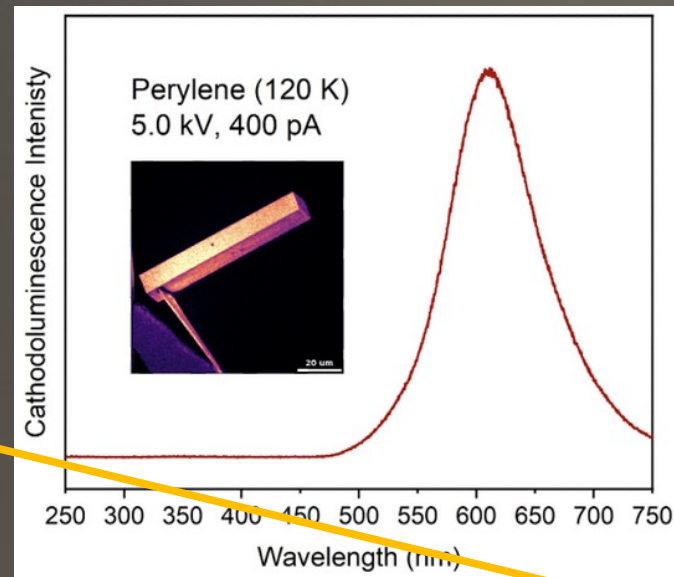
## Calibration of Anisotropy via Photoabsorption



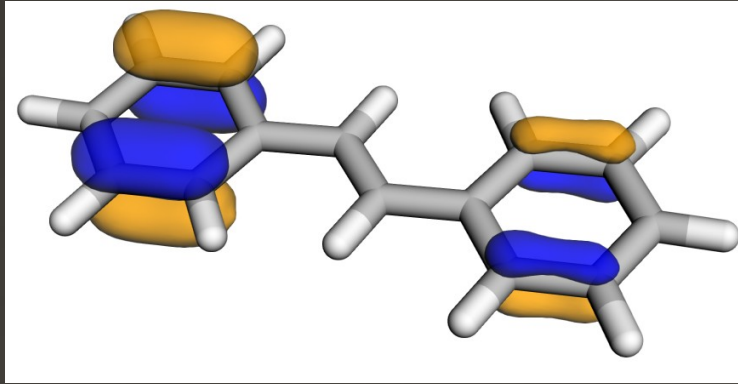
$$A_{UV} \sim n_{\text{mol}} |\vec{\epsilon} \cdot \langle \Psi_f | \mathbf{r} | \Psi_i \rangle|^2 \quad A_{UV} \approx \text{const} - T_{UV}$$

