

# The ComPW project

## Speeding up amplitude analysis with a Computer Algebra System

REMCO DE BOER<sup>1</sup>, MIRIAM FRITSCH<sup>1</sup>, WOLFGANG GRADL<sup>2</sup>, STEFAN PFLÜGER<sup>1</sup>, and LEONARD WOLLENBERG<sup>1</sup>  
<sup>1</sup>Ruhr-Universität Bochum — <sup>2</sup>Johannes Gutenberg Universität Mainz

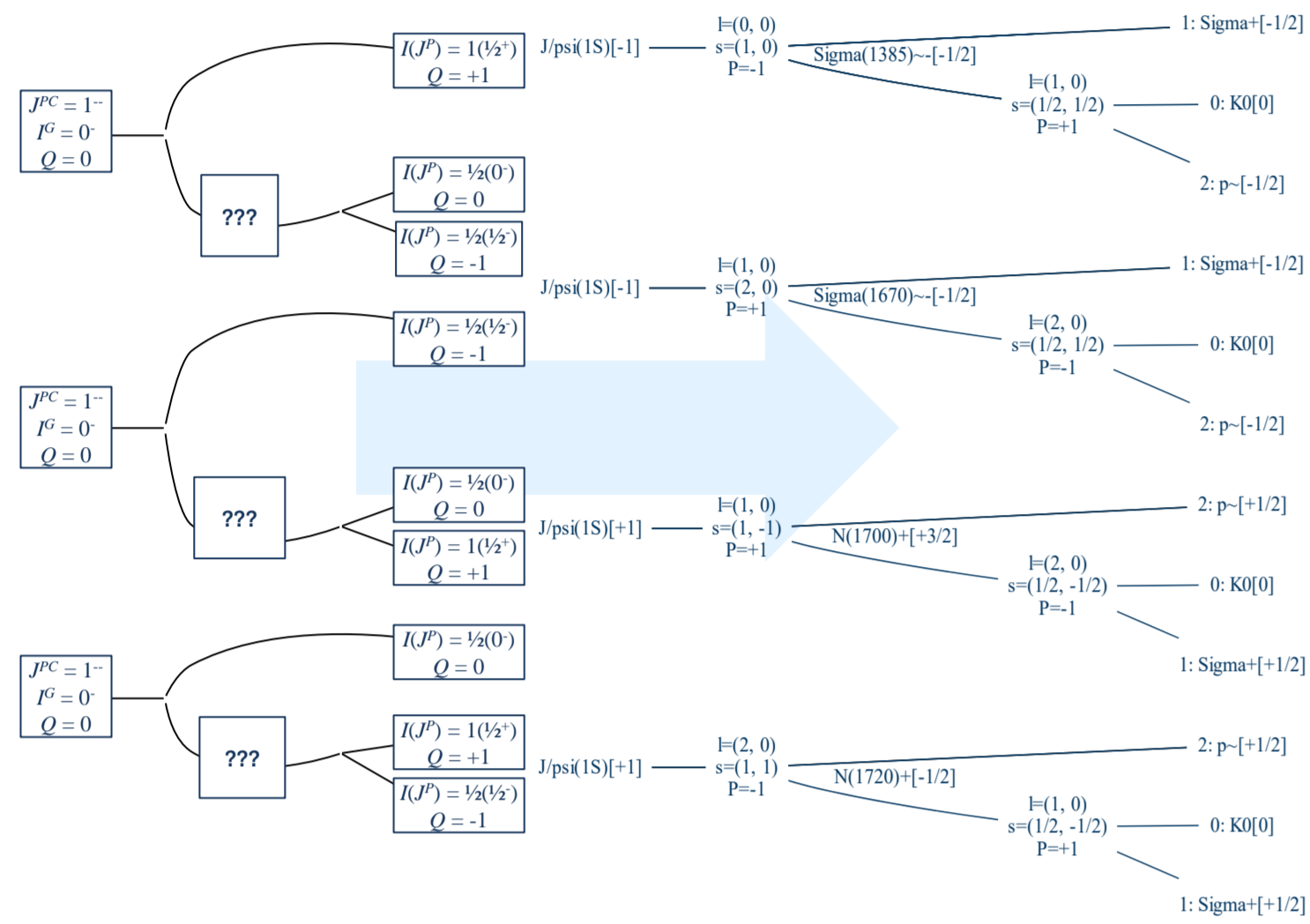
remco.deboer@rub.de

### Three Python Libraries for a full amplitude analysis

#### QRules: Automated quantum number conservation rules

```
reaction = qrules.generate_transitions(
    initial_state="J/psi(1S)",
    final_state=["K0", "Sigma+", "p-"],
    allowed_interaction_types=["strong"],
)
```

