

# NaCl : what is it?

**Mahmoud Osman**

2nd year PhD student

Laboratoire de Physique Nucléaire et des Hautes Énergies (LPNHE)

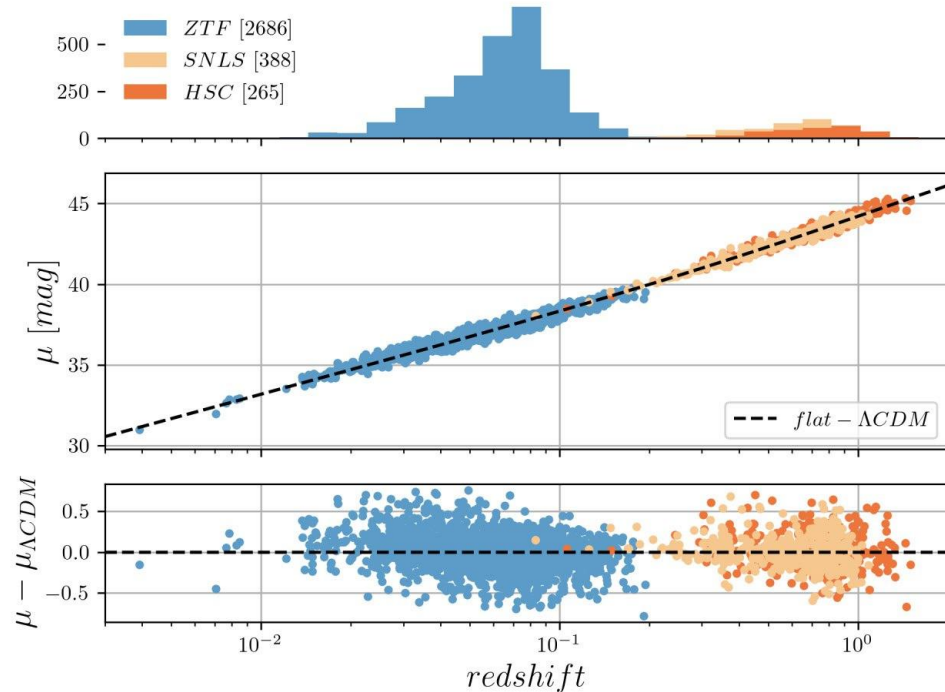
Supervised by **Nicolas Regnault** & **Pauline Zarrouk**

11/12/2024

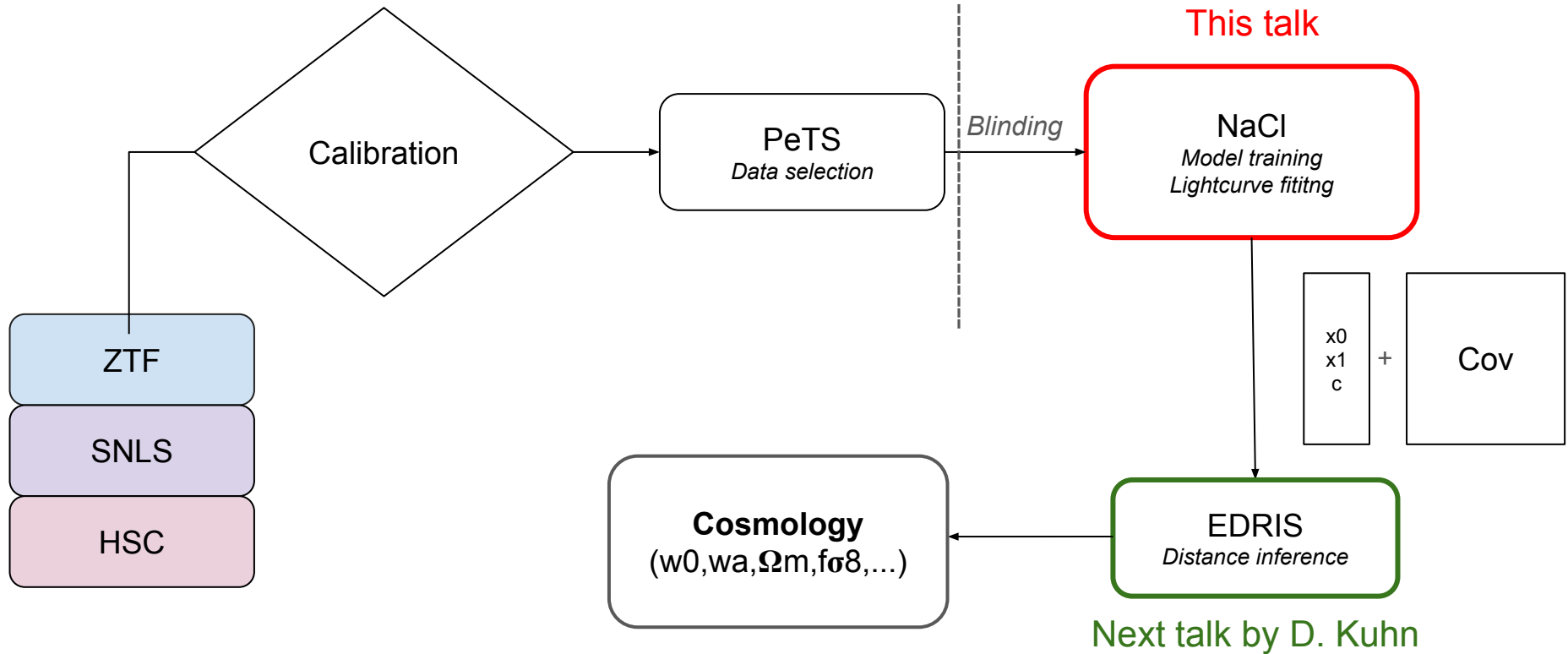
# LEMAÎTRE (Latest Extensive Mapping of Acceleration with Independent Troves of Redshifted Explosions)



- The LEMAÎTRE project aims at constructing a new and independent Hubble-Lemaître diagram
- It contains its own **new analysis pipeline** from extracting SN Ia lightcurves from pixels to the cosmological analysis
- This allows us to tackle multiple known issues that were previously complicated to address