# New Target Materials

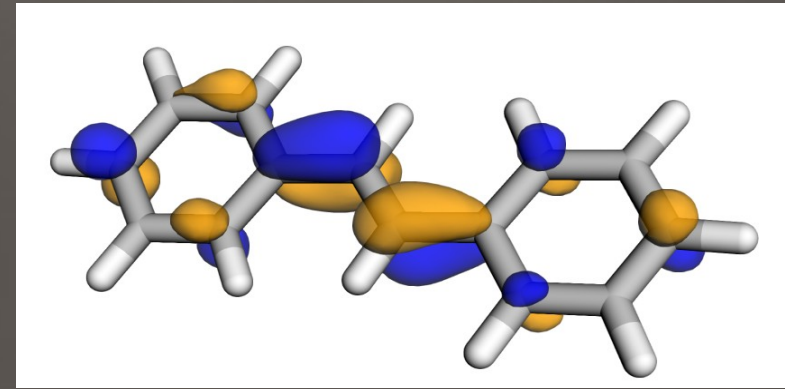
			
1. Count: 62465	2. Count: 34577	3. Count: 23404	4. Count: 20976
			
5. Count: 10118	6. Count: 8686	7. Count: 16948	8. Count: 23387
			
9. Count: 9292	10. Count: 7187	11. Count: 5760	12. Count: 4748
			
13. Count: 5508	14. Count: 5772	15. Count: 4468	16. Count: 5058
			
17. Count: 4584	18. Count: 4258	19. Count: 4383	20. Count: 3794



# Validating AI Predictions



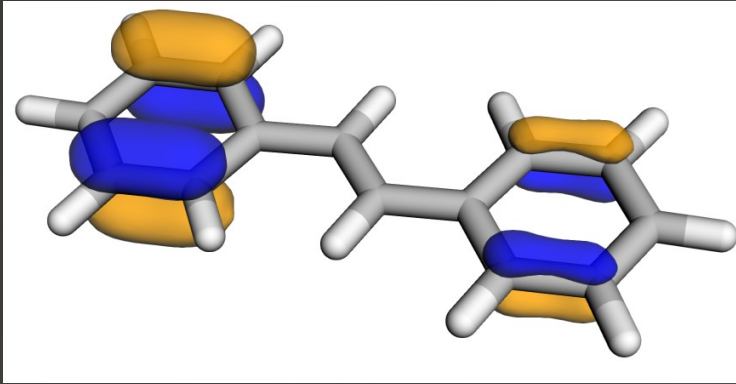
Ground State



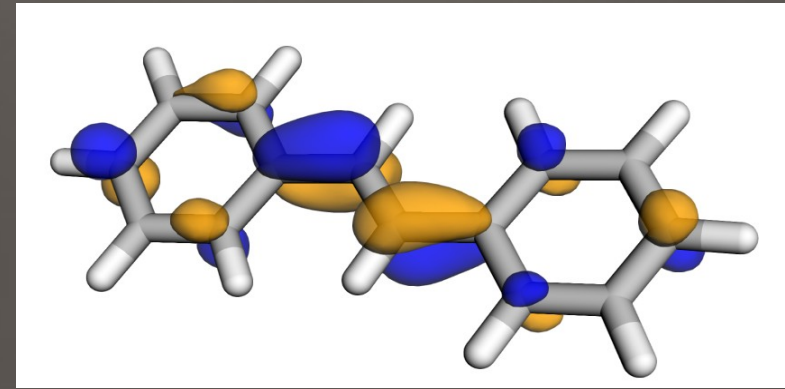
Excited State

Use time-dependent DFT to compute accurate many-body states

# Validating AI Predictions

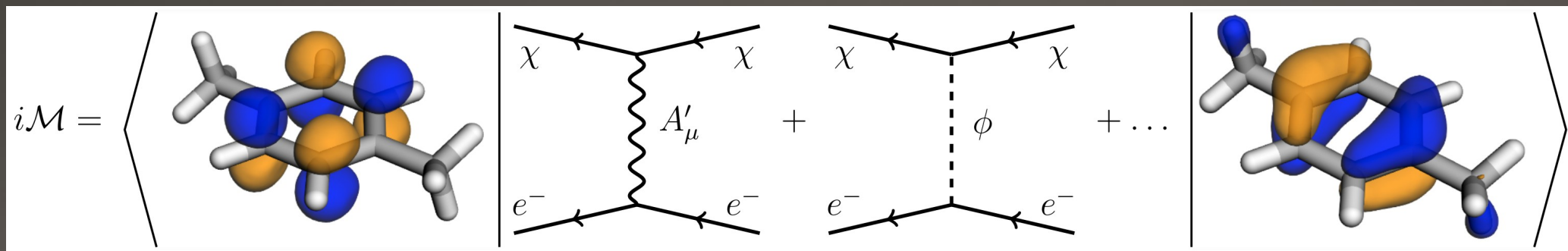


Ground State



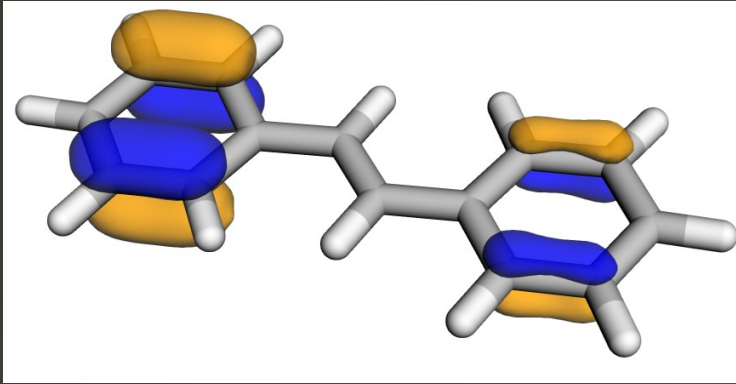
Excited State

Use time-dependent DFT to compute accurate many-body states

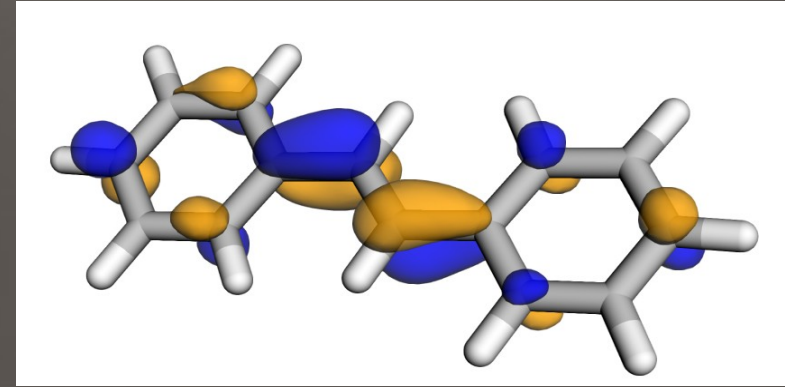


Then compute full matrix elements

# Validating AI Predictions

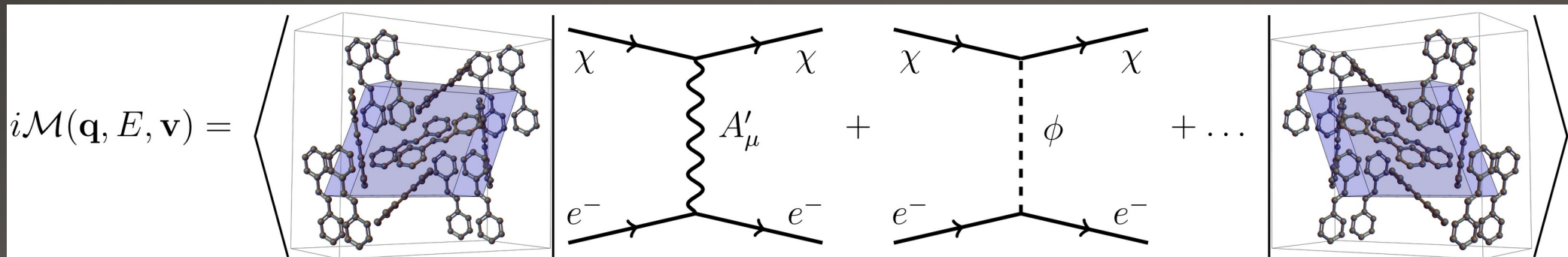


Ground State



Excited State

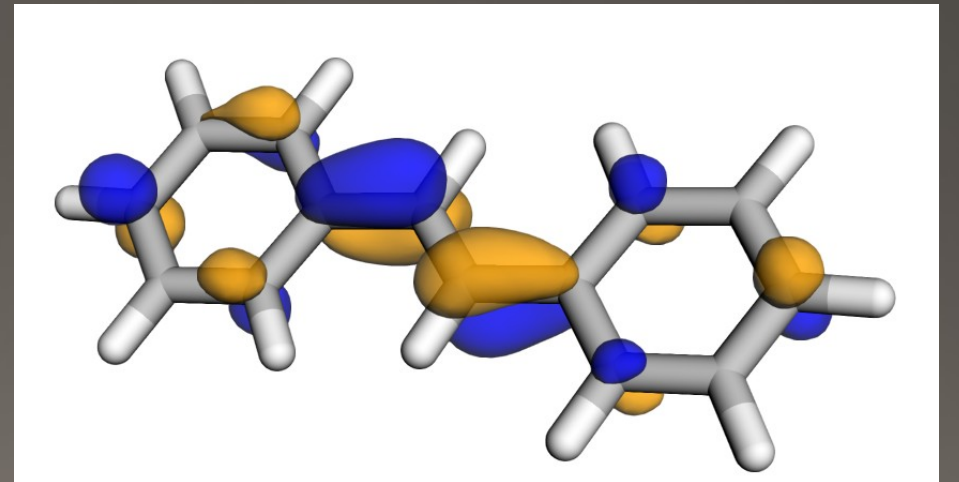
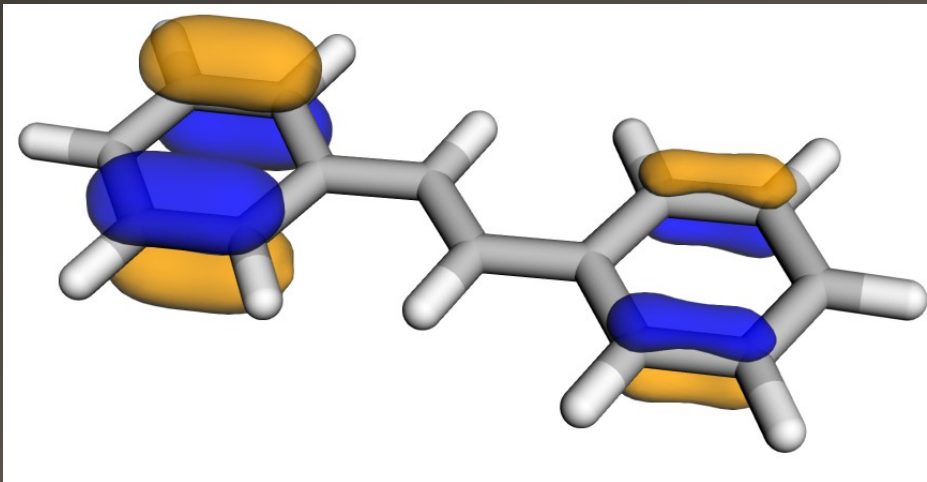
Use time-dependent DFT to compute accurate many-body states