Input: boundary conditions  
• initial and final state particles  
• optional intermediate state restrictions



Output: state transitions  
• Determines possible decay topologies  
• Gets corresponding particle properties from the PDG (or custom definitions)  
• Propagates quantum numbers through intermediate states  
• Selects all allowed transitions with its conservation laws

qrules.rtdf.io

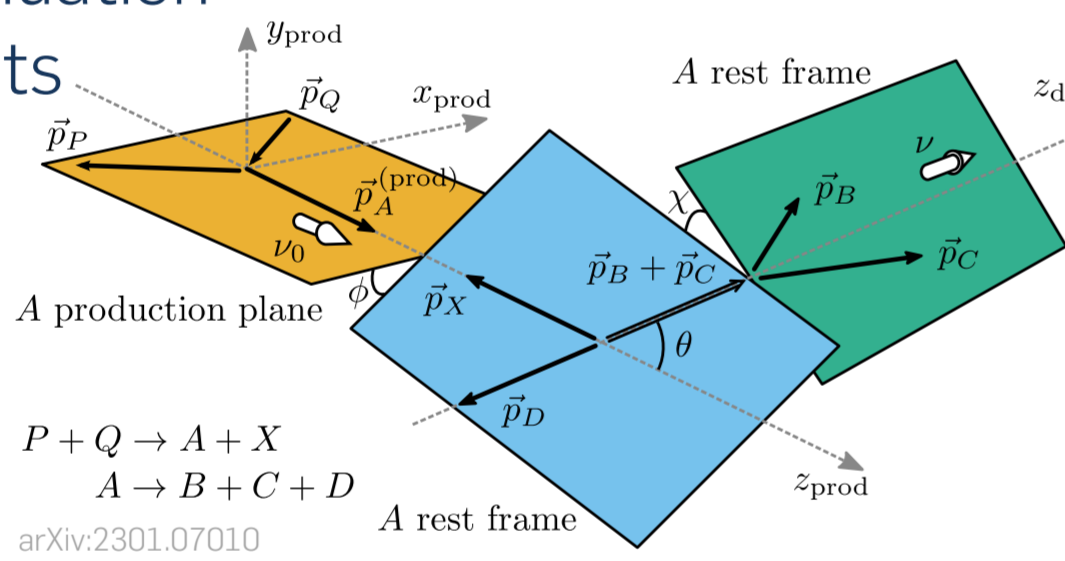
#### AmpForm: Symbolic amplitude models

##### Core responsibilities:

- Library of spin formalisms and dynamics
- Formulate amplitude model for QRules output

##### Implemented formalisms:

- Helicity formalism for multibody decays
- Transformation to canonical basis
- Dynamics parametrisations, such as analytic continuation
- Spin alignments



```
n = sp.Symbol("n_R")
matrix = RelativisticKMatrix.formulate(
    n_channels=1,
    n_poles=n,
)
matrix[0, 0]
```

$$\rho(s) \sum_{R=1}^{n_R} \frac{\Gamma(s) \gamma_{R,0}^2 m_R}{-s + m_R^2} - i\rho(s) \sum_{R=1}^{n_R} \frac{\Gamma(s) \gamma_{R,0}^2 m_R}{-s + m_R^2} + 1$$

```
matrix = NonRelativisticKMatrix.formulate(
    n_channels=2,
    n_poles=1,
).doit()
matrix[0, 0].simplify()
```

