

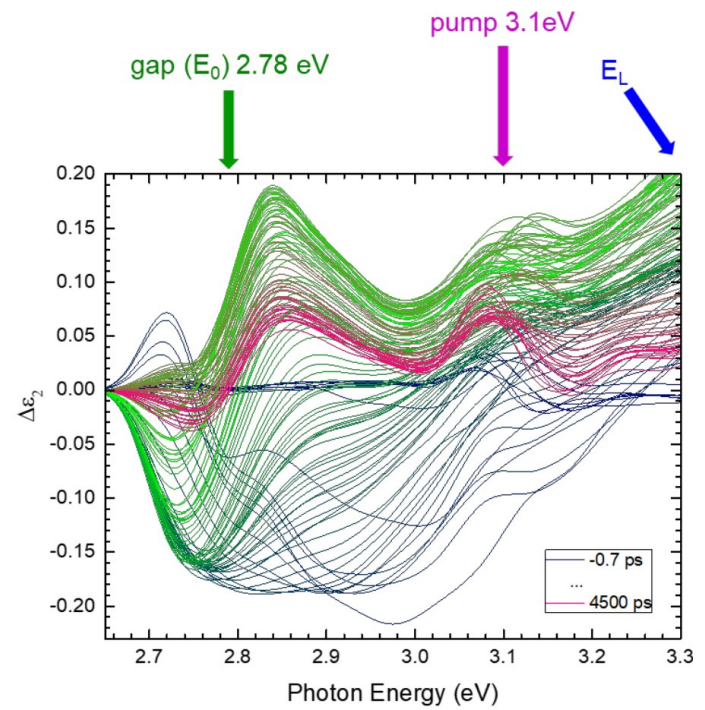
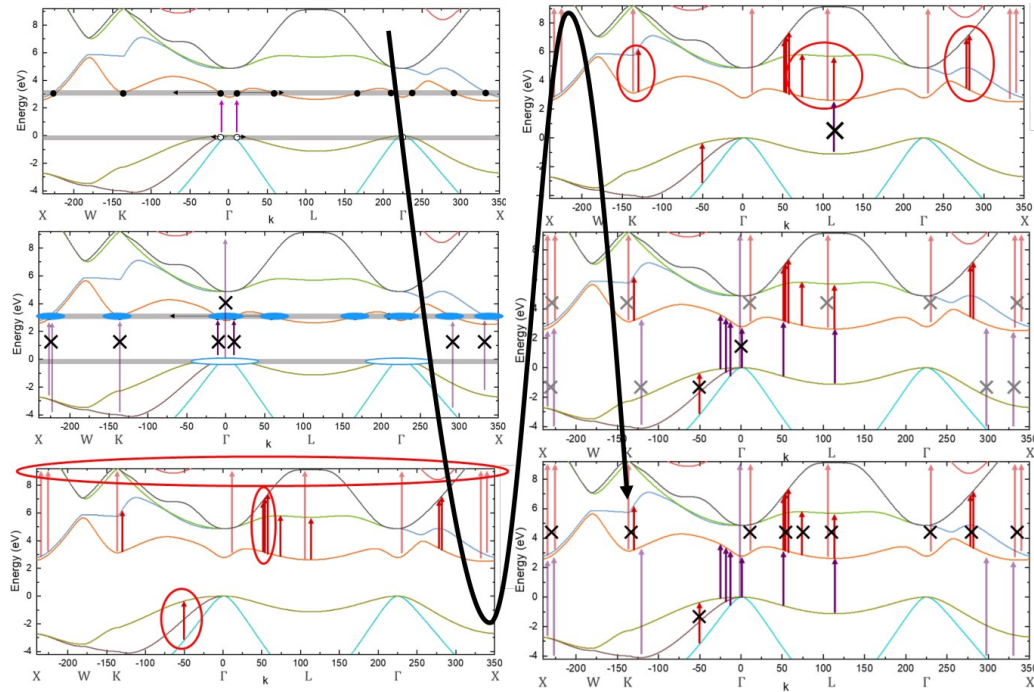
A Graph-Based Model Building Approach for Time-Resolved Ellipsometry

