

All gBlock data

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Instructions

This file recreates the analyses in the publication, ‘A Statistical Model for Calibration and Computation of Detection and Quantification Limits for Low Copy Number Environmental DNA samples’ by Lesperance, Allison, Bergman, Hocking, Helbing, *eDNA*, 2021.

- Create a folder call Outputs in your working directory.
- Put files in working directory: PoissonCalib-Functions-7April2021.R, GEDWG_LOD_DATA3.csv
- Data set csv file requirements. Columns: Target, Lab, Cq, Sq
- DO NOT DUPLICATE Target names over different Labs!!
- Uses observations with nonempty data and where $\text{phat} < 1$ (num detect=num technical replicates)
- Only the SQ’s up to the first one with $\text{phat} == 1$ are used.
- Allows for variable numbers of SQ levels per Target.
- Assumes SQs == NA are zero, i.e. are negative controls.

- Adds in Negative Control (ntc) zeroes (24/(48 for Monroe) technical replicates).
- The code uses the R function optim. A convergence code 0 indicates successful completion.
- Ignore warnings if results are sensible.

Process/Summarize samples by Lab; Compute the Poisson estimates of SQ

Hindson et al “High-Throughput Droplet Digital PCR System for Absolute Quantitation of DNA Copy Number”, Anal. Chem., 2011, 83 (22), pp 8604–8610 use a Poisson approximation for quantitation. Before that, Dube et al. 2008, “Mathematical analysis of copy number variation in a DNA sample using digital PCR on a nanofluidic device”, POIS One, Vol 3, Issue 8, e2876, model the number of molecules in each chamber as a Poisson process, giving the relationship between p and λ .

```
## [1] 300    6

## dim, before negative controls added

##      Target      SQ      detect      n
## Hno      : 14  Min.    :      0  Min.    : 0.00  Min.    : 1.00
## MYPI-6 : 12  1st Qu.:      2  1st Qu.: 8.00  1st Qu.:24.00
## MYPI-6t: 12  Median :      50  Median :24.00  Median :24.00
## eASMO9 : 10  Mean    :    8414  Mean    :27.73  Mean    :36.58
## eASTR4 : 10  3rd Qu.:    1250  3rd Qu.:48.00  3rd Qu.:48.00
## eFISH1 : 10  Max.    :1000000  Max.    :96.00  Max.    :96.00
## (Other):232
##      Cqmean      Lab
## Min.    :16.32  UVIC   :110
## 1st Qu.:29.06  Monroe : 79
## Median :33.85  CERC   : 36
## Mean    :33.02  UMESC  : 31
## 3rd Qu.:37.70  Lance  : 14
## Max.    :47.20  Wilson : 14
## NA's    :24    (Other): 16

## [1] 325   13

## dim, Added 24/48 negative controls for Targets with zero ntc

##
##
## Table: APHIS
##
## Target      SQ      detect      n      phat
## -----
## D-loop      0.0         0      24      0.000
## D-loop      0.1         7      48      0.146
## D-loop      0.5        18      47      0.383
## D-loop      1.0        33      48      0.688
## D-loop     10.0        48      48      1.000
## D-loop     100.0       48      48      1.000
##
##
## Table: CERC
##
##      Target      SQ      detect      n      phat
```

##	---	-----	-----	---	-----
## 7	AC1_CID_1	0.000	0	24	0.000
## 8	AC1_CID_1	4.000	3	22	0.136
## 9	AC1_CID_1	8.000	12	23	0.522
## 10	AC1_CID_1	19.200	19	24	0.792
## 11	AC1_CID_1	192.000	24	24	1.000
## 12	AC1_CID_1	1920.000	24	24	1.000
## 13	AC1_CID_1	19200.000	24	24	1.000
## 14	CID	0.000	0	24	0.000
## 15	CID	15.625	22	22	1.000
## 16	CID	31.250	24	24	1.000
## 17	CID	62.500	24	24	1.000
## 18	CID	125.000	24	24	1.000
## 19	CID	250.000	24	24	1.000
## 20	CID	500.000	24	24	1.000
## 21	MYPI-6	0.000	0	24	0.000
## 22	MYPI-6	0.244	0	24	0.000
## 23	MYPI-6	0.489	0	24	0.000
## 24	MYPI-6	0.978	0	24	0.000
## 25	MYPI-6	1.955	0	24	0.000
## 26	MYPI-6	3.906	0	24	0.000
## 27	MYPI-6	7.812	0	24	0.000
## 28	MYPI-6	15.625	3	24	0.125
## 29	MYPI-6	31.250	2	24	0.083
## 30	MYPI-6	62.500	10	24	0.417
## 31	MYPI-6	125.000	4	24	0.167
## 32	MYPI-6	250.000	23	24	0.958
## 33	MYPI-6	500.000	24	24	1.000
## 34	MYPI-6t	0.000	0	24	0.000
## 35	MYPI-6t	0.244	3	24	0.125
## 36	MYPI-6t	0.488	4	24	0.167
## 37	MYPI-6t	0.975	9	24	0.375
## 38	MYPI-6t	1.950	9	24	0.375
## 39	MYPI-6t	3.906	22	24	0.917
## 40	MYPI-6t	7.812	22	24	0.917
## 41	MYPI-6t	15.625	24	24	1.000
## 42	MYPI-6t	31.250	24	24	1.000
## 43	MYPI-6t	62.500	24	24	1.000
## 44	MYPI-6t	125.000	24	24	1.000
## 45	MYPI-6t	250.000	24	24	1.000
## 46	MYPI-6t	500.000	23	24	0.958