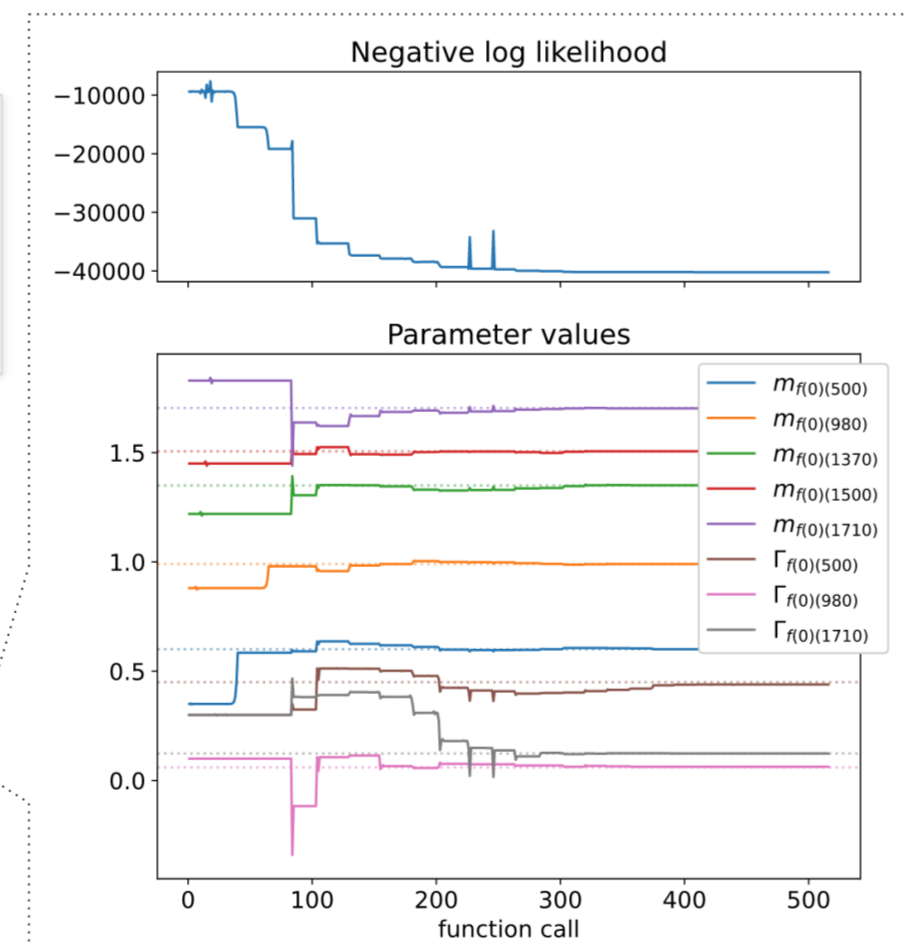
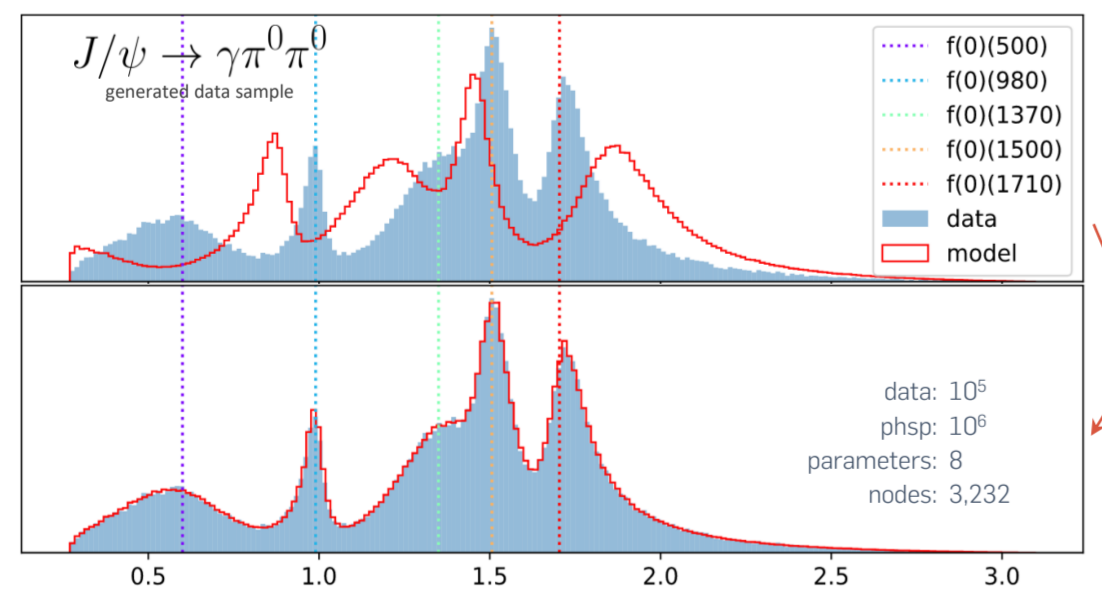
$$\frac{\Gamma_{1,0} \gamma_{1,0}^2 m_1}{s + i\Gamma_{1,0} \gamma_{1,0}^2 m_1 + i\Gamma_{1,1} \gamma_{1,1}^2 m_1 - m_1^2}$$