$$i\mathcal{M}(\mathbf{q}, E, \mathbf{v}) = \left\langle \left[ \text{Crystal Structure} \right] \left[ \begin{array}{c} \chi \rightarrow \chi \\ \left. \begin{array}{c} \text{---} \\ \text{---} \end{array} \right\} A'_\mu \\ e^- \rightarrow e^- \end{array} \right] + \left[ \begin{array}{c} \chi \rightarrow \chi \\ \left. \begin{array}{c} \text{---} \\ \text{---} \end{array} \right\} \phi \\ e^- \rightarrow e^- \end{array} \right] + \dots \right] \right\rangle$$
The diagram illustrates the Dyson equation for the imaginary part of the conductivity,  $i\mathcal{M}(\mathbf{q}, E, \mathbf{v})$ . It is shown as a sum of terms enclosed in large square brackets. The first term is a diagram showing a crystal structure (a 3D lattice of atoms) with a blue shaded plane representing a Brillouin zone. This is followed by a vertex diagram where an incoming photon  $\chi$  and an incoming electron  $e^-$  meet at a vertex, and an outgoing photon  $\chi$  and an outgoing electron  $e^-$  meet at another vertex. The interaction between the photon and electron is represented by a wavy line labeled  $A'_\mu$ . The second term is a similar vertex diagram, but the interaction is represented by a dashed line labeled  $\phi$ . The diagram is followed by an ellipsis  $+\dots$  and another crystal structure diagram.

Then compute full matrix elements... accounting for crystal structure

# Conclusions

- 1) We have done an extremely effective job looking for WIMPs above a GeV, now we must look beyond.
- 2) By developing the formalism that describes the interaction between dark matter and molecules, we can develop detection strategies capable of *delving deep* and *searching wide* across the dark matter parameter space.



# ML for DM Direct Detection

Property prediction

Energies & Matrix elements

Step 1. Small molecules : Train on exhaustive database (< 9 atoms)

Characterize neural nets & architecture

→ Prove it's possible to learn from small subsample

Dataset of energies + oscillator strengths:

QM9: “All” 134K organic mols up to 9 heavy atoms - *computed* using DFT

# ML for DM Direct Detection

Property prediction

Energies & Matrix elements

Step 1. Small molecules : Train on exhaustive database (< 9 atoms)

Characterize neural nets & architecture

→ Prove it's possible to learn from small subsample

Dataset of energies + oscillator strengths:

QM9: "All" 134K organic mols up to 9 heavy atoms - *computed* using DFT

Step 2. Large molecules: Train on dataset up to 10s of atoms

Scale architecture

→ Generalize property prediction

Dataset of energies + oscillator strengths:

PubChemQ: "Sparse" ~3M organic molecules - *computed* using DFT

# Regression Model in Action

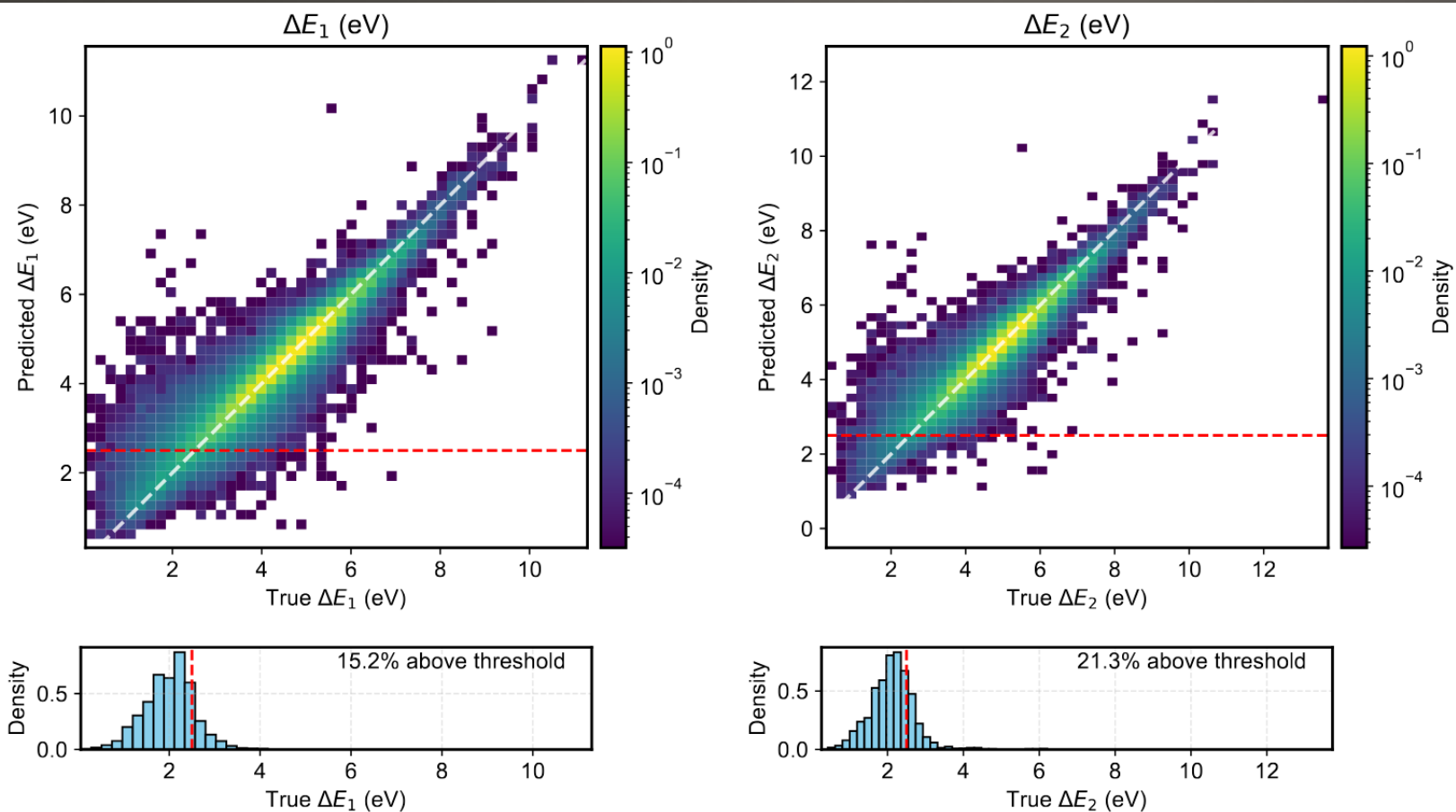


FIG. 19. We show the density of predicted values vs true values for the transition energies of our validation set, with a red line that corresponds to the threshold placed on the predicted value. The Histogram shows the distribution of true values for molecules whose predicted value is above the threshold.

