# RTModel: automatic fast real-time modelling of microlensing events



# Issues in microlensing modelling

- Microlensing is a **non-repeatable** phenomenon
- Data quantity and quality cannot be improved if insufficient.





- The **computational time** of a single model point is long.
- The number of modelling parameters is large.





- There is an **extreme variety** of light curve morphologies.
- Chi square is **wildly sensitive** to small variations in the parameters.





 Many discrete and continuous degeneracies exist.

# RTModel

- Since 2011 we have been running our computational platform **RTModel** on our good old 8-core workstation.
- More than 600 events have been modelled in 4 years.



- RTModel automatically responds to anomaly alerts by **ARTEMIS**.
- Data are downloaded and pre-processed.
- Initial conditions are automatically set.
- Downhill fitting is performed and higher order effects are considered.
- Models are automatically displayed on a **public webpage**.
- The total time for a single run is kept within **3** hours.





### **Basic calculation**

- First step: for given lens and source positions, we must compute the gravitational lensing magnification.
- Inverse ray-shooting amounts to shooting back light rays from the observer to the source plane.
- Light rays are counted if they hit the source disk.
- Magnification maps re-usable (save for orbital motion cases). Limb darkening naturally included.



Optimizations are possible.



- We use **contour integration**: boundaries of the images are calculated; area is obtained by Green's theorem.
- Elegant and fast. Limb darkening requires multiple contours.
- Computational time is somewhat less of ms.
- With thousands of points a single lightcurve may exceed one second.

# Initial conditions

- **Grid search** might cover the interesting regions of the parameter space,
- but is always redundant and needs sufficiently small steps.



- **Template matching** (*Mao & Di Stefano 1995*) avoids redundancy and promises to be exhaustive.
- More vulnerable to the presence of local minima within a given class.
- Peaks in the datasets are identified and classified by their prominence.
- The two most prominent peaks are matched to the peaks in the templates.
- If there is only one peak in the data, the anomaly alert time is taken as the second "would-be" peak.



# **Classification of light curves**

- The completeness of the template library is of crucial importance for the effectiveness of this approach.
- We have now published **the first complete catalogue of light curves** in equal-mass binary microlensing

(Liebig, D'Ago, Bozza and Dominik arXiv:1501.02219).



- Every peak in a microlensing light curve can be traced to an interaction of the source with a **caustic**:
- Fold crossing
- Cusp crossing
- Fold approach
- **Cusp approach**

### $[b_{b} a_{b1} a_{t2}] A_{2}$

Light curve morphologies are classified by their specific sequence of peaks.

### The catalogue of light curves





• We have scanned the **parameter space** distinguishing all **regions** corresponding to different morphologies.

- In the equal-mass case, we have identified 73 different morphologies, arising from 232 different regions of the parameter space.
- We can link any observed morphology to the respective regions of the parameter space.
- The classification can be naturally extended to arbitrary mass-ratios.