ampform.rtdf.io

#### TensorWaves: Fit data with multiple computational back-ends

- Mathematical expressions → Computational back-end
- Generation of amplitude-based Monte Carlo samples
- Fits with TensorFlow, NumPy, JAX, ...
- Different optimizers: Minuit2, SciPy, ...

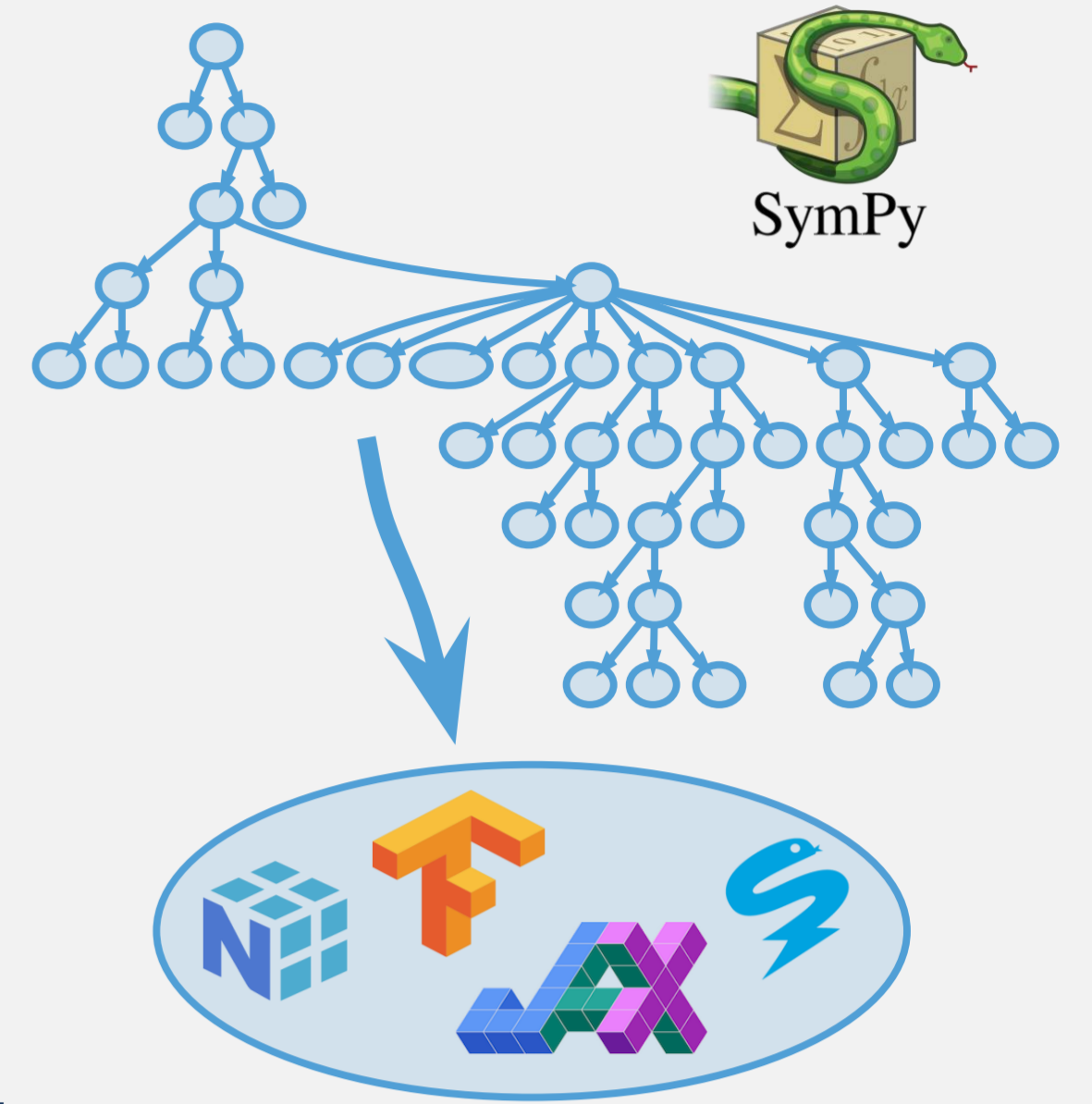
```
function = create_parametrized_function(expression, parameters, backend="jax")
estimator = UnbinnedNLL(function, data, phsp, backend="jax")
minuit2 = Minuit2(Callback=CSVSummary("fit_traceback.csv"))
fit_result = minuit2.optimize(estimator, initial_parameters)
```