# Regression Model in Action

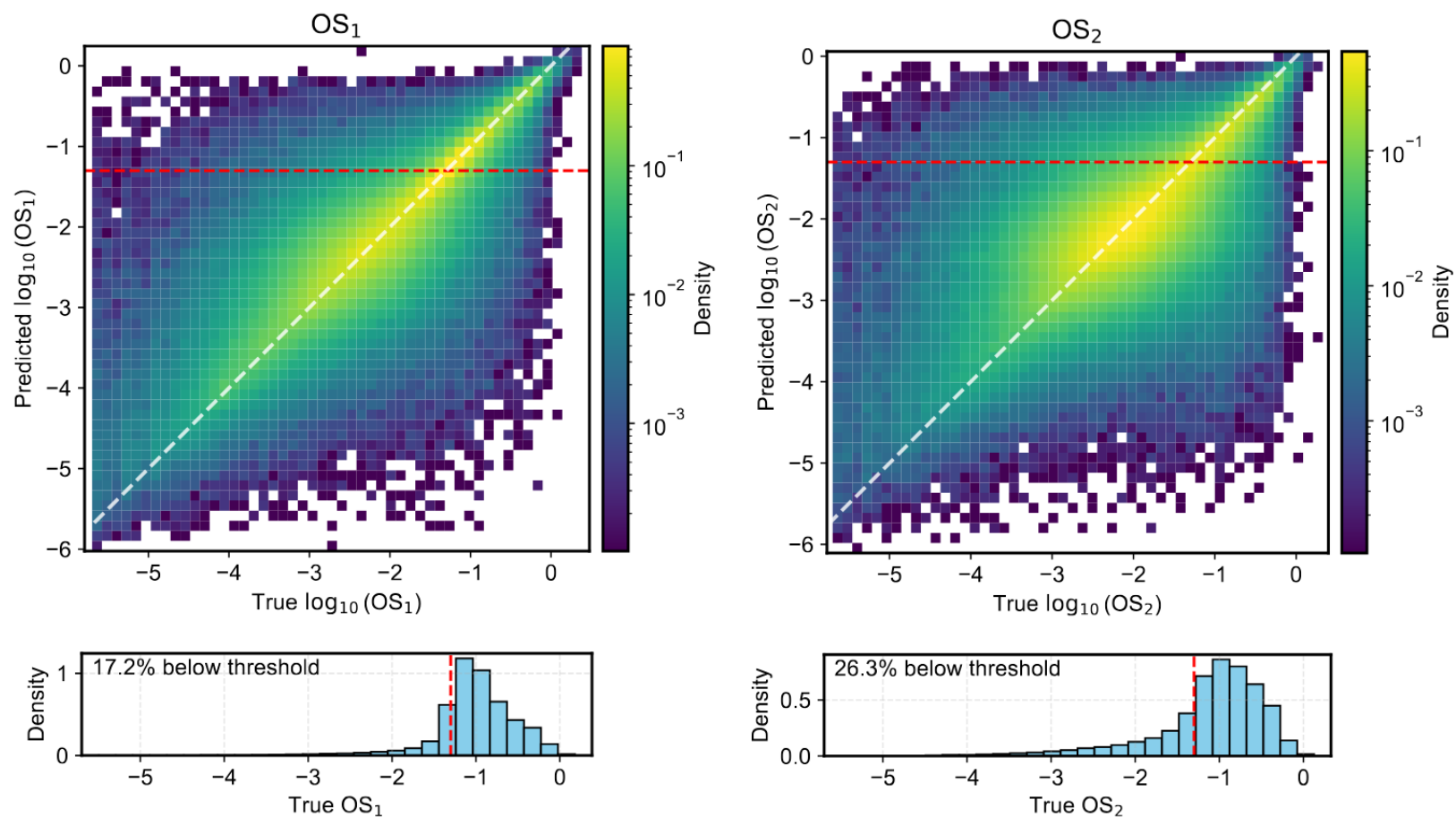


FIG. 20. We show the density of predicted values vs true values for the oscillator strength of our validation set, with a red line that corresponds to the threshold placed on the predicted value. The Histogram shows the distribution of true values for molecules whose predicted value is above the threshold.

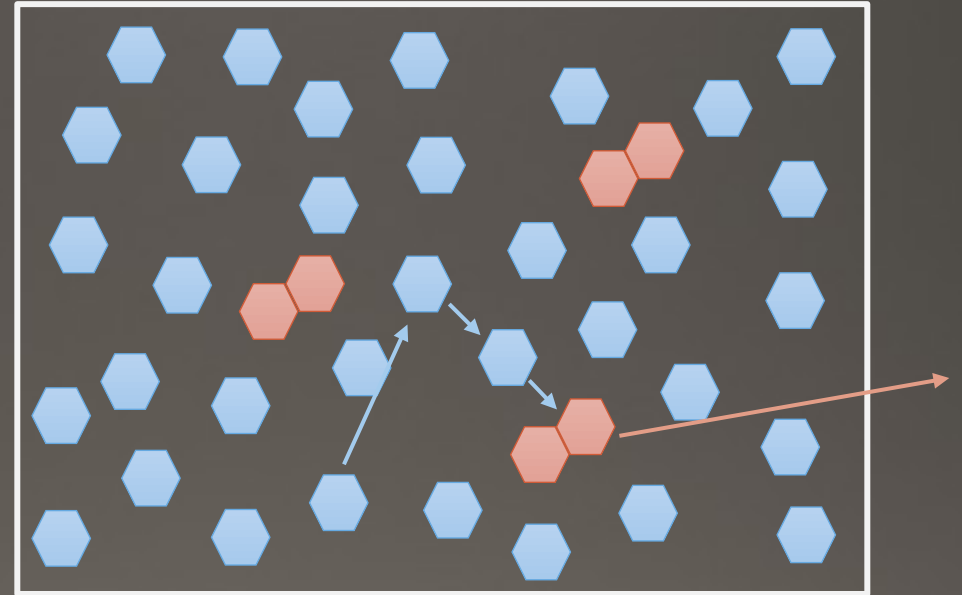
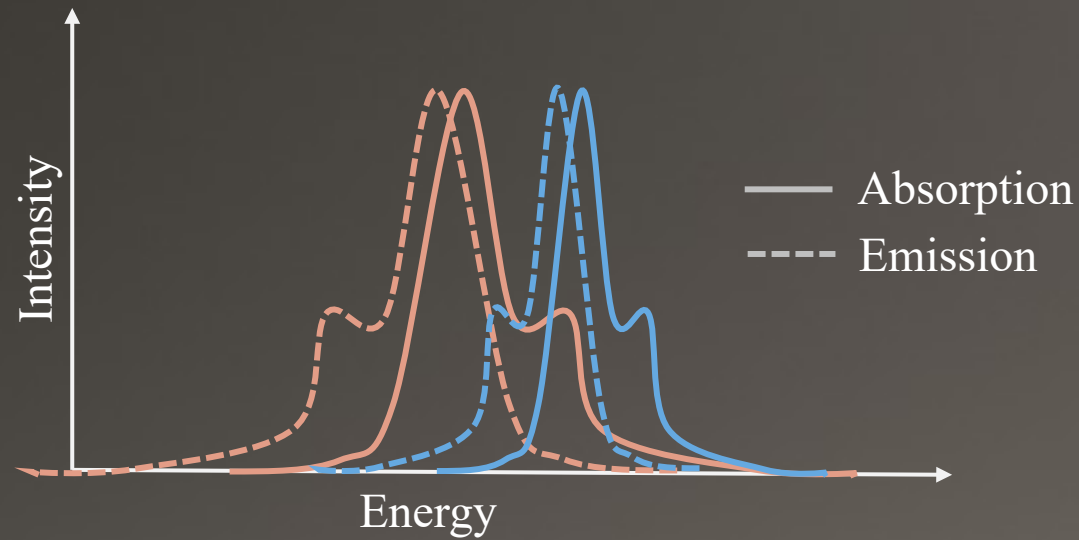
# Fluorescence: Binary Scintillators