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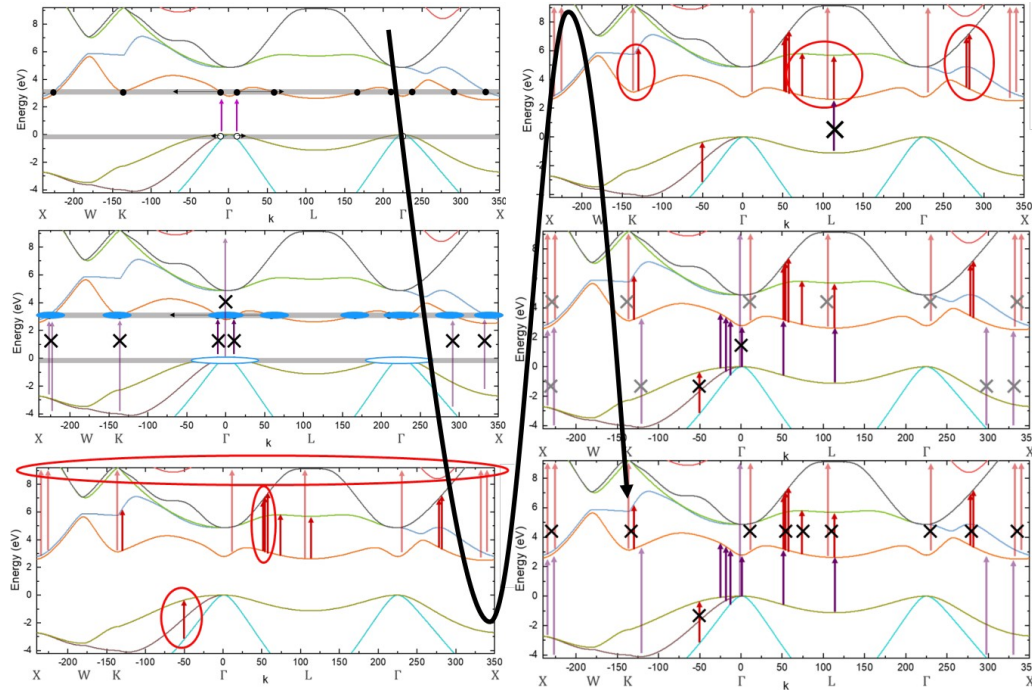
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