	Morphology Class	Close	Intermediate	Wide
I	С	outside caustics	outside caustics	outside caustics
		between caustics		between caustics
п	F-F	$[a_{ip1}a_{ip2}], [a_{is1}a_{is2}], [a_{bp1}a_{ip2}], [a_{is1}b_{i}], [a_{bp1}a_{ip1}] \}$	$[a_{i1}a_{i2}], [a_{b1}a_{i2}], [a_{i1}b_{i}], [a_{b1}b_{i}], [b_{b}b_{i}]$	$[a_{b1}b_{r1}], [b_{b1}b_{r1}]$
	CC	$B_{t1}B_{t2}, A_1B_{t1}, A_1C_{tp}, A_1C_{tx}, A_1B_{t2}, C_{tx}B_{t2}$	$B_{t1}B_{t2}, A_1B_{t1}$	$B_{t1}B_{t2}, A_1B_{t1}, B_{b1}D_1, B_{b1}B_{t2}$
	C-C	[A1A2]	[A <sub>1</sub> A <sub>2</sub> ]	-
	C-F	$[a_{ta1}B_{t2}], [A_1a_{tp2}]$	$[a_{e1}B_{e2}], [A_1a_{e2}], [B_{b1}b_e], [A_1a_{e2}]$	$[B_{b1}b_{t1}]$
	ČF-F	$\begin{array}{l} A_1[a_{ip1}a_{ip2}], A_1[a_{bp1}a_{ip2}], [a_{ix1}b_1]B_{l2}, A_1[a_{ix1}b_1], [a_{bp1}a_{ip2}]B_{l2}, \\ [a_{bp1}a_{ip1}]B_{l1}, [a_{a_{1}}a_{a_{2}}]B_{l2}, [a_{a_{1}}a_{ip2}]B_{l2}, [a_{bp1}a_{ip2}]B_{l2}, \\ [a_{bp1}a_{ip1}]C_{ip}, \\ [a_{bp1}a_{ip1}]C_{ip}, \\ [a_{bp1}a_{ip1}]C_{ip}, \\ [a_{bp1}a_{ip1}]C_{ip}, \\ \end{array}$	$\begin{array}{l} A_1[a_{i1}a_{i2}], A_1[a_{b1}a_{i2}], [a_{i1}b_1]B_{i2}, A_1[a_{i1}b_1], [a_{b1}a_{i2}]B_{i2}, [a_{b1}b_1]B_{i2}, \\ B_{b1}[a_{b1}b_1], [a_{b1}a_{i1}]B_{i1}, [a_{i1}a_{i2}]B_{i2}, [a_{b1}b_1]B_{i2}, B_{b1}[b_2b_1] \end{array}$	$ \begin{array}{l} [a_{i1}b_{i1}]B_{i2},A_1[a_{i1}b_{i1}],B_{k1}[a_{k1}b_{i1}],[a_{k1}a_{i1}]B_{i1},[a_{k1}b_{i1}]B_{i2},\\ B_{k1}[b_{k1}b_{i1}],A_1[b_{k1}b_{i1}],[a_{k1}b_{k1}]B_{i2} \end{array} $
ш	F-F-F	$[a_{tx1}b_t a_{tx2}], [a_{bp1}a_{tp1}a_{tp2}]$	$[a_{l1}b_{l}a_{l2}], [a_{b1}a_{l1}a_{l2}], [a_{b1}a_{l1}b_{l}], [b_{b}a_{b1}b_{l}], [a_{b1}a_{l2}b_{l}], [a_{b1}b_{l}a_{l2}]$	$[a_{b1}a_{c1}b_{c1}], [a_{b1}b_{b1}b_{c1}]$
	CF-C	$[N_1a_{ip2}]N_2, C_{bp}[abp1C_{ip}]$	$[A_1a_{t2} A_2, A_1[a_{t1}B_{t2}], [A_1b_t]B_{t2}$	$[A_1b_{t1}]B_{t2}$
	CCC	$B_{b1}A_1B_{c1}, A_1C_{tx}B_{c2}$	$B_{b1}A_1B_{c1}$	$B_{b1}A_1B_{c1}, B_{b1}D_1B_{c2}, A_1B_{c1}B_{c2}, B_{b1}A_1B_{c2}$
	C-F-F	-	$[A_1b_1a_{i2}]$	-
	C C-F	$A_1[C_{ts}b_t]$	-	-
	CF-FC	$\begin{array}{l} A_1[a_{tp1}a_{tp2}]A_2,A_1[a_{bp1}a_{tp2}]A_2,A_1[a_{ts1}a_{ts2}]B_{t2},B_{b1}[a_{bp1}a_{tp2}]B_{t2},\\ B_{b1}[a_{bp1}b_{tp1}]B_{t1},C_{bp}[a_{bp1}a_{tp2}]C_{tp} \end{array}$	$\begin{array}{l} A_1[a_{t1}a_{t2}]A_2, A_1[a_{b1}a_{t2}]A_2, A_1[a_{t1}b_1]B_{t2}, A_1[a_{t1}a_{t2}]B_{t2}, \\ B_{b1}[a_{b1}a_{t2}]B_{t2}, B_{b1}[a_{b1}a_{t1}]B_{t1}, A_1[a_{b1}b_1]B_{t2}, B_{b1}[a_{b1}b_1]B_{t2} \end{array}$	$A_1[a_{t1}b_{t1}]B_{t2}, B_{b1}[b_{b1}b_{t1}]B_{t2}, B_{b1}[a_{b1}a_{t1}]B_{t1}, A_1[a_{b1}b_{t1}]B_{t2}$
	F-FF-F	$[a_{t+1}b_t][b_ta_{t+2}], [a_{bp},a_{tp1}][a_{tp1}a_{tp2}], [a_{bp1},a_{tp1}][a_{t+1}b_t]$	$[a_{t1}b_t   [b_ta_{t2}], [a_{b1}a_{t1}   [a_{t1}a_{t2}], [a_{b1}a_{t1}]   [a_{t1}b_{t}], [b_{b}a_{b1}] ] [a_{b1}b_{t}], [a_{b1}a_{t2}] [a_{t2}b_{t}], [a_{b1}b_{t}] [b_{t}a_{t2}]$	$[a_{t1}b_{t1}][b_{t2}a_{t2}], [a_{b1}a_{t1}][a_{t1}b_{t1}], [a_{b1}b_{b1}][b_{b1}b_{t1}], [a_{b1}b_{b1}][b_{t2}a_{t2}]$
	CF-F-F	$[a_{bp1}a_{ip1}a_{ip2}]B_{i2},A_1[a_{ix1}a_{ix2}b_l]$	$[a_{b1}a_{t1}a_{t2}]B_{t2}, [a_{b1}a_{t1}b_{t}]B_{t2}, B_{b1}[a_{b1}a_{t1}b_{t}], A_{1}[a_{t1}b_{t}a_{t2}], A_{1}[a_{b1}b_{t}a_{t2}], B_{b1}[a_{b1}a_{t2}b_{t}]$	$B_{b1}[a_{b1}a_{c1}b_{11}], [a_{b1}b_{b1}b_{c1}]B_{c2}$
IV	F.F.F.F		$[b_b a_{b1} a_{t1} b_t], [a_{b1} a_{t1} a_{t2} b_t], [b_b a_{b1} a_{t2} b_t]$	-
	C-FF-F	-	$[a_{b1}a_{t1}][a_{t1}B_{t2}], [A_1b_t][b_ta_{t2}]$	$[A_1b_{i1}][b_{i2}a_{i2}]$
	F-FČČ	$[a_{bp1}a_{cp1}]C_{ip}B_{c2}, [a_{bp1}a_{cp1}]C_{ix}B_{c2}, A_1C_{tx}[a_{cx1}b_t], A_1C_{tx}[a_{cx2}b_t], [a_{bp1}a_{cp1}]C_{ix}B_{c1}$	-	[a <sub>b1</sub> b <sub>b1</sub> ]D <sub>1</sub> B <sub>c2</sub>
	C-C-C-C	-	-	$[A_1D_1][D_2A_2]$
	F-F C-F	$[a_{bp1}a_{ip1}][C_{isbi}]$	-	-
	C-FCC	-	-	$[A_1b_{c1}]D_1B_{c2}$
	CFCC	$C_{bp}[a_{bp1}C_{lp}]B_{l2}$	-	-