##

##

Table: Goldberg

##

##	Target	SQ	detect	n	phat
##	---	-----	-----	---	-----
## 47	NZMS	0	0	1	0.000
## 48	NZMS	5	13	20	0.650
## 49	NZMS	10	22	22	1.000
## 50	NZMS	50	20	20	1.000
## 51	NZMS	100	22	22	1.000
## 52	NZMS	1000	2	2	1.000

```

## 53  NZMS      10000      2    2    1.000
## 54  SAFO        0      0   24    0.000
## 55  SAFO        5     14   19    0.737
## 56  SAFO       10     20   20    1.000
## 57  SAFO       50     20   20    1.000

```

```

##
##

```

```

## Table: Lance

```

```

##

```

##	Target	SQ	detect	n	phat
## ---	-----	-----	-----	---	-----
## 58	Elod-2-NU	0	0	48	0.000
## 59	Elod-2-NU	2	22	48	0.458
## 60	Elod-2-NU	10	47	47	1.000
## 61	Elod-2-NU	50	48	48	1.000
## 62	Elod-2-NU	250	48	48	1.000
## 63	Elod-2-NU	1250	48	48	1.000
## 64	Elod-2-NU	6250	48	48	1.000
## 65	MYPI	0	0	48	0.000
## 66	MYPI	2	31	48	0.646
## 67	MYPI	10	48	48	1.000
## 68	MYPI	50	48	48	1.000
## 69	MYPI	250	48	48	1.000
## 70	MYPI	1250	48	48	1.000
## 71	MYPI	6250	48	48	1.000

```

##
##

```

```

## Table: Monroe

```

```

##

```

##	Target	SQ	detect	n	phat
## ----	-----	-----	-----	---	-----
## 72	AC1	0.0	0	48	0.000
## 73	AC1	0.4	26	48	0.542
## 74	AC1	2.0	45	48	0.938
## 75	AC1	10.0	48	48	1.000
## 76	AC1	50.0	48	48	1.000
## 77	AC1	250.0	48	48	1.000
## 78	AC1	1250.0	48	48	1.000
## 79	AC1	6250.0	48	48	1.000
## 80	AC1	31250.0	48	48	1.000
## 81	AC3	0.0	0	48	0.000
## 82	AC3	0.4	25	48	0.521
## 83	AC3	2.0	45	48	0.938
## 84	AC3	10.0	48	48	1.000
## 85	AC3	50.0	48	48	1.000
## 86	AC3	250.0	48	48	1.000
## 87	AC3	1250.0	48	48	1.000
## 88	AC3	6250.0	48	48	1.000
## 89	AC3	31250.0	48	48	1.000
## 90	BH1	0.0	0	48	0.000
## 91	BH1	0.4	19	48	0.396
## 92	BH1	2.0	38	48	0.792
## 93	BH1	10.0	47	48	0.979

## 94	BH1	50.0	48	48	1.000
## 95	BH1	250.0	48	48	1.000
## 96	BH1	1250.0	48	48	1.000
## 97	BH1	6250.0	48	48	1.000
## 98	BH1	31250.0	48	48	1.000
## 99	BH2	0.0	0	48	0.000
## 100	BH2	0.4	20	48	0.417
## 101	BH2	2.0	34	48	0.708
## 102	BH2	10.0	47	48	0.979
## 103	BH2	50.0	48	48	1.000
## 104	BH2	250.0	48	48	1.000
## 105	BH2	1250.0	48	48	1.000
## 106	BH2	6250.0	48	48	1.000
## 107	BH2	31250.0	48	48	1.000
## 108	GCTM10	0.0	0	48	0.000
## 109	GCTM10	0.4	24	48	0.500
## 110	GCTM10	2.0	43	48	0.896
## 111	GCTM10	10.0	48	48	1.000
## 112	GCTM10	50.0	48	48	1.000
## 113	GCTM10	250.0	48	48	1.000
## 114	GCTM10	1250.0	48	48	1.000
## 115	GCTM10	6250.0	48	48	1.000
## 116	GCTM10	31300.0	48	48	1.000
## 117	GCTM22	0.0	0	48	0.000
## 118	GCTM22	0.4	22	48	0.458
## 119	GCTM22	2.0	44	48	0.917
## 120	GCTM22	10.0	48	48	1.000
## 121	GCTM22	50.0	48	48	1.000
## 122	GCTM22	250.0	48	48	1.000
## 123	GCTM22	1250.0	48	48	1.000
## 124	GCTM22	6250.0	48	48	1.000
## 125	GCTM22	31300.0	48	48	1.000
## 126	GCTM32	0.0	0	48	0.000
## 127	GCTM32	0.4	20	48	0.417
## 128	GCTM32	2.0	44	48	0.917
## 129	GCTM32	10.0	48	48	1.000
## 130	GCTM32	50.0	48	48	1.000
## 131	GCTM32	250.0	48	48	1.000
## 132	GCTM32	1250.0	48	48	1.000
## 133	GCTM32	6250.0	48	48	1.000
## 134	GCTM32	31300.0	48	48	1.000
## 135	Goby	0.0	0	6	0.000
## 136	Goby	1.0	26	46	0.565
## 137	Goby	10.0	46	46	1.000
## 138	Goby	100.0	46	46	1.000
## 139	Goby	1000.0	46	46	1.000
## 140	Goby	10000.0	46	46	1.000
## 141	Goby	100000.0	46	46	1.000
## 142	SC4	0.0	0	48	0.000
## 143	SC4	0.4	13	48	0.271
## 144	SC4	2.0	41	48	0.854
## 145	SC4	10.0	48	48	1.000
## 146	SC4	50.0	48	48	1.000