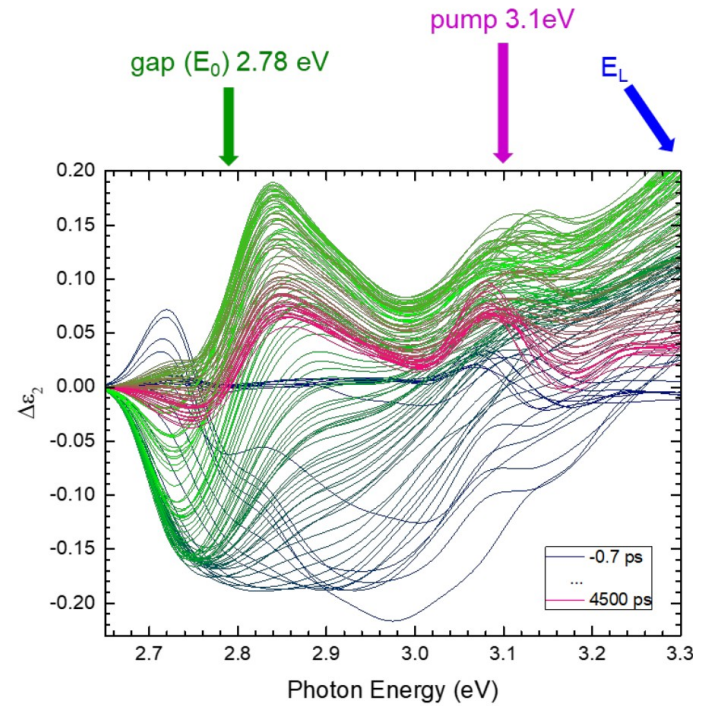
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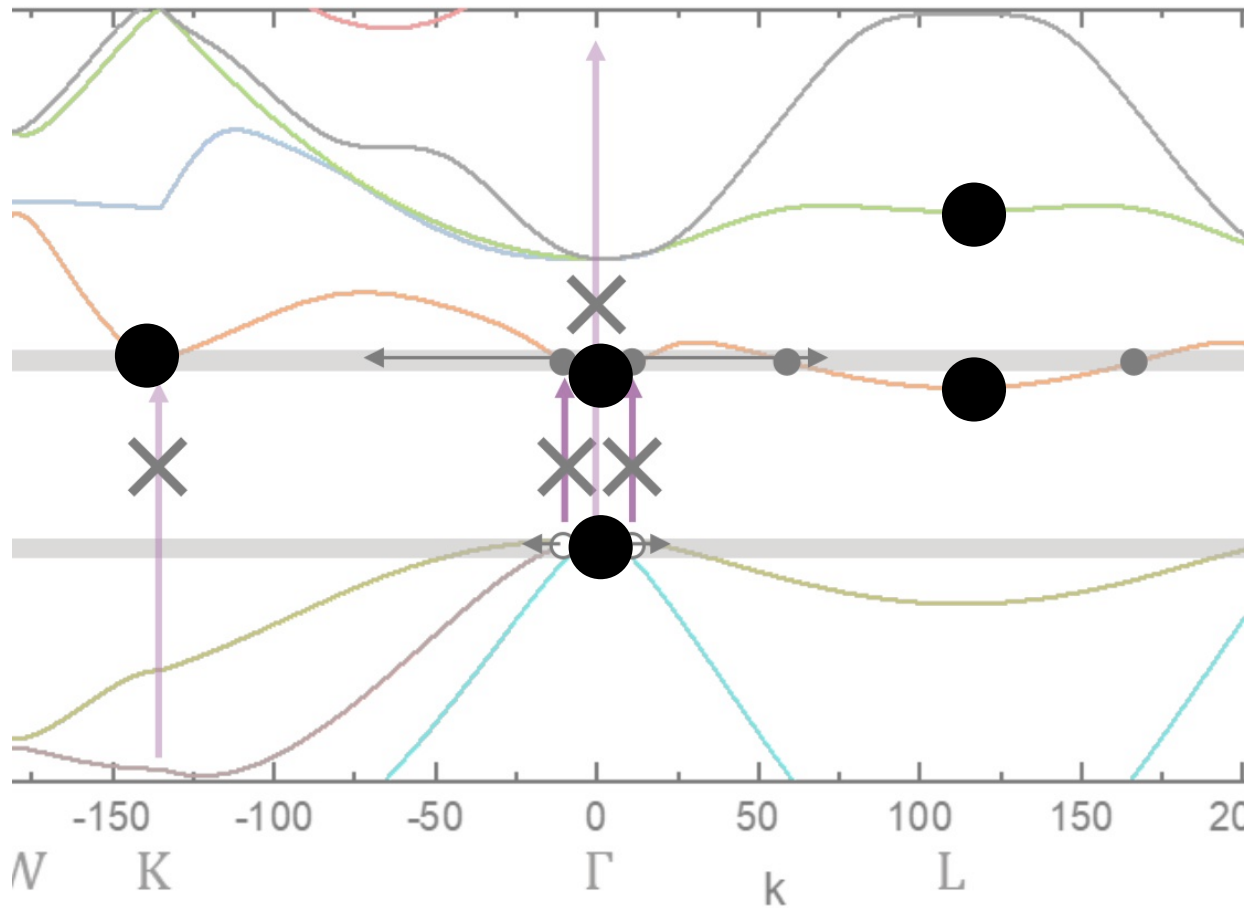