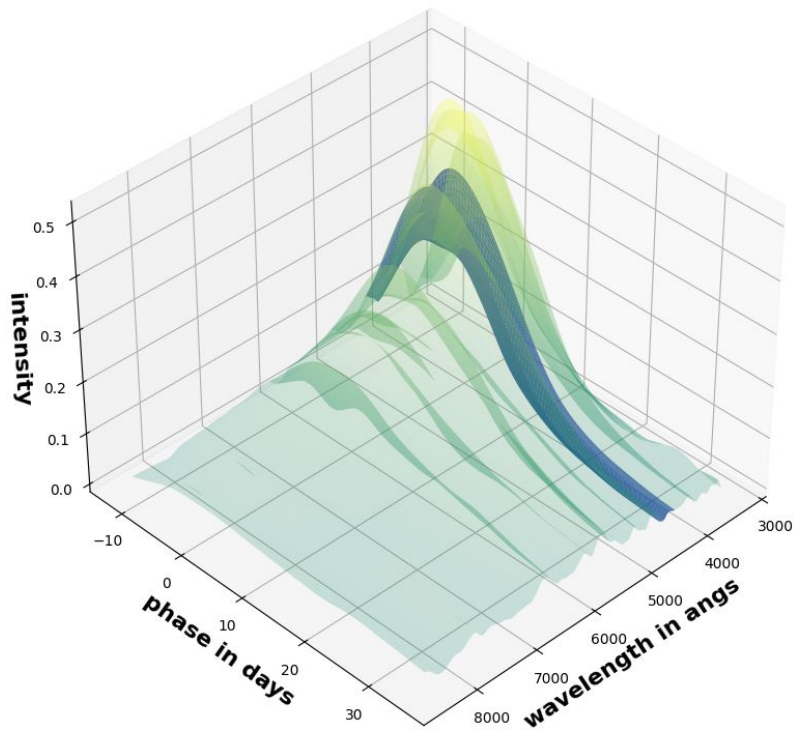
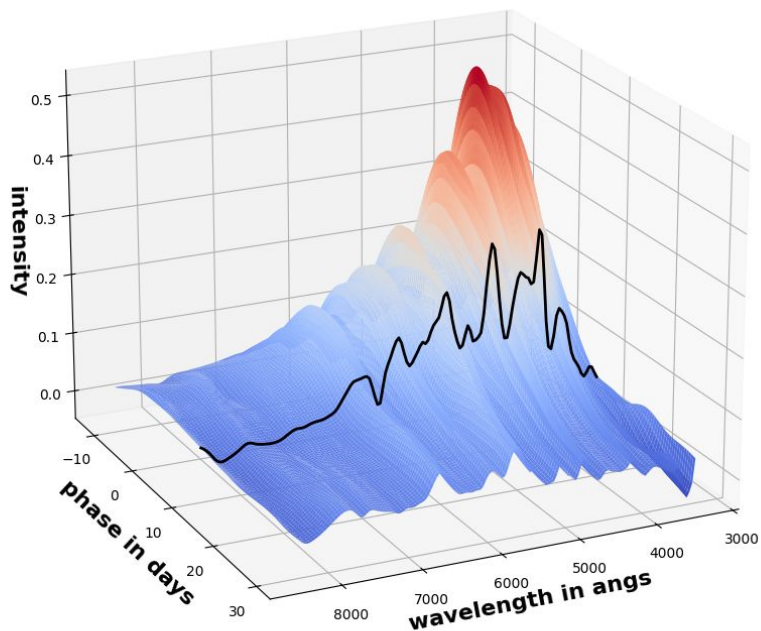


# LEMAÎTRE pipeline



# SN Ia spectrophotometric modelling

- We interpolate the SN Ia flux from a spectrophotometric model : slice in phase



# SN Ia spectrophotometric modelling

- To describe SNe diversity we use spectrophotometric models like SALT2

$$S(\lambda, p) = X_0 \times [M_0(\lambda, p) + X_1 \times M_1(\lambda, p)] 10^{0.4cCL(\lambda)}$$