## 147	SC4	250.0	48	48	1.000
## 148	SC4	1250.0	48	48	1.000
## 149	SC4	6250.0	48	48	1.000
## 150	SC4	31250.0	48	48	1.000
## 151	SC5	0.0	0	48	0.000
## 152	SC5	0.4	14	48	0.292
## 153	SC5	2.0	42	48	0.875
## 154	SC5	10.0	48	48	1.000
## 155	SC5	50.0	48	48	1.000
## 156	SC5	250.0	48	48	1.000
## 157	SC5	1250.0	48	48	1.000
## 158	SC5	6250.0	48	48	1.000
## 159	SC5	31250.0	48	48	1.000

##

##

Table: UMESC

##

##	Target	SQ	detect	n	phat
## ----	-----	-----	-----	---	-----
## 160	BHC	0	0	96	0.000
## 161	BHC	1	25	96	0.260
## 162	BHC	5	59	96	0.615
## 163	BHC	10	96	96	1.000
## 164	BHC	100	96	96	1.000
## 165	BHC	1000	96	96	1.000
## 166	BHC	10000	96	96	1.000
## 167	Dre16s	0	1	96	0.010
## 168	Dre16s	1	9	96	0.094
## 169	Dre16s	5	42	96	0.438
## 170	Dre16s	10	68	96	0.708
## 171	Dre16s	100	96	96	1.000
## 172	Dre16s	1000	96	96	1.000
## 173	Dre16s	10000	96	96	1.000
## 174	SS	0	0	96	0.000
## 175	SS	1	44	96	0.458
## 176	SS	10	96	96	1.000
## 177	SS	100	96	96	1.000
## 178	SS	1000	96	96	1.000
## 179	SVC	0	0	96	0.000
## 180	SVC	1	25	96	0.260
## 181	SVC	5	59	96	0.615
## 182	SVC	10	96	96	1.000
## 183	SVC	100	96	96	1.000
## 184	SVC	1000	96	96	1.000
## 185	SVC	10000	96	96	1.000
## 186	YPC	0	0	96	0.000
## 187	YPC	1	31	96	0.323
## 188	YPC	10	94	96	0.979
## 189	YPC	100	96	96	1.000
## 190	YPC	1000	96	96	1.000