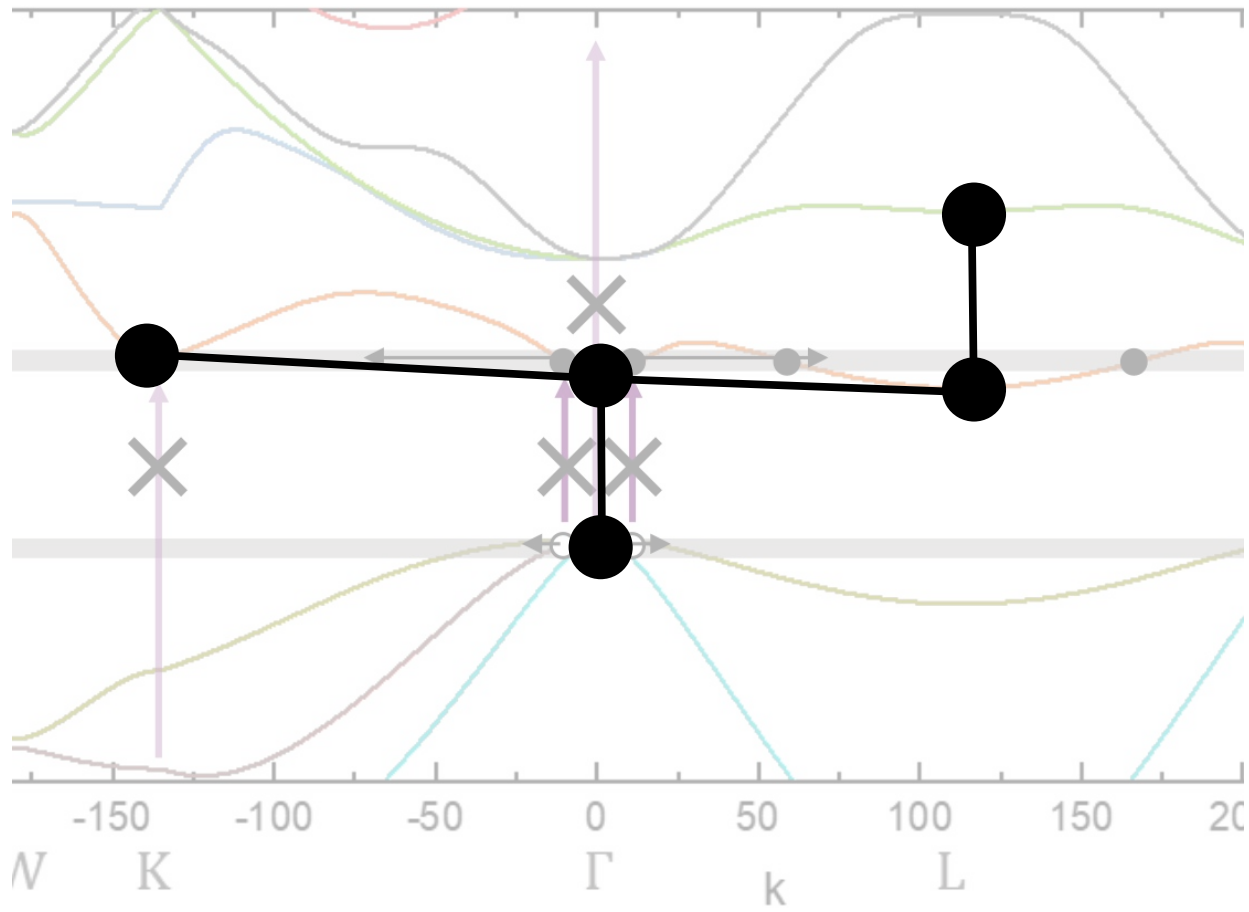
mental model