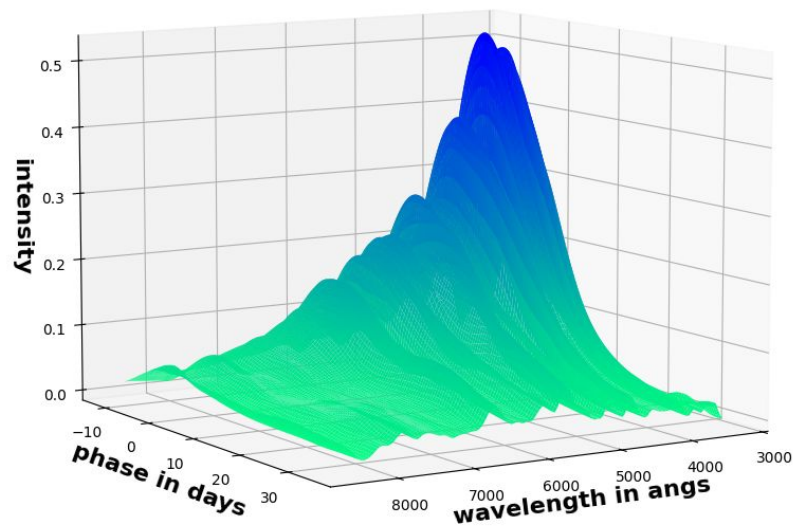
Model

Amplitude

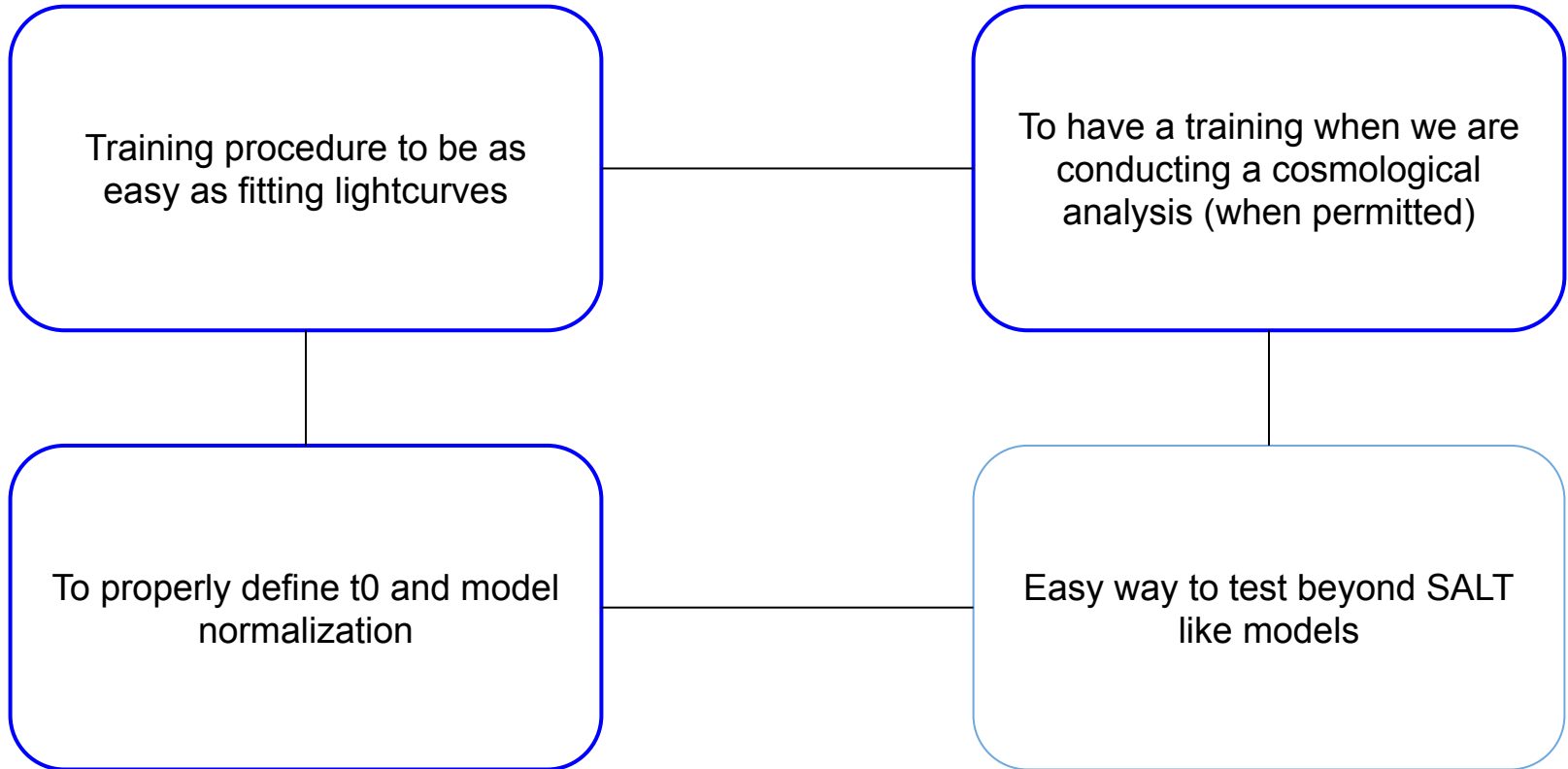
Stretch

Color

SALT2.4

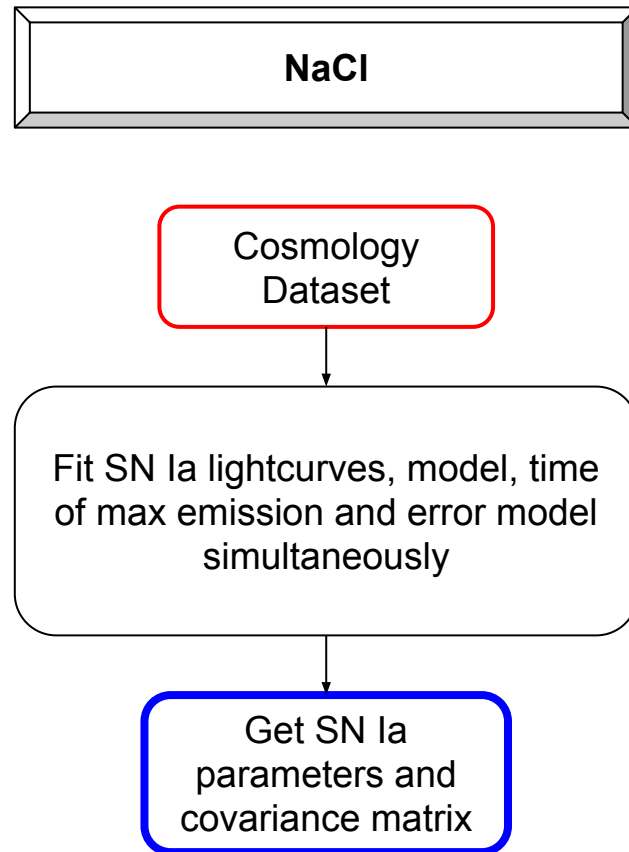


# The idea behind NaCl



# NaCl : a new framework for training spectrophotometric models

- The training procedure is **simplified into one log likelihood minimisation** which encapsulates the **propagation of measurement, model and calibration uncertainties**
- NaCl is **user friendly, fast** for training models and can be easily **reparameterized** to train more sophisticated models



# **NaCl** : Requirements for such a framework

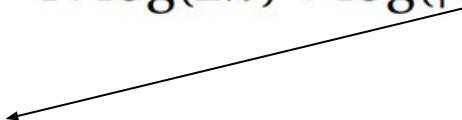
1. Error model fitted during training
2. Fitting  $t_0$  and having constraints on model
3. Regularize model
4. Include calibration during training
5. Include color scatter



# NaCl : Requirements for such a framework

**Error model (error snake)** : To capture the variabilities from one SN to another we fit an error snake model

$$-2\log(\mathcal{L}(\beta)) = N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R}$$


$$\mathbf{V}^{-1} = (\mathbf{V}_{\text{meas}} + \mathbf{V}_{\text{model}})^{-1}$$

Where

$$\mathbf{V}_{\text{model}} = \begin{cases} (\sigma(\lambda, p) \times \gamma_{\text{SN}} \times \text{flux})^2 \\ \dots \\ (\gamma \times \text{flux})^2 \end{cases}$$