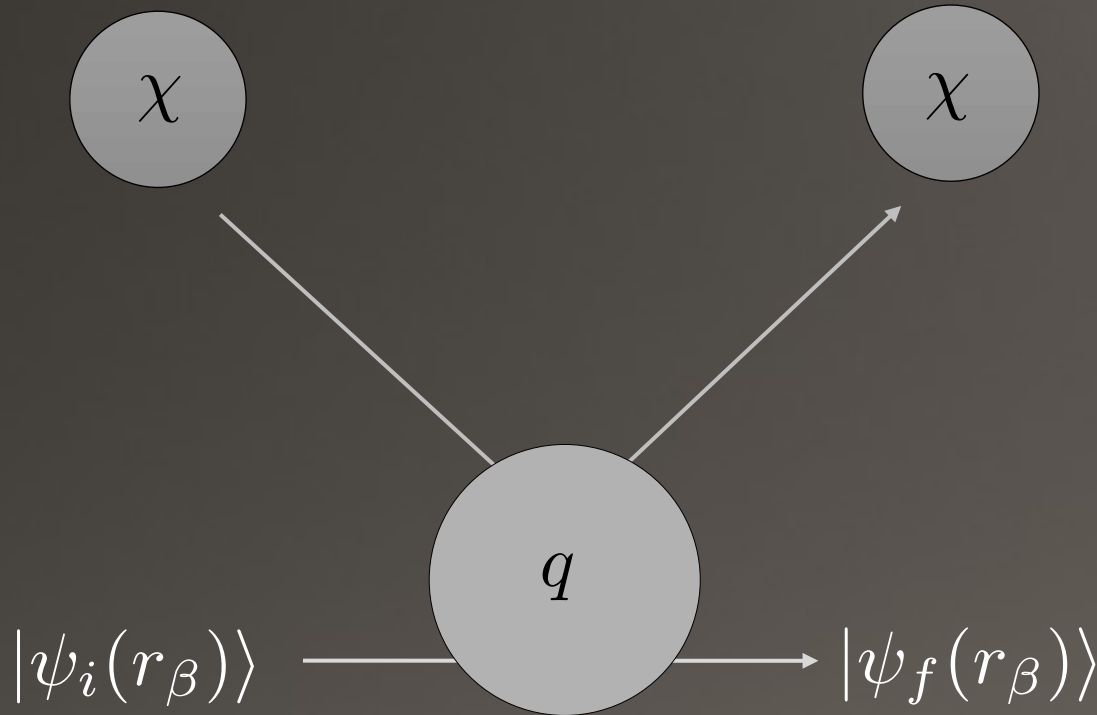
Solvent: Primary target starts the signal



Solute: Dilute fluor gets the signal out of the bulk



# Electron Recoil: Charge Signal



Electron scattering

$$\Delta E_r = (m_\chi^2 / m_T) \times 10^{-6}$$

$$\Delta E \sim \mathcal{O}(\text{few eV}) \left( \frac{m_\chi}{1 \text{ MeV}} \right)^2$$

What has such transition energies?

- Semiconductor band gaps
- Maybe atomic ionization

Electrons in crystals (exciton generation)

Electrons in atoms (ionization)

$$|\psi_i\rangle \sim u_v(r) e^{ik' \cdot r}$$

$$|\psi_f\rangle \sim u_c(r) e^{ik \cdot r}$$

$$|\psi_i\rangle \sim \psi_{\text{STO}}(r_\beta)$$

$$|\psi_f\rangle \sim e^{ik \cdot r}, r \gg a_0$$

# Trans-Stilbene

$s$	Platt Symbol	Symmetry	$\Delta E$ [eV]	Configuration amplitudes			
$s_1$	${}^1B$	$B_u$	4.240	$d_{7,8} = 0.94,$	$d_{4,11} = -0.24$		
$s_2$	${}^1G^-$	$B_u$	4.788	$d_{7,10} = 0.53,$	$d_{5,8} = 0.53,$	$d_{6,11} = 0.37,$	$d_{4,9} = -0.37$
$s_3$	${}^1G^-$	$A_g$	4.800	$d_{7,9} = 0.53,$	$d_{6,8} = 0.53,$	$d_{5,11} = 0.37,$	$d_{4,10} = -0.37$
$s_4$	${}^1(C, H)^+$	$A_g$	5.137	$d_{7,11} = 0.41,$	$d_{5,9} = -0.41,$	$d_{6,10} = -0.41,$	$d_{4,8} = -0.59$
$s_5$	${}^1H^+$	$B_u$	5.791	$d_{5,10} = 0.54,$	$d_{6,9} = 0.54,$	$d_{7,12} = 0.33,$	$d_{3,8} = 0.33$
$s_6$	${}^1G^+$	$A_g$	6.264	$d_{7,9} = 0.68,$	$d_{6,8} = -0.68$		
$s_7$	${}^1C^-$	$A_g$	6.013	$d_{7,11} = 0.66,$	$d_{4,8} = 0.54,$		
$s_8$	${}^1G^+$	$B_u$	6.439	$d_{7,10} = 0.65,$	$d_{5,8} = -0.65$		

Table 1: The first eight excited states  $s_{n=1\dots 8}$ , with their energy eigenvalues  $\Delta E(s_n)$  with respect to the ground state and coefficients  $d_{ij}^{(n)}$  as calculated by Ting and McClure.