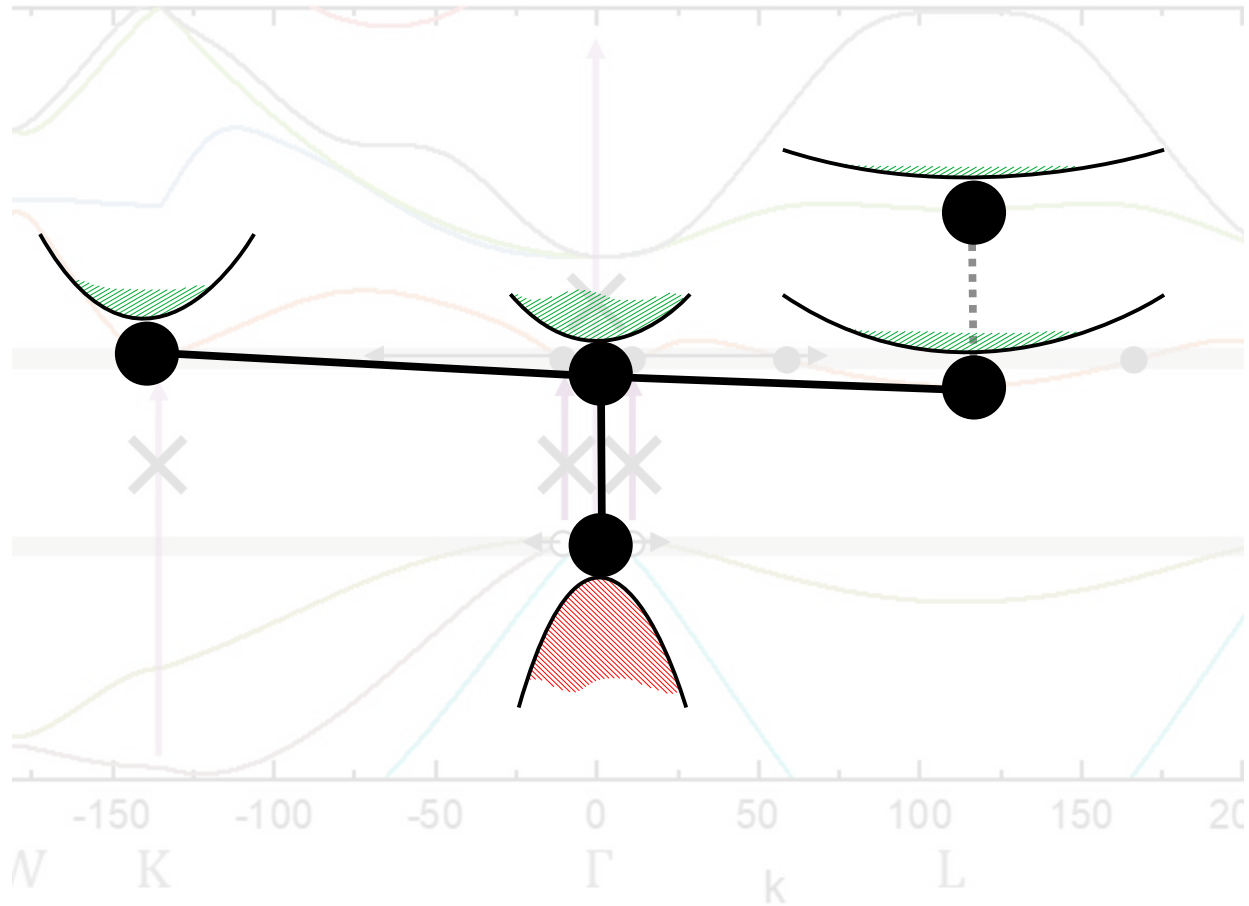


“steady state” model(s)