tensorwaves.rtdf.io

#### Core idea:

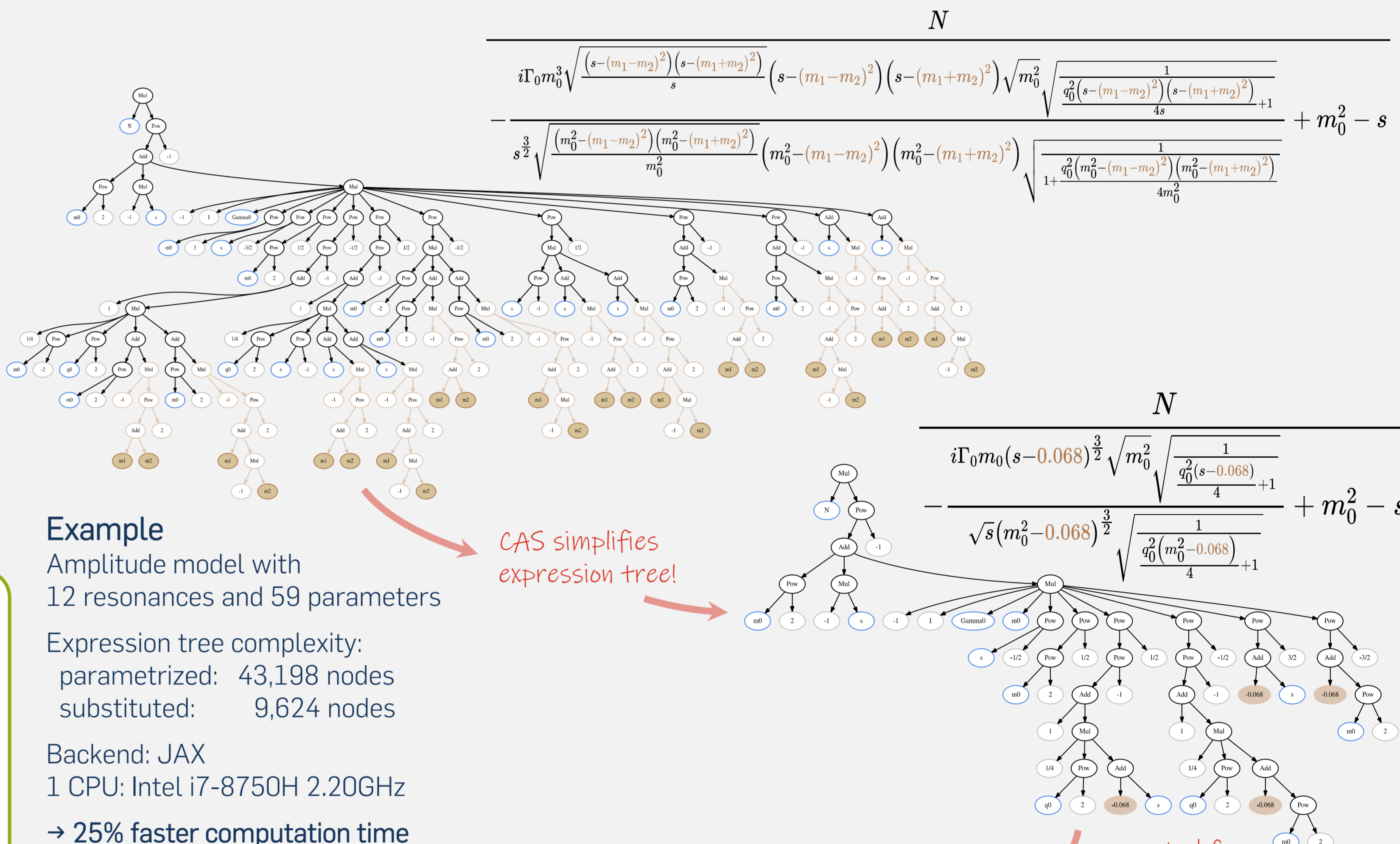
1. Formulate amplitude model as formulas → Computer Algebra System
2. Usage of fast numerical back-ends