##

##

Table: UVIC

##	Target	SQ	detect	n	phat
##	----	-----	-----	---	-----
## 191	eASM09	0.00e+00	0	24	0.000
## 192	eASM09	3.20e-02	0	24	0.000
## 193	eASM09	1.60e-01	6	24	0.250
## 194	eASM09	8.00e-01	11	24	0.458
## 195	eASM09	4.00e+00	22	24	0.917
## 196	eASM09	2.00e+01	24	24	1.000
## 197	eASM09	1.00e+02	24	24	1.000
## 198	eASM09	5.00e+02	8	8	1.000
## 199	eASM09	2.50e+03	8	8	1.000
## 200	eASM09	1.25e+04	8	8	1.000
## 201	eASM09	6.25e+04	8	8	1.000
## 202	eASTR4	0.00e+00	0	24	0.000
## 203	eASTR4	3.20e-02	0	24	0.000
## 204	eASTR4	1.60e-01	4	24	0.167
## 205	eASTR4	8.00e-01	9	24	0.375
## 206	eASTR4	4.00e+00	19	24	0.792
## 207	eASTR4	2.00e+01	24	24	1.000
## 208	eASTR4	1.00e+02	24	24	1.000
## 209	eASTR4	5.00e+02	8	8	1.000
## 210	eASTR4	2.50e+03	8	8	1.000
## 211	eASTR4	1.25e+04	8	8	1.000
## 212	eASTR4	6.25e+04	8	8	1.000
## 213	eFISH1	0.00e+00	0	24	0.000
## 214	eFISH1	3.20e-02	1	24	0.042
## 215	eFISH1	1.60e-01	0	24	0.000
## 216	eFISH1	8.00e-01	10	24	0.417
## 217	eFISH1	4.00e+00	18	24	0.750
## 218	eFISH1	2.00e+01	24	24	1.000
## 219	eFISH1	1.00e+02	24	24	1.000
## 220	eFISH1	5.00e+02	8	8	1.000
## 221	eFISH1	2.50e+03	8	8	1.000
## 222	eFISH1	1.25e+04	8	8	1.000
## 223	eFISH1	6.25e+04	8	8	1.000
## 224	eLIPI1	0.00e+00	0	24	0.000
## 225	eLIPI1	3.20e-02	0	24	0.000
## 226	eLIPI1	1.60e-01	5	24	0.208
## 227	eLIPI1	8.00e-01	11	24	0.458
## 228	eLIPI1	4.00e+00	23	24	0.958
## 229	eLIPI1	2.00e+01	24	24	1.000
## 230	eLIPI1	1.00e+02	24	24	1.000
## 231	eLIPI1	5.00e+02	8	8	1.000
## 232	eLIPI1	2.50e+03	8	8	1.000
## 233	eLIPI1	1.25e+04	8	8	1.000
## 234	eLIPI1	6.25e+04	8	8	1.000
## 235	eMID01	0.00e+00	0	24	0.000
## 236	eMID01	3.20e-02	0	24	0.000
## 237	eMID01	1.60e-01	2	24	0.083
## 238	eMID01	8.00e-01	16	24	0.667
## 239	eMID01	4.00e+00	23	24	0.958
## 240	eMID01	2.00e+01	24	24	1.000

## 241	eMID01	1.00e+02	24	24	1.000
## 242	eMID01	5.00e+02	8	8	1.000
## 243	eMID01	2.50e+03	8	8	1.000
## 244	eMID01	1.25e+04	8	8	1.000
## 245	eMID01	6.25e+04	8	8	1.000
## 246	eMISA2	0.00e+00	0	24	0.000
## 247	eMISA2	3.20e-02	0	24	0.000
## 248	eMISA2	1.60e-01	3	24	0.125
## 249	eMISA2	8.00e-01	10	24	0.417
## 250	eMISA2	4.00e+00	21	24	0.875
## 251	eMISA2	2.00e+01	24	24	1.000
## 252	eMISA2	1.00e+02	24	24	1.000
## 253	eMISA2	5.00e+02	8	8	1.000
## 254	eMISA2	2.50e+03	8	8	1.000
## 255	eMISA2	1.25e+04	8	8	1.000
## 256	eMISA2	6.25e+04	8	8	1.000
## 257	eONKI4	0.00e+00	0	24	0.000
## 258	eONKI4	3.20e-02	3	24	0.125
## 259	eONKI4	1.60e-01	6	24	0.250
## 260	eONKI4	8.00e-01	14	24	0.583
## 261	eONKI4	4.00e+00	21	24	0.875
## 262	eONKI4	2.00e+01	24	24	1.000
## 263	eONKI4	1.00e+02	24	24	1.000
## 264	eONKI4	5.00e+02	8	8	1.000
## 265	eONKI4	2.50e+03	8	8	1.000
## 266	eONKI4	1.25e+04	8	8	1.000
## 267	eONKI4	6.25e+04	8	8	1.000
## 268	eRAAU1	0.00e+00	0	24	0.000
## 269	eRAAU1	3.20e-02	1	24	0.042
## 270	eRAAU1	1.60e-01	2	24	0.083
## 271	eRAAU1	8.00e-01	7	24	0.292
## 272	eRAAU1	4.00e+00	20	24	0.833
## 273	eRAAU1	2.00e+01	24	24	1.000
## 274	eRAAU1	1.00e+02	24	24	1.000
## 275	eRAAU1	5.00e+02	8	8	1.000
## 276	eRAAU1	2.50e+03	8	8	1.000
## 277	eRAAU1	1.25e+04	8	8	1.000
## 278	eRAAU1	6.25e+04	8	8	1.000
## 279	eRACA2	0.00e+00	0	24	0.000
## 280	eRACA2	3.20e-02	0	24	0.000
## 281	eRACA2	1.60e-01	0	24	0.000
## 282	eRACA2	8.00e-01	11	24	0.458
## 283	eRACA2	4.00e+00	21	24	0.875
## 284	eRACA2	2.00e+01	24	24	1.000
## 285	eRACA2	1.00e+02	24	24	1.000
## 286	eRACA2	5.00e+02	8	8	1.000
## 287	eRACA2	2.50e+03	8	8	1.000
## 288	eRACA2	1.25e+04	8	8	1.000
## 289	eRACA2	6.25e+04	8	8	1.000
## 290	eRALU2	0.00e+00	0	24	0.000
## 291	eRALU2	3.20e-02	1	24	0.042
## 292	eRALU2	1.60e-01	2	24	0.083
## 293	eRALU2	8.00e-01	5	24	0.208


```

## 294 eRALU2 4.00e+00 20 24 0.833
## 295 eRALU2 2.00e+01 24 24 1.000
## 296 eRALU2 1.00e+02 24 24 1.000
## 297 eRALU2 5.00e+02 8 8 1.000
## 298 eRALU2 2.50e+03 8 8 1.000
## 299 eRALU2 1.25e+04 8 8 1.000
## 300 eRALU2 6.25e+04 8 8 1.000
## 301 eRAPR2 0.00e+00 0 24 0.000
## 302 eRAPR2 3.20e-02 1 24 0.042
## 303 eRAPR2 1.60e-01 2 24 0.083
## 304 eRAPR2 8.00e-01 12 24 0.500
## 305 eRAPR2 4.00e+00 22 24 0.917
## 306 eRAPR2 2.00e+01 24 24 1.000
## 307 eRAPR2 1.00e+02 24 24 1.000
## 308 eRAPR2 5.00e+02 8 8 1.000
## 309 eRAPR2 2.50e+03 8 8 1.000
## 310 eRAPR2 1.25e+04 8 8 1.000
## 311 eRAPR2 6.25e+04 8 8 1.000