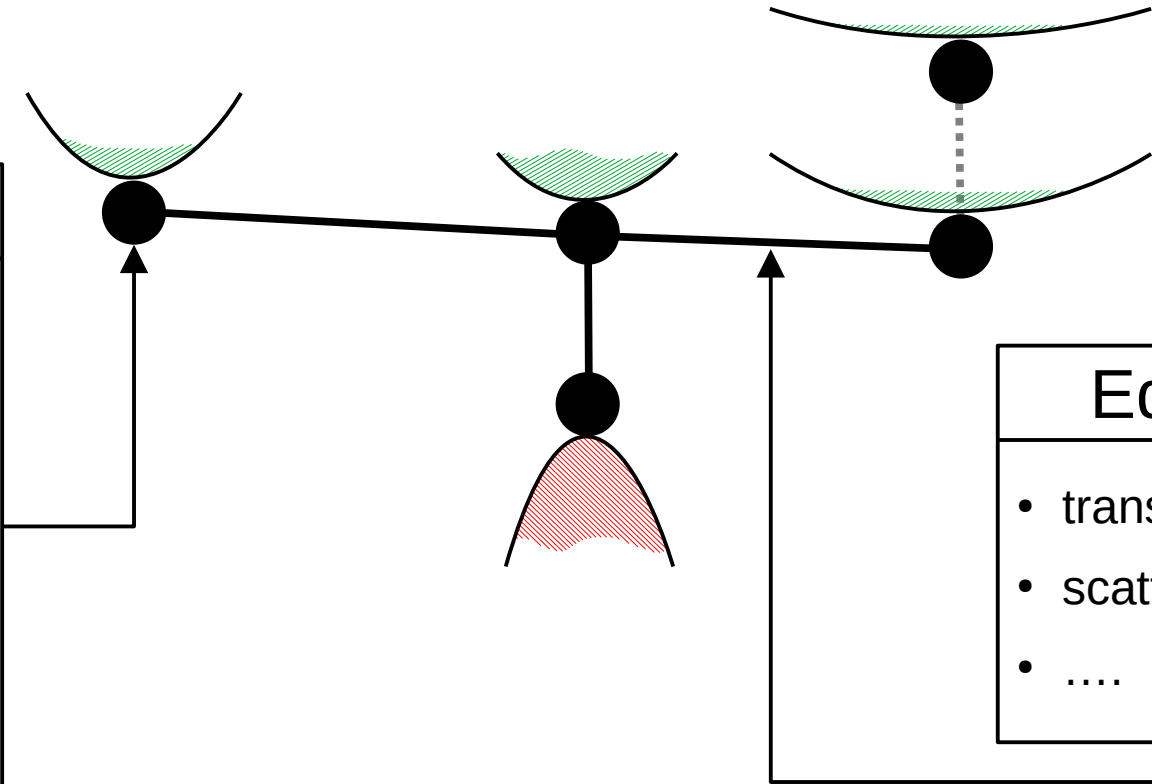








Node Data
<ul style="list-style-type: none"> • effective mass • (k-)volume • population • carrier temp • BGR / BMS •



Edge Data
<ul style="list-style-type: none"> • transition type • scattering rate •

$$\dot{n}_m = \sum_{i \rightarrow m} g(\Delta E_{i \rightarrow m}, t) \cdot \gamma_{i \rightarrow m} \cdot n_i (N_m - n_m) - \sum_{m \rightarrow f} g(\Delta E_{m \rightarrow f}, t) \cdot \gamma_{m \rightarrow f} \cdot n_m (N_f - n_f)$$

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$$\varepsilon(\omega) = \sum_k M_k(\omega; n_i, n_f [, \vec{p}(t)])$$

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$$\varepsilon(\omega) = \sum_k M_k(\omega; n_i, n_f [, \vec{p}(t)])$$

$$\dot{\varepsilon}(\omega) = \sum_k \left[\frac{\partial M_k(\omega)}{\partial n_i} \dot{n}_i + \frac{\partial M_k(\omega)}{\partial n_f} \dot{n}_f \right] \left[+ \frac{\partial M_k(\omega)}{\partial \vec{p}} \dot{\vec{p}} \right]$$

Compatibility / Extensibility of the Model

$$\begin{aligned}\dot{n} &= \dots \\ \dot{\varepsilon} &= \dots \\ \dot{T} &= \dots \\ \dot{\Delta} &= \dots \\ &\vdots\end{aligned}$$

- Extending the model to include more effects (approximately) is in principle straightforward
- Having the ODE system enables access to a large ecosystem of ODE analysis and parameter estimation tools (nonlinear optimization, particle swarm, Bayesian, neural nets, etc...)
- Including spatial diffusion is also possible at higher computational costs (solving one instance of the model for every mesh element in a finite elements / finite differences simulation)

“Node Based” User Interface Design Philosophy

Do not forget to show the video here 😊

Conclusion

- A system of ODEs is a flexible description of TRSE measurements

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Thank you for your attention!