### Fitting

- The nightmare of modellers is getting stuck in a local minimum.
- Local minima may exist within each region of the parameter space corresponding to a specific morphology class.
- But the presence of **gaps** in the data may copiously generate **see-saw patterns** in the chi square.



# Fitting

- Markov chains have a finite probability (depending on the temperature) to jump out of a local minimum.
- However, they require the calculation of a large number of models from any given initial condition.
- We use a Levenberg-Marquardt algorithm (interpolating between Newton's and steepest descent).
- In order to jump out of local minima, we fill the minima with penalty functions and let the fit roll to the next minimum.







### Higher-order effects

- We refine the best static solutions by including annual parallax and orbital motion.
- For **parallax** we start from  $\pi_{\perp} = \pi_{\mu} = 0$ , which is fine for not too large effects.
- For orbital motion we consider **circular orbits** with arbitrary inclination, parameterized by  $(ds/dt)_{t0}$ ,  $(d\alpha/dt)_{t0}$  and  $(\omega_z)_{t0}$  starting from zero velocities.
- For comparison and completeness, we also calculate the following models:

PSPL

- PSPL with parallax
- Finite source single lens
- Finite source single lens with parallax
- Binary source
- Binary source with parallax
- Binary source with parallax and orbital motion.

### Publication of the results

A webpage at Salerno University is automatically updated with automatically generated plots