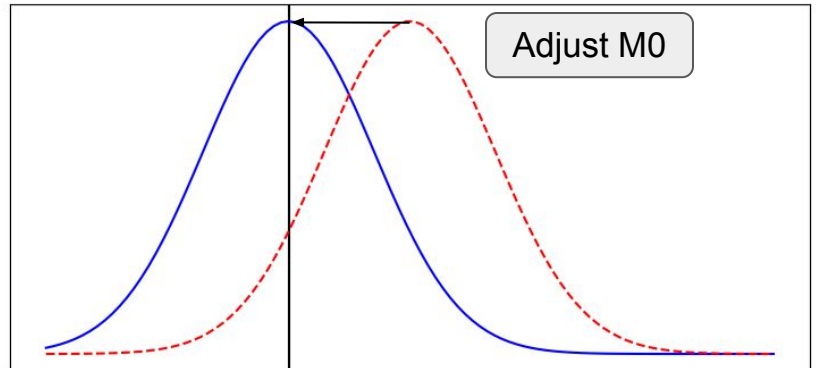
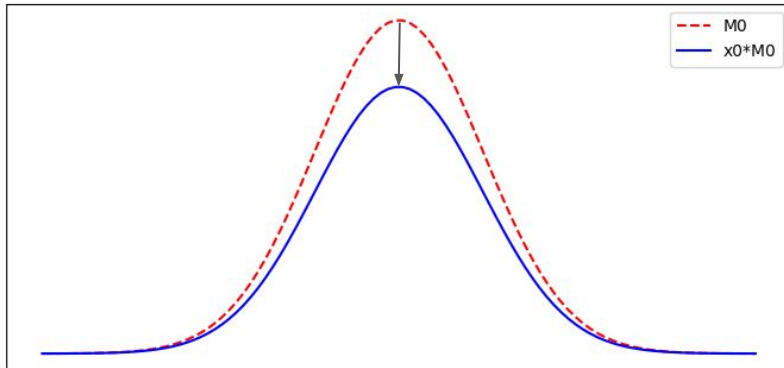
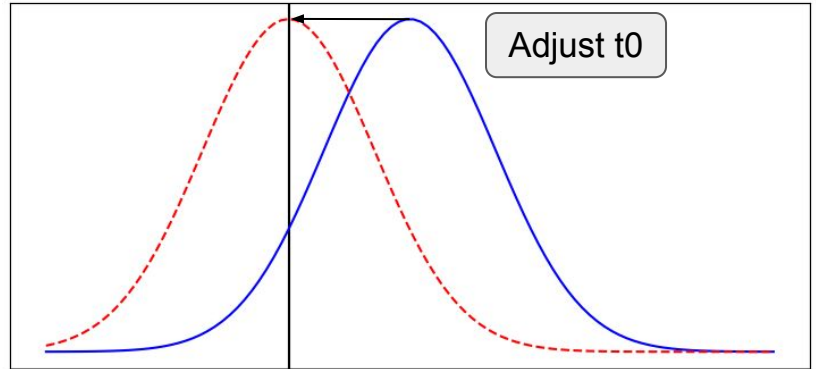
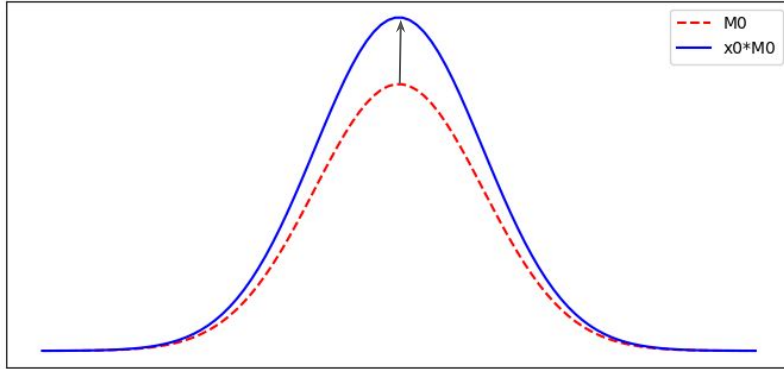
## NaCl : Requirements for such a framework

**Model constraints and tmax fit** : The model contains degeneracies between the model parameters and SN parameters when it comes to the definition of **color**, **normalization** and **time of maximum emission in the B band**

$$\begin{aligned} -2\log(\mathcal{L}(\beta)) = & N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ & + \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta) \end{aligned}$$

# NaCl : Requirements for such a framework

## Model constraints and tmax fit : examples



# NaCl : Requirements for such a framework

**Model constraints and tmax fit** : The constraints in NaCl are different to what we have in SALT

$$\int M_0(\lambda, p = 0) T_B(\lambda) \frac{\lambda}{hc} d\lambda = 1$$

$$\int \frac{\partial M_0(\lambda, p = 0)}{\partial t} \Big|_{t=t_B} = 0$$

$$\int M_1(\lambda, p = 0) T_B(\lambda) \frac{\lambda}{hc} d\lambda = 0$$

$$\int \frac{\partial M_1(\lambda, p = 0)}{\partial t} \Big|_{t=t_B} = 0$$

The constraints in NaCl are evolving to be fully independent of the dataset

$$\langle (X_1 - \langle X_1 \rangle)^2 \rangle = 1 \rightarrow \int M_1(\lambda, p = 15) T_B(\lambda) \frac{\lambda}{hc} d\lambda = 1$$

## NaCl : Requirements for such a framework

**Model regularization** : Parts of the model won't be constrained by any data so we add a penalty term, model regularization, to adjust the model in those areas

$$\begin{aligned} -2\log(\mathcal{L}(\beta)) &= N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ &\quad + \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta) + \boxed{\mu_{reg} \beta^T \mathbf{P} \beta} \end{aligned}$$

# NaCl : Requirements for such a framework

**Calibration uncertainties** : We take into account the calibration uncertainties by fitting **a parameter per band** which are held by **prior** calculated with the **predetermined covariance matrix of the filters**

$$\begin{aligned} -2\log(\mathcal{L}(\beta)) &= N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ &+ \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta) + \mu_{reg} \beta^T \mathbf{P} \beta \\ &+ \eta^T \mathbf{V}_\eta^{-1} \eta \end{aligned}$$

# NaCl : Requirements for such a framework

Color scatter : We take into account the intrinsic variation of the color of each SN

Currently we fit a **parameter per lightcurve** held by **prior** that is fitted during the training

This current implementation will change due to the fact that we have SNe measured in **only 2 bands** with ZTF

$$\begin{aligned} -2\log(\mathcal{L}(\beta)) &= N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ &+ \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta) + \mu_{reg} \beta^T \mathbf{P} \beta \\ &+ \eta^T \mathbf{V}_\eta^{-1} \eta + \log(|\mathbf{V}_\kappa(\sigma_\kappa)|) + \kappa^T \mathbf{V}_\kappa^{-1}(\sigma_\kappa) \kappa \end{aligned}$$

# Nacl : Likelihood

Model constraints

Classic likelihood with  
 $\mathbf{V}^{-1} = (\mathbf{V}_{\text{meas}} + \mathbf{V}_{\text{model}})^{-1}$

$$\begin{aligned} -2\log(\mathcal{L}(\beta)) = & N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ & + \mu_{\text{cons}} \mathbf{C}(\beta)^T \mathbf{C}(\beta) + \mu_{\text{reg}} \beta^T \mathbf{P} \beta \\ & + \eta^T \mathbf{V}_{\eta}^{-1} \eta + \log(|\mathbf{V}_{\kappa}(\sigma_{\kappa})|) + \kappa^T \mathbf{V}_{\kappa}^{-1}(\sigma_{\kappa}) \kappa \end{aligned}$$