#### Main benefits:

- Physics separated from number crunching
- Switch back-end without changing analysis code
- Symbolic expressions result in self-documenting workflow
- Computations outsourced to fast, optimized back-ends with a large user-base
- Out-of-the-box GPU and multithreading support

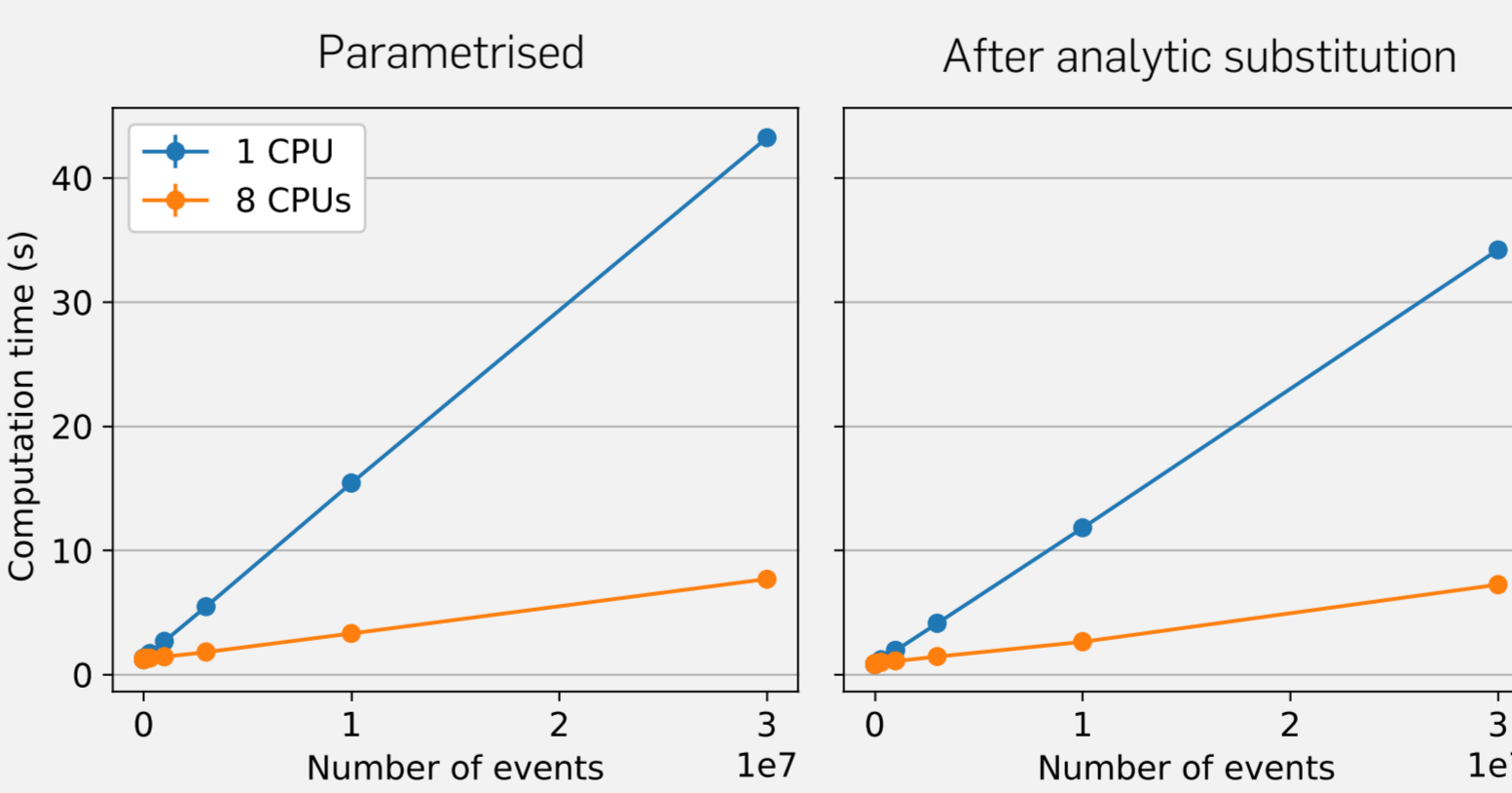
#### CAS simplifications result in performance boosts