```

```

##
##
## Table: Wilson

```

```

##
##      Target      SQ  detect    n    phat
## ---- -
## 312 Hno          0.0      0    48  0.000
## 313 Hno          0.2      7    24  0.292
## 314 Hno          1.0     37    48  0.771
## 315 Hno          5.0     24    24  1.000
## 316 Hno         10.0     24    24  1.000
## 317 Hno         25.0     24    24  1.000
## 318 Hno        100.0     24    24  1.000
## 319 Hno       125.0     24    24  1.000
## 320 Hno       625.0     24    24  1.000
## 321 Hno      1000.0     24    24  1.000
## 322 Hno     3125.0     24    24  1.000
## 323 Hno    10000.0     24    24  1.000
## 324 Hno   100000.0     24    24  1.000
## 325 Hno  1000000.0     24    24  1.000

```

Plot the Poisson estimates (and CI) of SQ for levels that had non-detects

Only the first levels of SQ that had non-detects are analyzed. Blue line is least squares linear regression line.

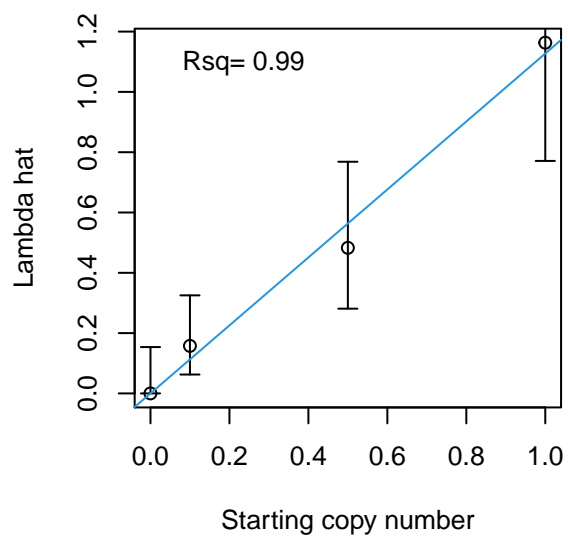
ML ??will error if all phats==1

```

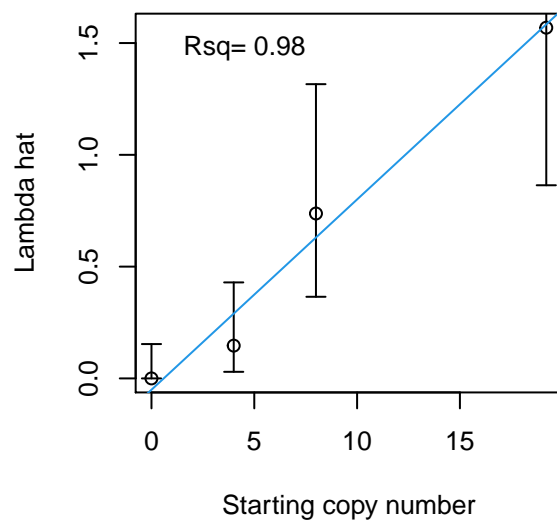
##
##
## Too few values for CID