Calibration uncertainty

Color scatter

Model regularization



## Nacl : Likelihood

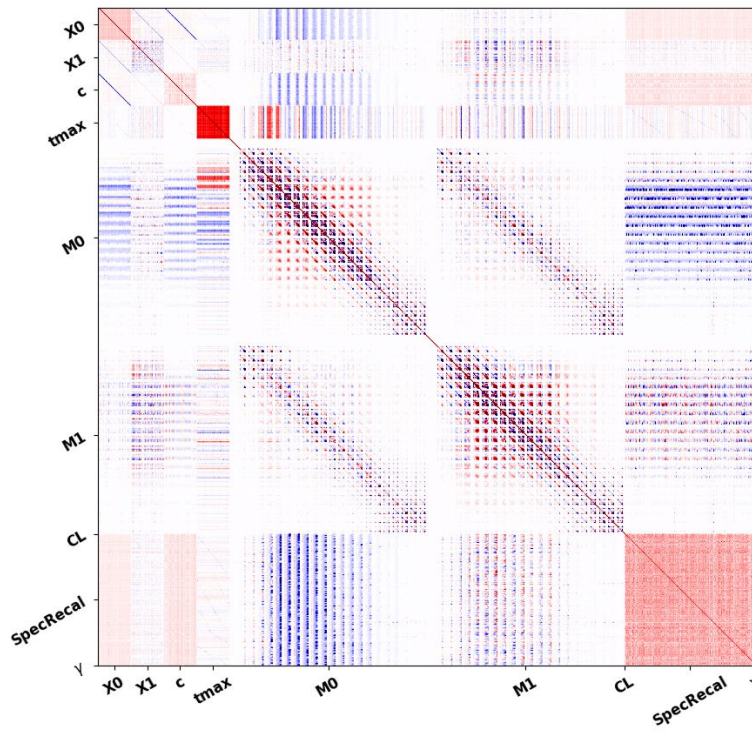
$$\begin{aligned} -2\log(\mathcal{L}(\beta)) &= N\log(2\pi) + \log(|\mathbf{V}|) + \mathbf{R}^T \mathbf{V}^{-1} \mathbf{R} \\ &\quad + \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta) + \mu_{reg} \beta^T \mathbf{P} \beta \\ &\quad + \eta^T \mathbf{V}_\eta^{-1} \eta + \log(|\mathbf{V}_\kappa(\sigma_\kappa)|) + \kappa^T \mathbf{V}_\kappa^{-1}(\sigma_\kappa) \kappa \end{aligned}$$

# NaCl : a new framework for training spectrophotometric models

- After a training NaCl marginalises over the model parameters and **only keeps the SN parameters used in the cosmological analysis** and their **covariance matrix**

One parameter per SN

A few thousand parameter

$$\begin{bmatrix} X_0 \\ X_1 \\ c \\ t_0 \\ M_0 \\ M_1 \\ CL \\ CL \\ \dots \end{bmatrix}$$


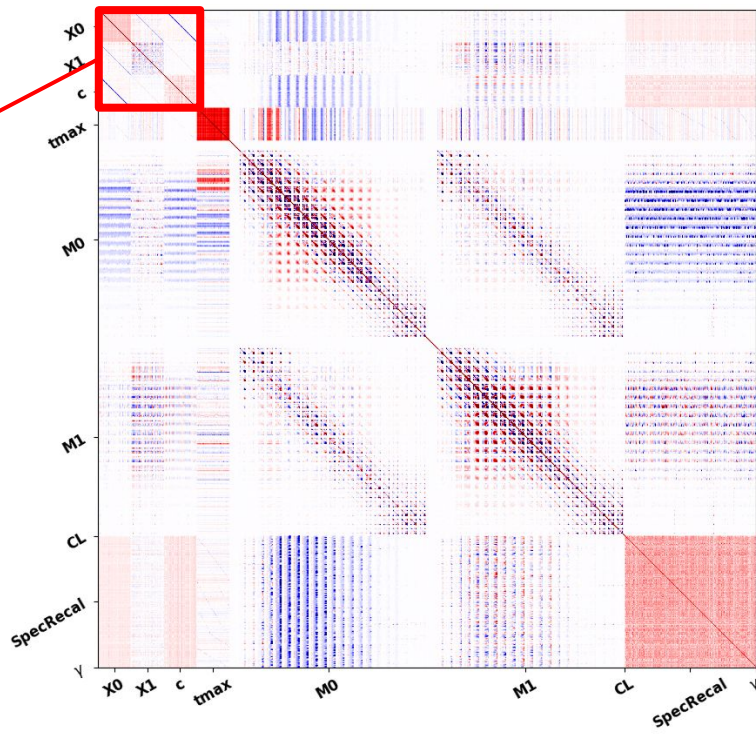
# NaCl : a new framework for training spectrophotometric models

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One parameter per SN

A few thousand parameter

$\begin{bmatrix} X_0 \\ X_1 \\ c \\ t_0 \\ M_0 \\ M_1 \\ CL \\ \dots \end{bmatrix}$



# Important technical notes on NaCl

NaCl contains 2 hyper-parameters managing the constraints and regularization

$$\chi_{reg}^2 = \mu_{reg} \beta^T \mathbf{P} \beta$$

$$\chi_{cons}^2 = \mu_{cons} \mathbf{C}(\beta)^T \mathbf{C}(\beta)$$

- If  **$\mu_{reg}$** 's value is too high the model becomes attenuated and the distances are biased
- If  **$\mu_{cons}$**  is too small the constraints are poorly respected in the training and degeneracies appear