$$|s_n\rangle = \sum_{i,j>i} d_{ij}^{(n)} |\psi_i^j\rangle,$$

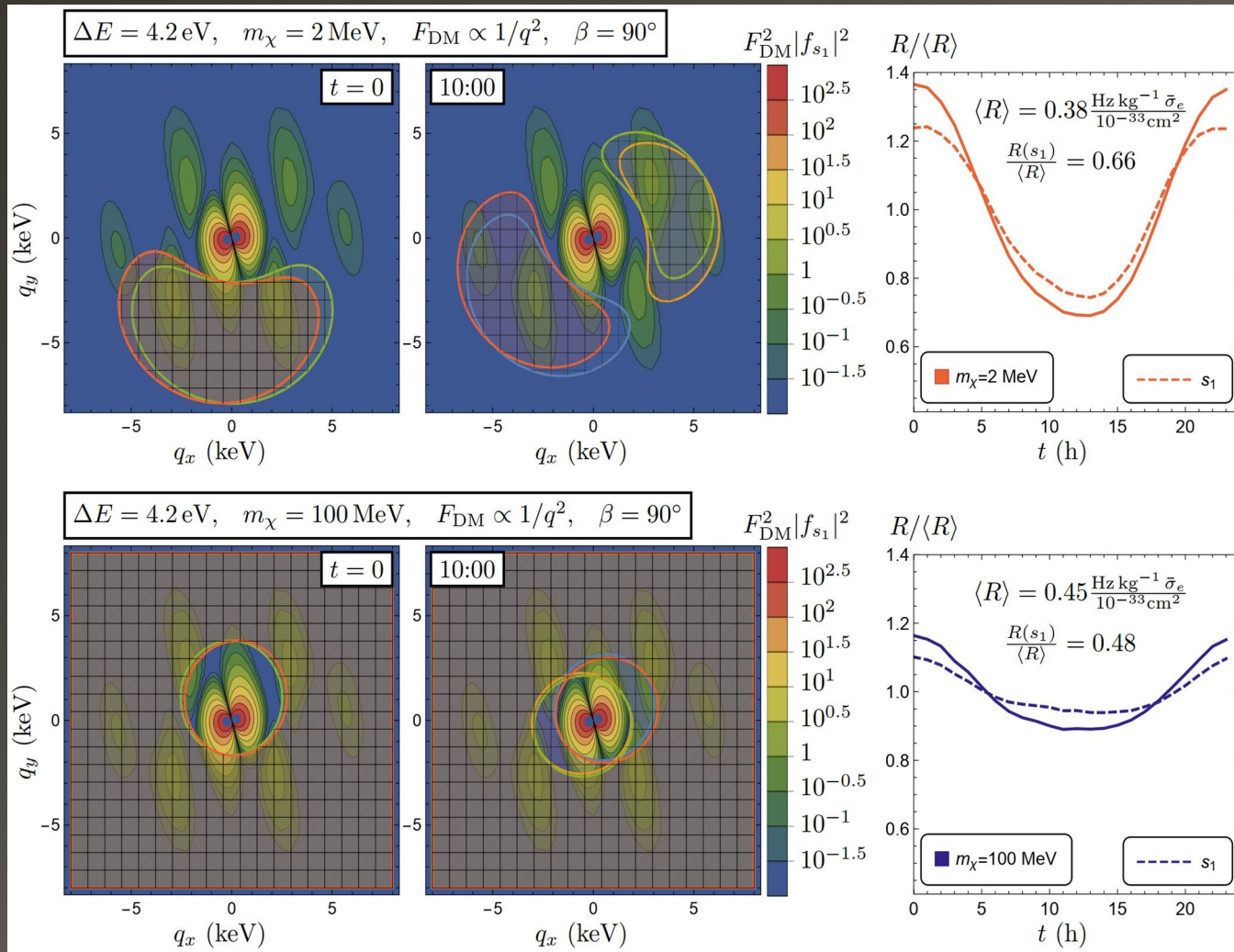
$$\sum_{ij} |d_{ij}^{(n)}|^2 = 1.$$

$$f_{g \rightarrow s_n}(\vec{q}) = \left\langle \psi_{s_n}(\vec{r}_1 \dots \vec{r}_{14}) \left| \sum_{m=1}^{14} e^{i\vec{q} \cdot \vec{r}_m} \right| \psi_G(\vec{r}_1 \dots \vec{r}_{14}) \right\rangle$$

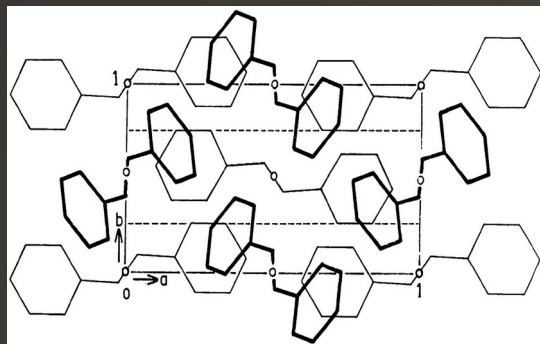
$$= \sum_{ij} d_{ij}^{(n)} \langle \psi_i^j | e^{i\vec{q} \cdot \vec{r}} | \psi_G \rangle$$

$$= \sqrt{2} \sum_{ij} d_{ij}^{(n)} \langle \Psi_j(\vec{r}) | e^{i\vec{q} \cdot \vec{r}} | \Psi_i(\vec{r}) \rangle.$$