http://www.fisica.unisa.it/GravitationAstrophysics/RTModel/2014/RTModel.htm



### Events modelled in 2014

Under investigation



About RTModel

#### Planetary





















### Event webpage

**RTModel** Real-Time Microlensing Modelling by Valerio Bozza

Gravitational Physics and Astrophysics

RTModel index

### OB140124 Planetary with parallax

Binary lens models						
	ł	$\frac{Model \ L1}{Model \ L1}  \chi^2 = 24940.6  g_{OGLE} = 1.00652 \pm 0.412237$				
		s=0.904511±0.0296468 q=0.000750416±0.000129067 u <sub>0</sub> =-0.250688±0.0589067 θ=1.37125±0.0371881 ρ-=0.0028	8639±0.00637904 t <sub>E</sub> =110.8±19.9386 t <sub>0</sub> =6836.09±0.939892			
	1.	Model L2 x <sup>2</sup> =31568.2 g <sub>OGLE</sub> =0.273378±0.0968608				
		s=0.848728±0.0127707 q=0.000784017±0.00010802 u <sub>0</sub> =-0.358492±0.0252235 θ=1.34231±0.0148411 ρ-=0.00774	437±0.00423333 t <sub>E</sub> =84.7621±3.96241 t <sub>0</sub> =6834.65±0.253834			
	1	Model L3 x <sup>2</sup> =42226.3 g <sub>OGLE</sub> =-0.766218±0.149419				
	-	s=0.621889±0.0161537 q=0.000768055±0.000107619 u <sub>0</sub> =-1.00737±0.0554097 θ=1.44582±0.0077895 p+=0.01048	57±0.0187656 t <sub>E</sub> =44.1633±2.46143 t <sub>0</sub> =6836.07±0.137982			
	1	Model L4 x <sup>2</sup> =50628. g <sub>OGLE</sub> =-0.925977±0.208842				
	ł	s=0.476124±0.0187407 q=0.000687306±0.000126724 u <sub>0</sub> =-1.64594±0.0901404 θ=1.47302±0.0227046 p=0.00001	02768±0.00172343 t <sub>E</sub> =31.6863±2.17851 t <sub>0</sub> =6836.52±1.17103			
	1.	Model L5 $\chi^2$ =127821. g <sub>OGLE</sub> =-0.289781±0.0915315				
	/	s=0.763874±0.0142947 q=0.000699372±0.000142686 u <sub>0</sub> =-0.545399±0.0316147 θ=1.2567±0.0218454 p=0.01110	98±0.00310309 t <sub>E</sub> =65.3398±4.13722 t <sub>0</sub> =6830.06±0.412378			
Binary lens models with parallax						
	1	$\frac{Model \times 1}{Model \times 1} = \chi^2 = 24202.1  g_{OGLE} = 5.86314 \pm 1.63923$				
	t	s=0.990599±0.0123037 q=0.000379833±0.000124866 u <sub>0</sub> =-0.0868334±0.0215884 θ=1.35672±0.066207 ρ-=0.0003 π <sub>⊥</sub> =0.0338196±0.469999 π <sub>  </sub> =-0.215736±0.321811	23885±0.00722684 t <sub>E</sub> =289.167±73.734 t <sub>0</sub> =6836.3±1.32788			

### Model pdf file



### Maximizing results

- We are moved by the idea that the **science output** of microlensing could be strongly improved, given the potential in the collected data.
- In order to speed-up the analysis and publication of the interesting events, we should make most of the work in a completely automatic way.
- Automatic pipelines and early warning systems are examples working on very large scales.
- Selecting anomalous events for intense follow-up observations is a very delicate task (ARTEMIS).
- Unfortunately, yet most planetary microlensing events are only discovered after the anomaly is over.

### Late-alert planets

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# Real-time modelling service

- As soon as an anomaly alert is issued, RTModel is able to automatically model the data and find preliminary models.
- Even if the final model may differ from those preliminary ones, the nature of the anomaly can be immediately guessed, ruling out competitors.



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14.

14.5

15

16

16.5

17.

Telescope

OGLE

Danish 1.54m

BaseLine

FB/FS

6.51565+0.0846791 2.05691+0.25551

LCOGT SAO A 20.3419±0.157467 -0.0127732±0.126623

16.8089±0.0962466 \_0.0103647±0.0610068



OGLE 16.8099±0.298534 2.97617±2.42739

### **Planetary Probability Indicator**

- For ongoing microlensing events we can build a planetary probability.
  - $\rightarrow$  Chi square
  - → Non-negative Blending constraint
  - $\rightarrow$  Source size
  - $\rightarrow$  Parallax
  - → Bayesian arguments
- A quantitative indicator to support follow-up decisions.

To Do!

### Facing future challenges

- NASA funded program to develop highly automated modeling code for the analysis of microlensing events.
  <u>PI: Rachel Street</u>, Co-Is: R. Barry, V. Bozza,
  Collaborators.: M. Dominik, K. Horne, M. Hundertmark, Y. Tsapras.
- Build on experience and capabilities of **RTModel**
- Develop the capacity to model microlensing events from WFIRST-AFTA
- Open Source Project: code will be publicly available
- Extensive verification: Data Challenge to test performance against existing packages



- 3yr **post-doctoral position** offered at LCOGT: deadline Feb 1, 2015
- See: lcogt.net/job/post-doc-microlensing jobregister.aas.org/node/50222

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### www.iiassvietri.it/en/ases2015.html

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