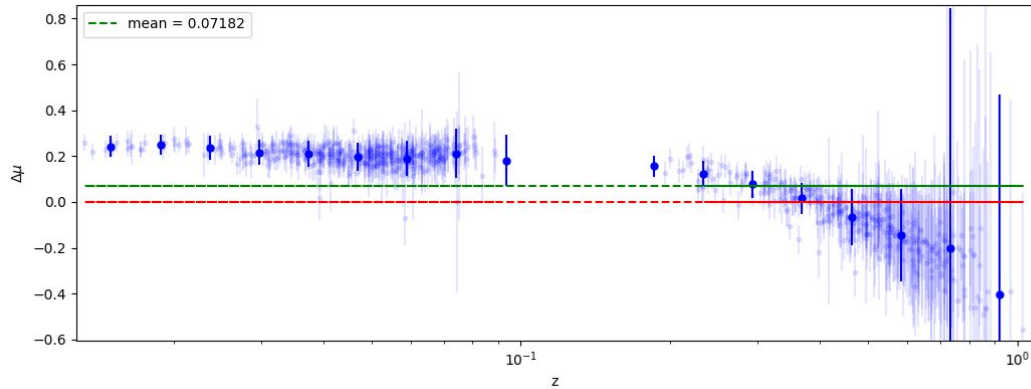
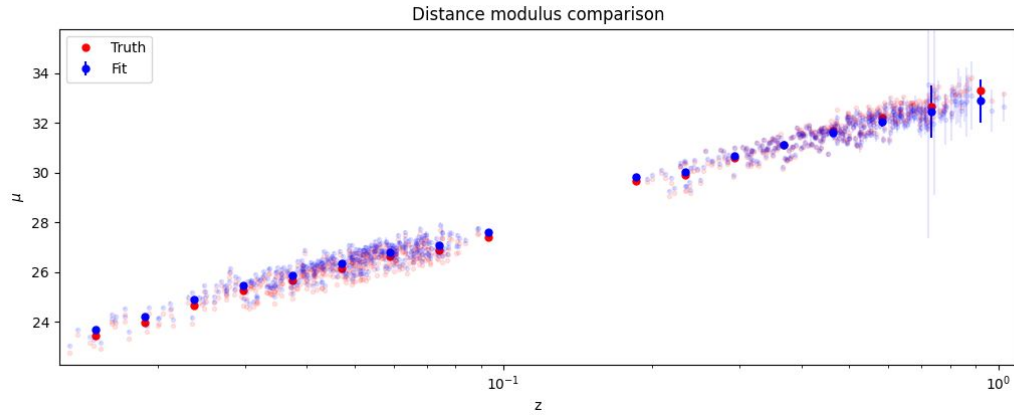
We have a heuristic with simulations that allows us to find the correct values of the hyper-parameters

# Important technical note on NaCl

- The hyperparameters' exact values doesn't need to be known for NaCl to be able to properly train a model
- We noticed issues when we change our hyperparameters by multiple orders of magnitude

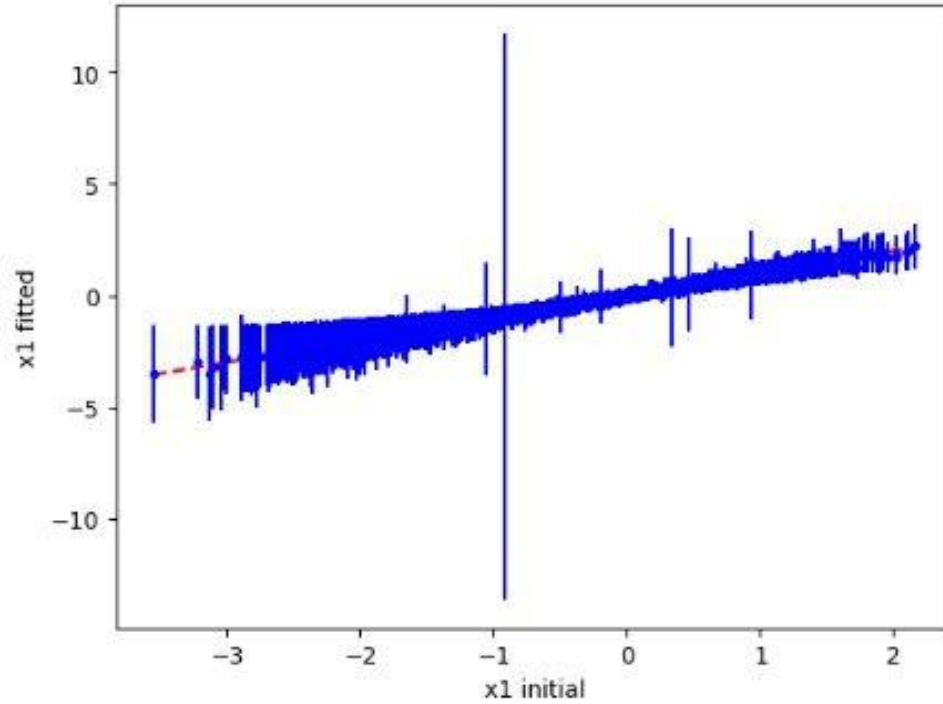
# Examples

$\mu_{\text{reg}}$   
increased by  
 $1e+4$



# Examples

$\mu_{\text{cons}}$   
decreased by  
 $1e+12$

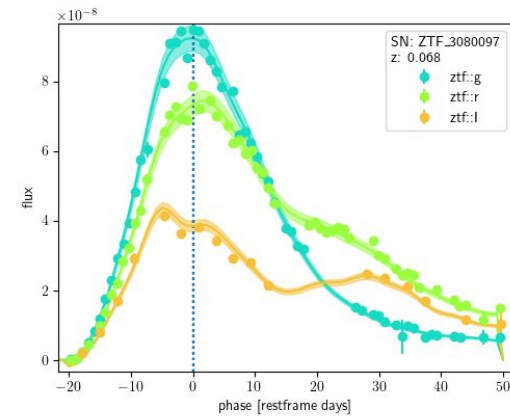
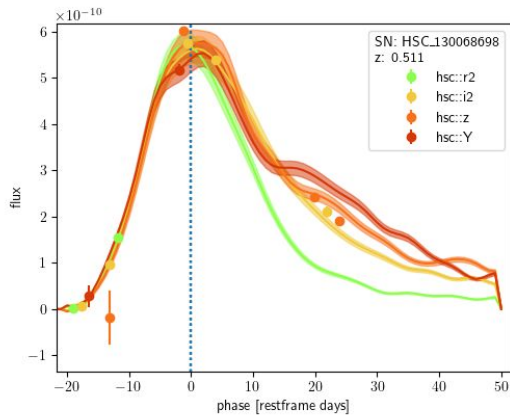
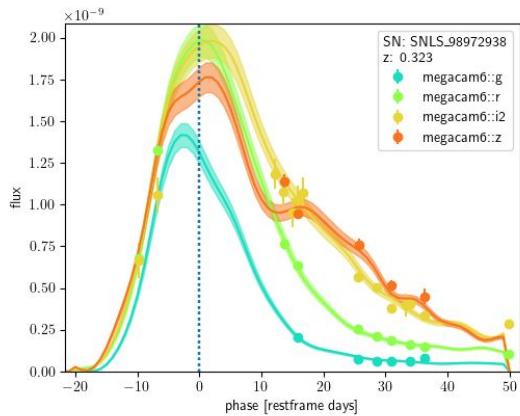
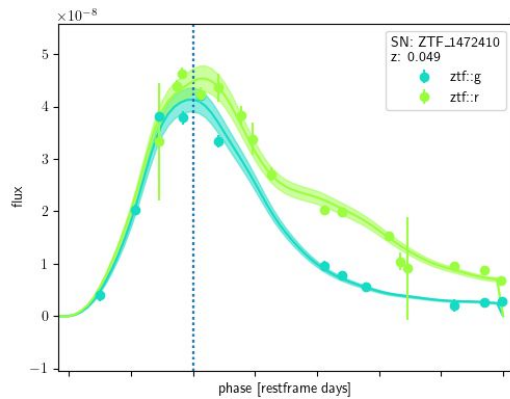
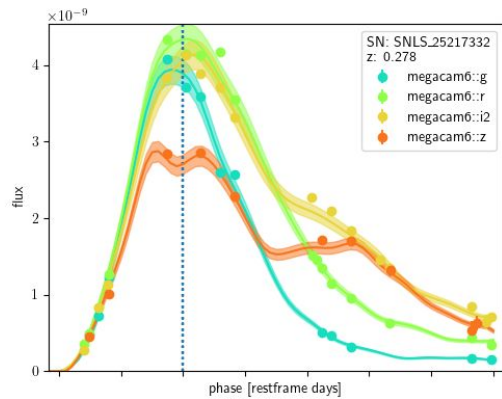
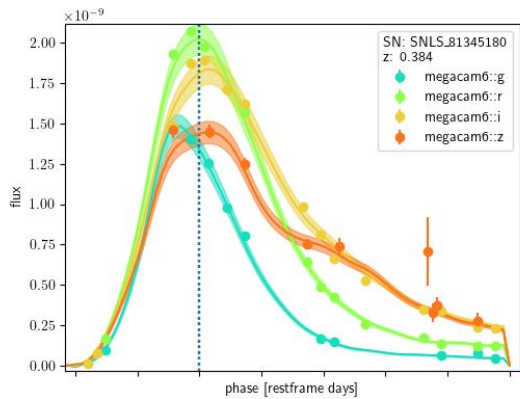