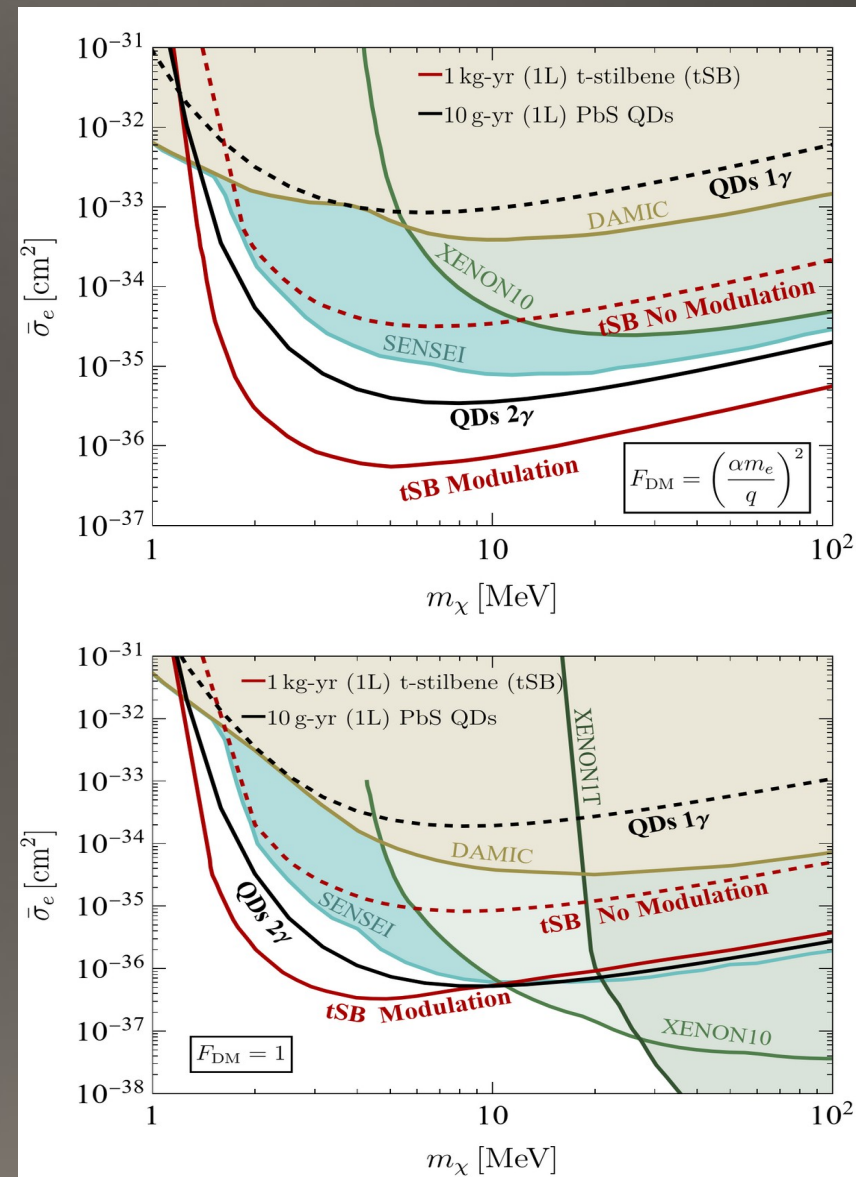
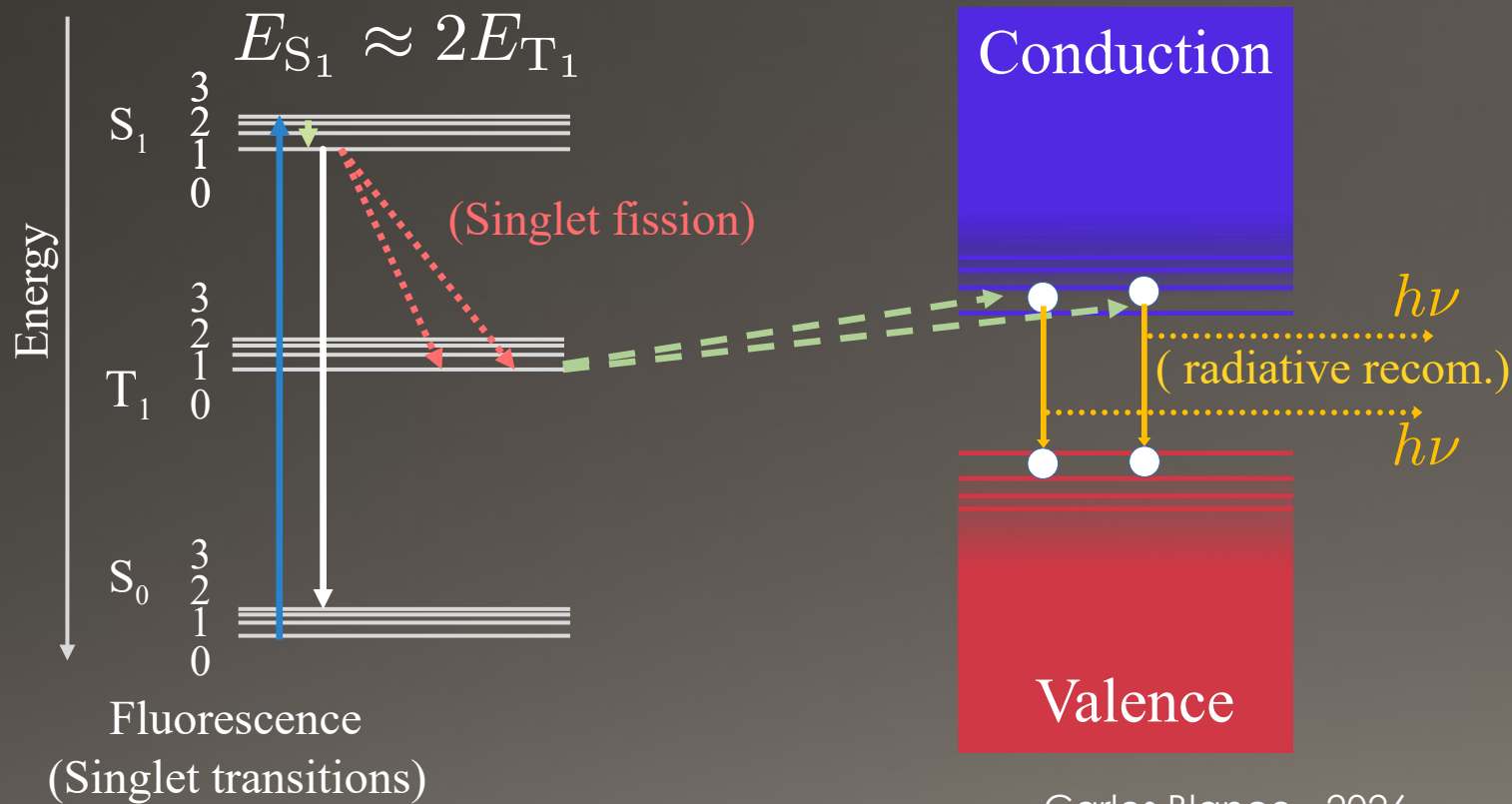
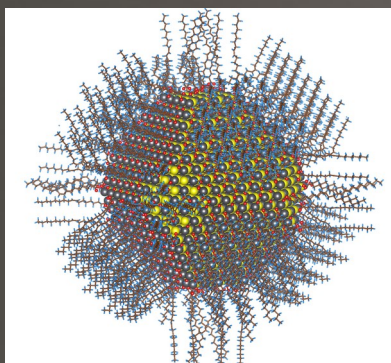
# Daily Modulation: Light Mediator