```

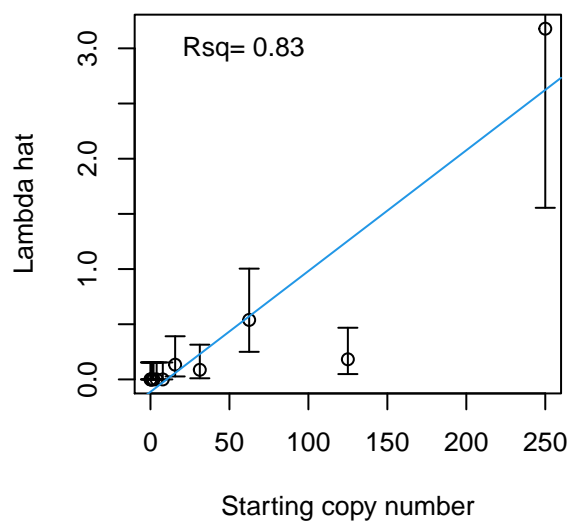
APHIS, D-loop



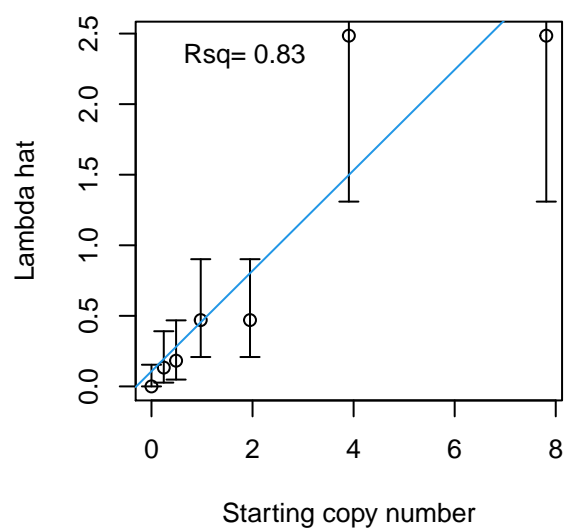
CERC, AC1_CID_1



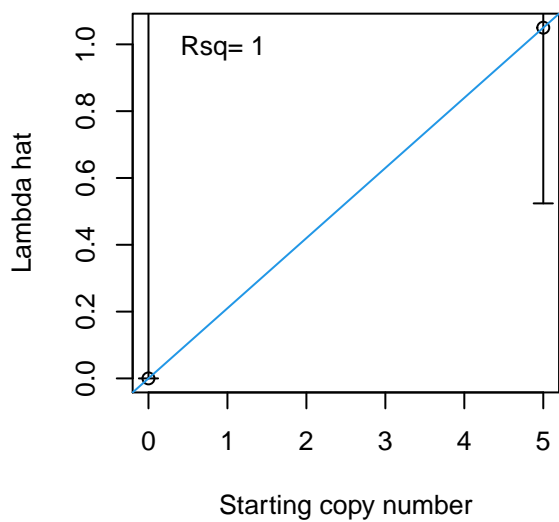
CERC, MYPI-6



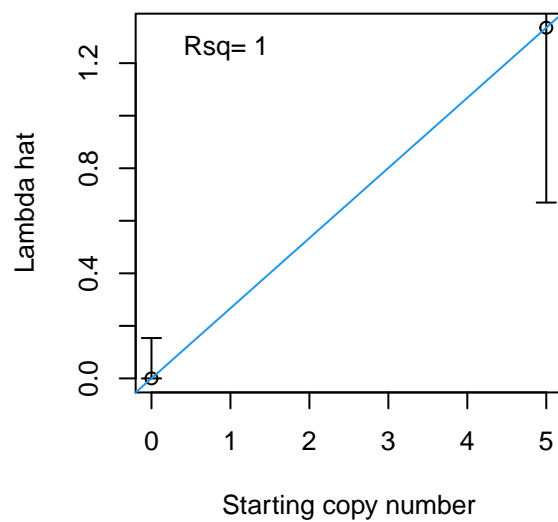
CERC, MYPI-6t



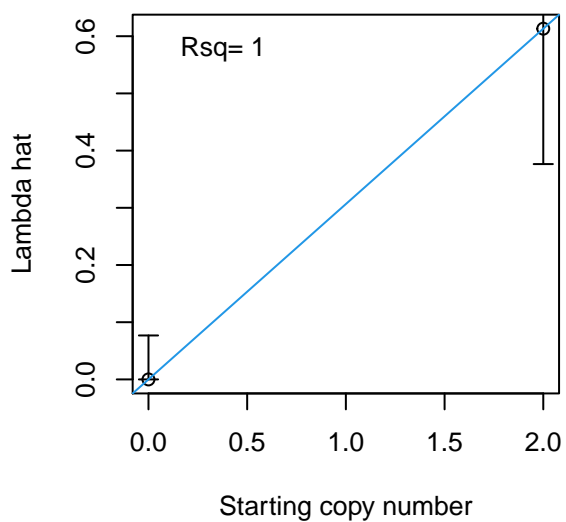
Goldberg, NZMS