# Current developments in NaCl

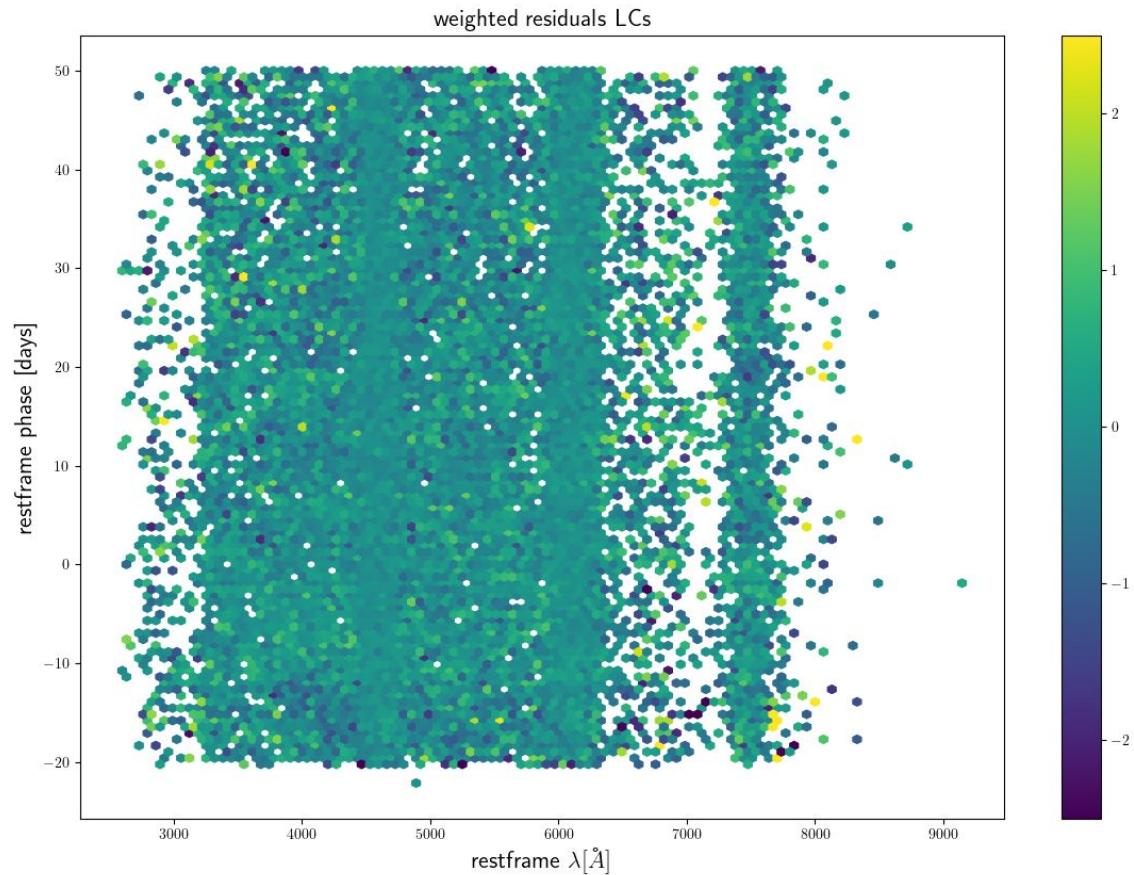
- Color scatter needs to be reworked due to the current method not being able to work for ZTF SNe measured in only 2 bands (DC3)
- Account for the calibration mean wavelength uncertainties (DC3)
- Improving what we call the “training recipe” to build a robust starting point to the training