Example  
Amplitude model with 12 resonances and 59 parameters

Expression tree complexity:  
parametrized: 43,198 nodes  
substituted: 9,624 nodes

Backend: JAX  
1 CPU: Intel i7-8750H 2.20GHz  
→ 25% faster computation time



```
@jax.jit
def _LambdifyGenerated(Gamma0, N, m0, s):
    return N / (
        -1j
        * Gamma0
        * m0
        * ((1 / 4) * m0**2 + 0.9831)
        * (s - 0.0676) ** (3 / 2)
        * sqrt(m0**2)
        / (sqrt(s) * (m0**2 - 0.0676) ** (3 / 2) * ((1 / 4) * s + 0.9831))
        + m0**2
        - s
    )
```

#### Self-documenting workflow

Search the docs... [Launch interactive examples](#)

**Pole parametrization**  
After all these matrix definitions, the final challenge is to choose a correct parametrization for the elements of  $K$  and  $P$  that accurately describes the resonances we observe. There are several choices, but a common one is the following summation over the poles  $R_i$ :

$$K_{ij} = \sum_R \frac{g_{Ri} g_{Rj}}{m_R^2 - s} + c_{ij}$$

$$K_{ij} = \sum_R \frac{g_{Ri}(s) g_{Rj}(s)}{(m_R^2 - s) \sqrt{\rho(p_i)}} + \tilde{c}_{ij}$$

with  $c_{ij}, \tilde{c}_{ij}$  some optional background characterization and  $g_{Ri}$  the residue functions. The residue functions are often further expressed as:

$$g_{Ri} = \gamma_{Ri} \sqrt{m_{Ri} \Gamma_{Ri}}$$

$$g_{Ri}(s) = \gamma_{Ri} \sqrt{m_{Ri} \Gamma_{Ri}(s)}$$

with  $\gamma_{Ri}$  some real constants and  $\Gamma_{Ri}$  the partial width of each pole, in the Breit-Wigner form, the latter often is implemented by an energy-dependent form:

• Thoroughly integrated with codebase  
• Continuously tested links and code

Powered by symbolic expressions  
• Explanations of implemented physics  
• Kind of an interactive book in the browser (see Executable Book Project)

#### Spin-off project: polarimetry vector field of $\Lambda_c^+ \rightarrow pK^-\pi^+$

Polarimeter vector field  $\vec{\alpha}$  for:  
• determination of polarisation  
• improvement of the sensitivity of amplitude models

$$|\mathcal{M}(\phi, \theta, \chi, \kappa)|^2 = I_0(\kappa) \left( 1 + \sum_{i,j=1}^3 P_i R_{ij}(\phi, \theta, \chi) \alpha_j(\kappa) \right)$$

$$\vec{\alpha}(\kappa) = \sum_{\nu, \nu', \lambda} A_{\nu, \nu', \lambda}^* \bar{\sigma}_{\nu, \nu'} A_{\nu, \nu', \lambda} / I_0(\kappa)$$

CompPWA offers the flexibility to compute vector fields and easily test several parametrisations

lc2pkpi-polarimetry.docs.cern.ch  
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