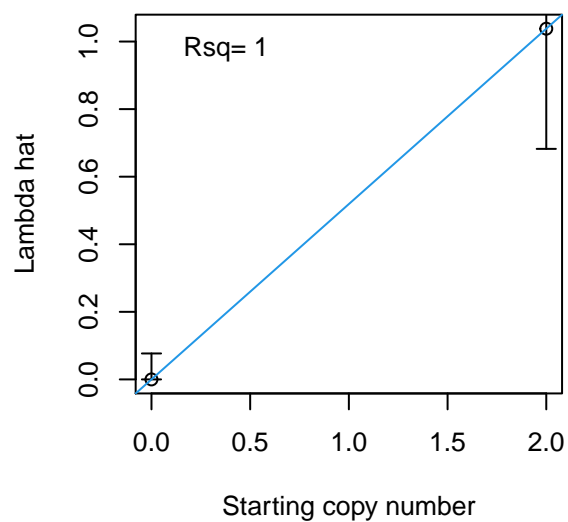
Goldberg, SAFO



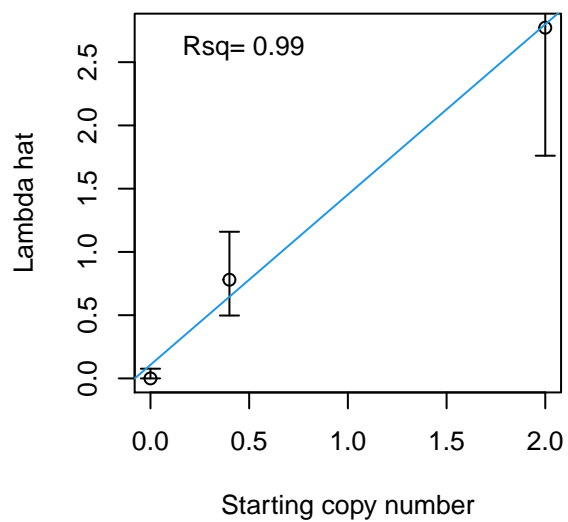
Lance, Elod-2-NU



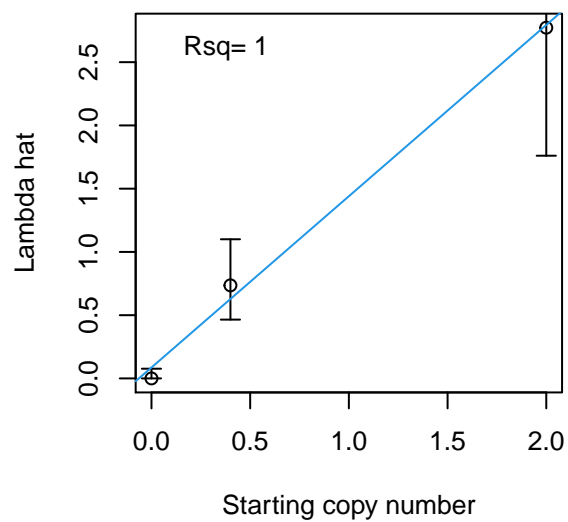
Lance, MYPI



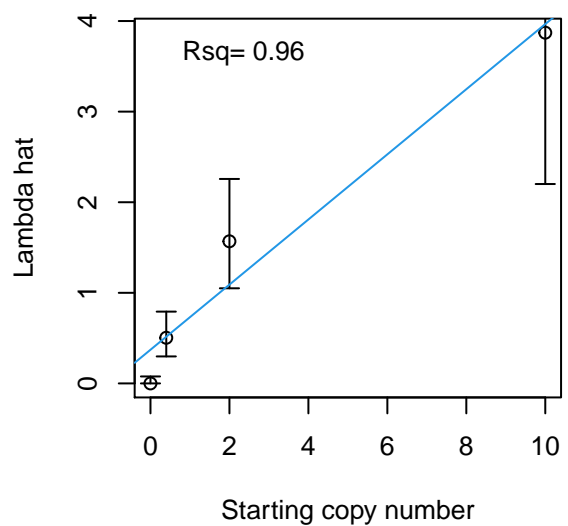
Monroe, AC1



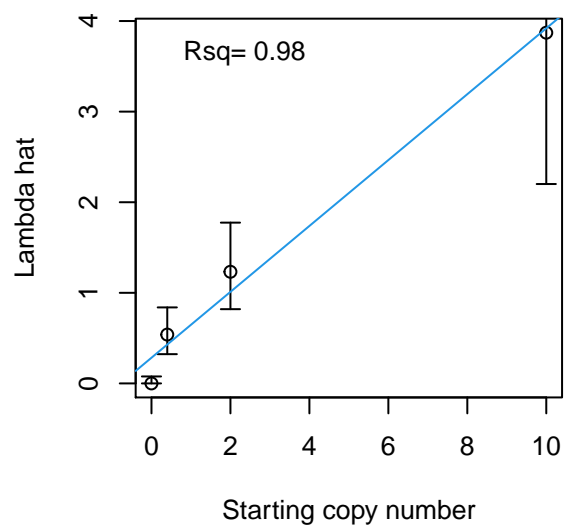
Monroe, AC3