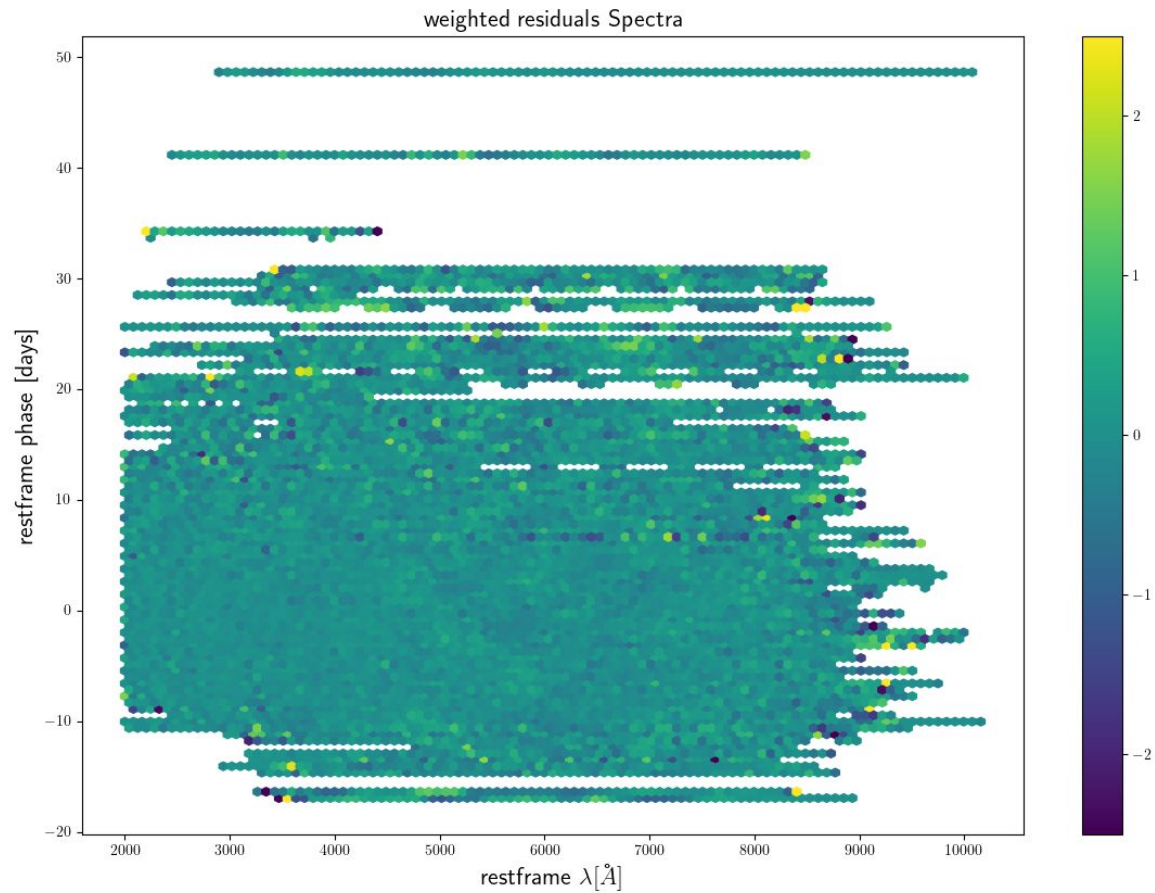
# NaCl output with DC1



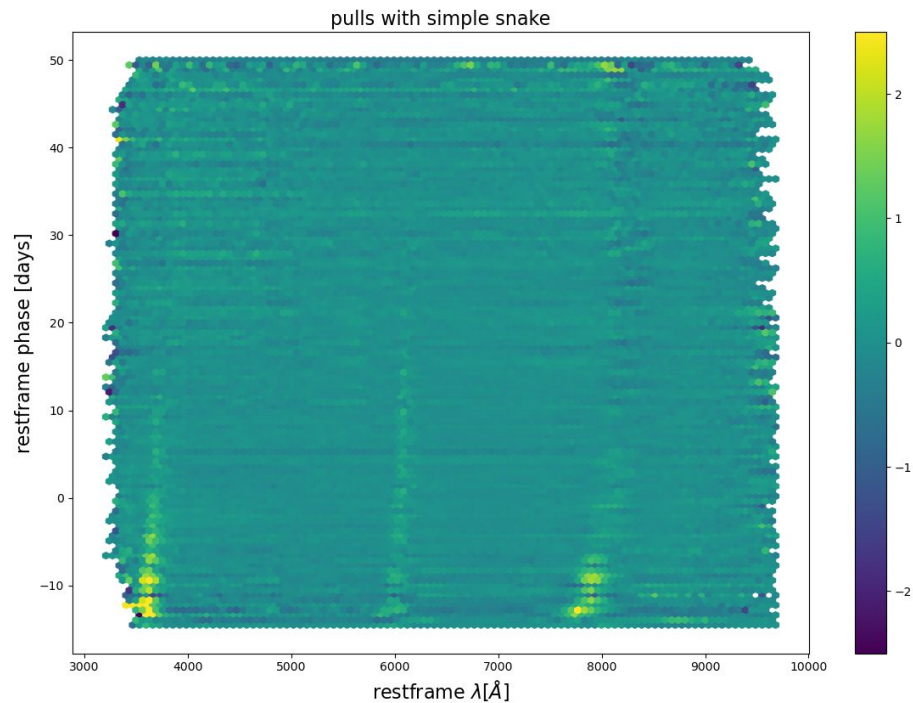
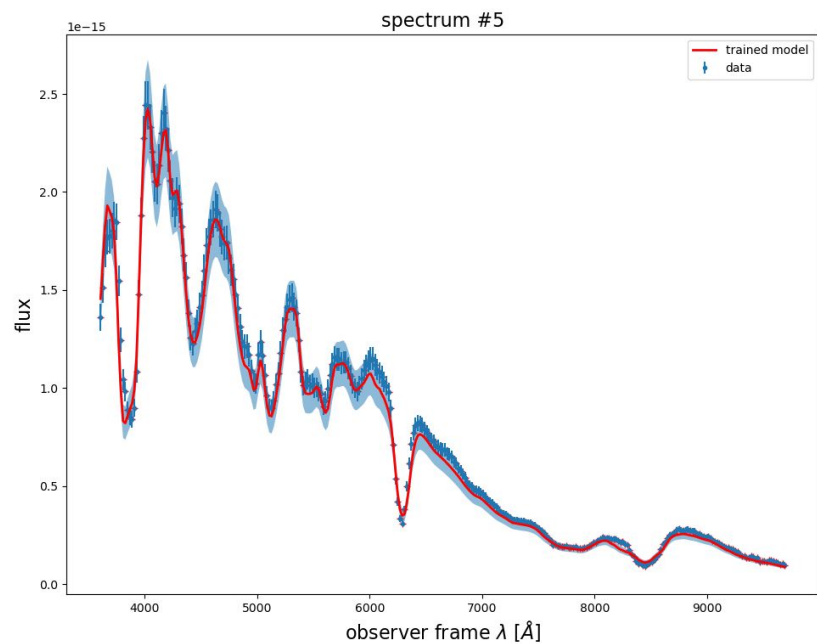
# NaCl output with DC1



# NaCl output with DC1



# NaCl output on SNFactory data



# NaCl

In short, NaCl is **a framework that trains a spectrophotometric model** on a **cosmological dataset** to ensure an accurate **propagation of measurement, model and calibration uncertainties**

We will show NaCl applied on simulations in the next presentation with the DC1 results...

# Important technical notes on NaCl

- NaCl's constraints on the model are different than we have in SALT
- We are working on a set of constraints that do not depend on the dataset, only the model
  
- If we are working with simulations then **we don't** necessarily **expect** to **reconstruct the same SN parameters** but we do expect to obtain the **same distances**