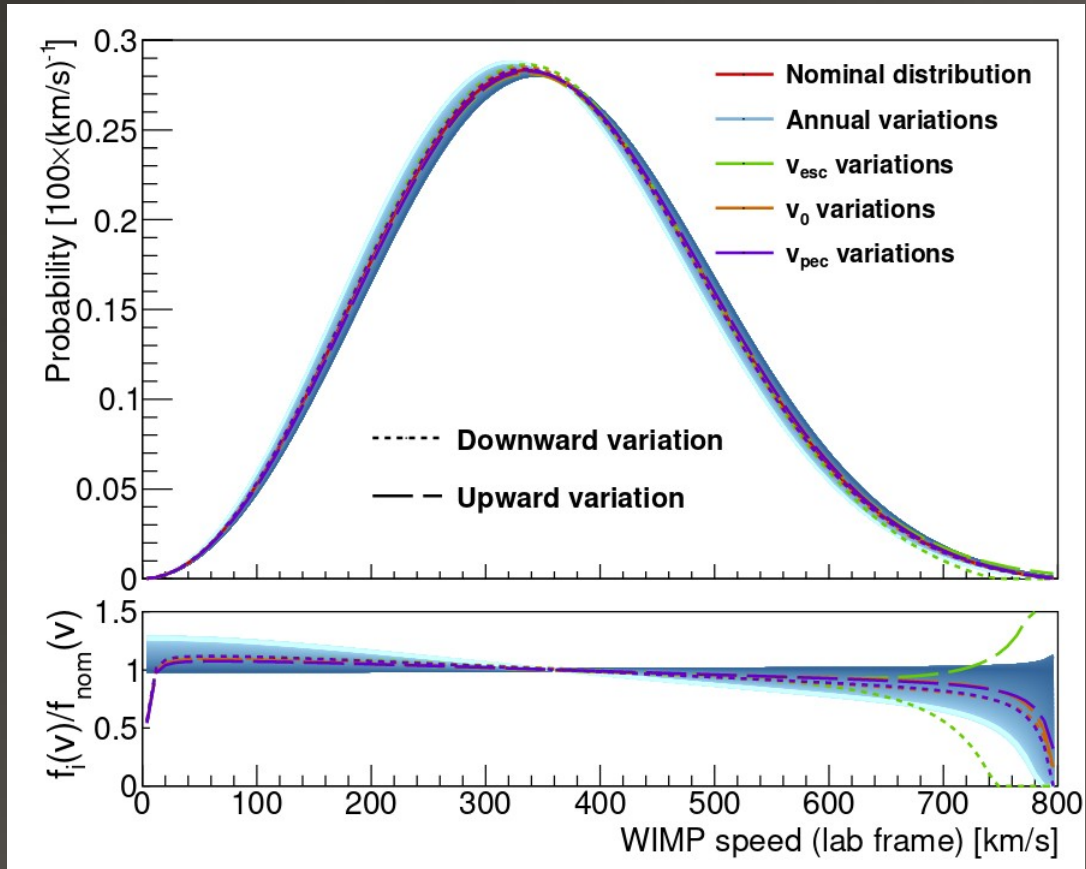
Same as previous figures (top) for a light mediator DM form factor  $F_{\text{DM}} = (\alpha m_e / q)^2$ . Here, the contour plots show  $F_{\text{DM}}^2 |f(s_1)|^2$  which appears in the rate integrand; the scattering is dominated by the smallest kinematically-allowed  $q$ . **Top:** Molecular form factors with  $q_z = 0$  and rate modulations for  $m_\chi = 2 \text{ MeV}$ . **Bottom:** Molecular form factors with  $q_z = 0$  and rate modulations for  $m_\chi = 100 \text{ MeV}$ .



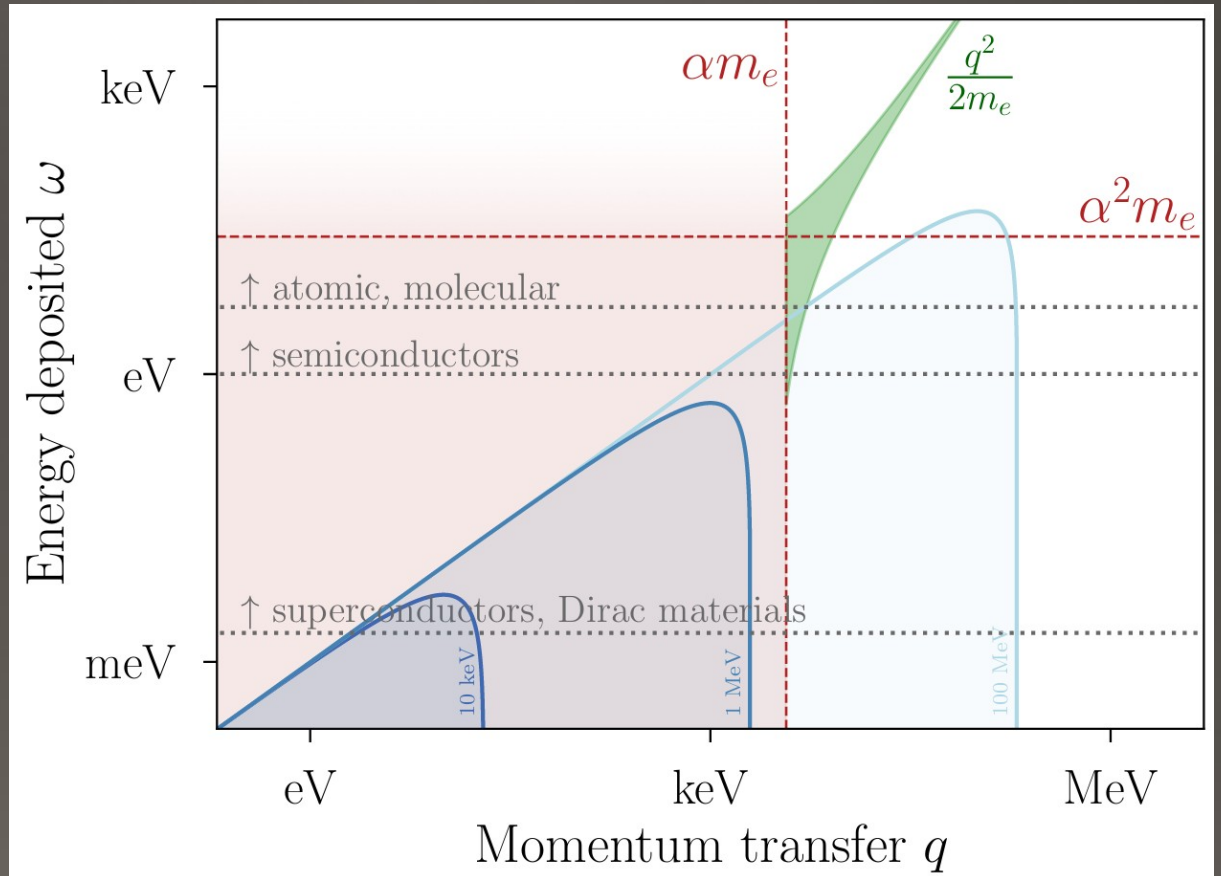
(Triplet diffusion)



# Local DM Phase Space



Baxter, D., et al. "Recommended conventions for reporting results from direct dark matter searches." The European Physical Journal C 81.10 (2021): 1-19.



Lin, Tongyan. "Sub-GeV dark matter models and direct detection." SciPost Physics Lecture Notes (2022): 043.