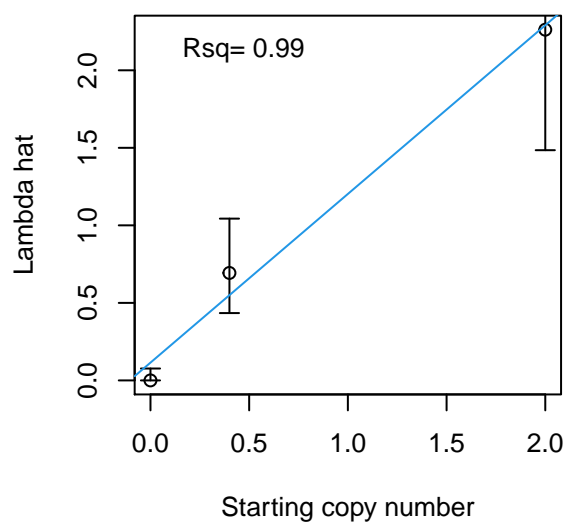
Monroe, BH1



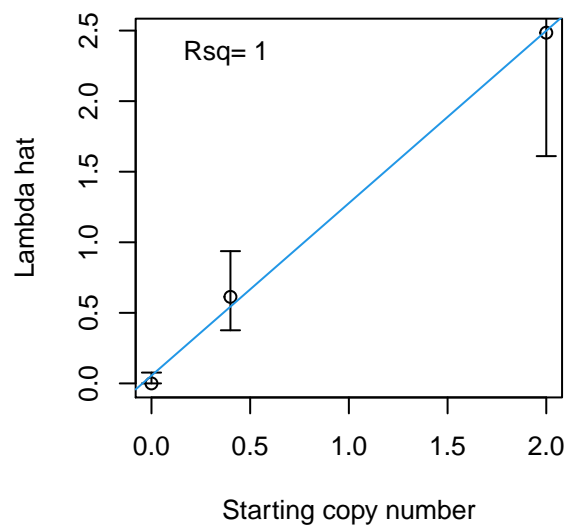
Monroe, BH2



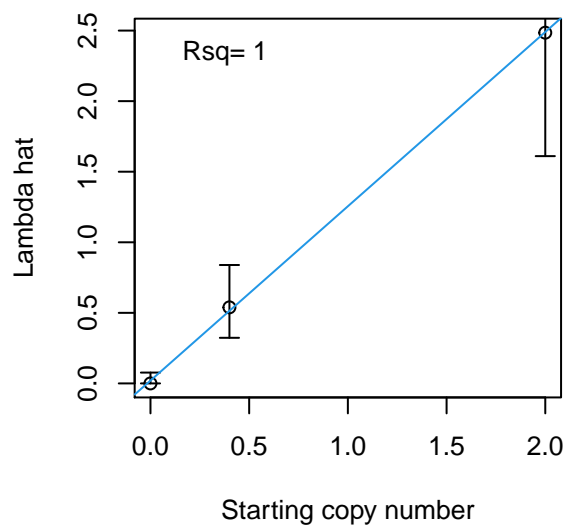
Monroe, GCTM10



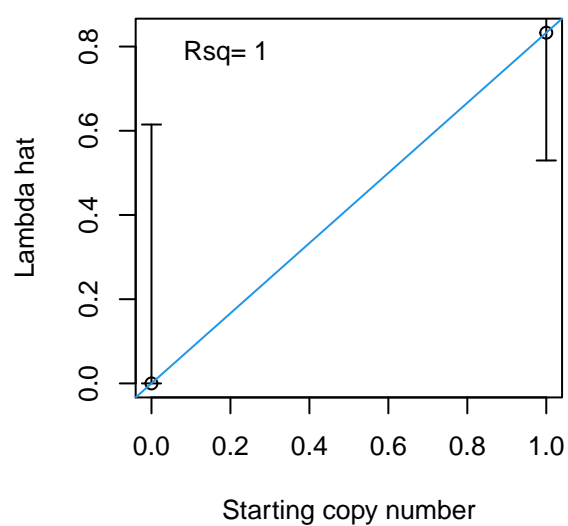
Monroe, GCTM22



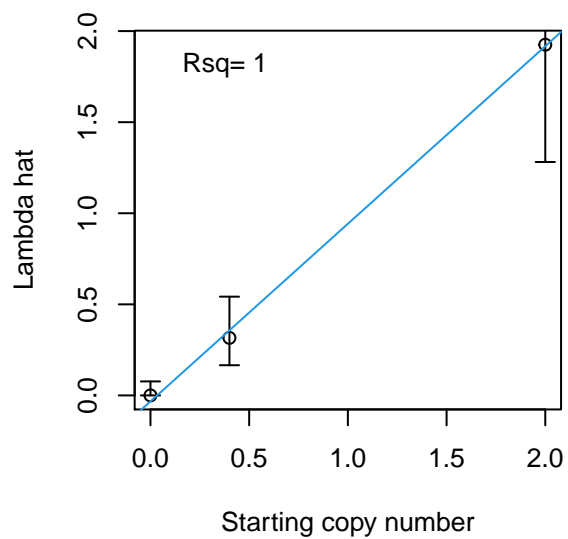
Monroe, GCTM32



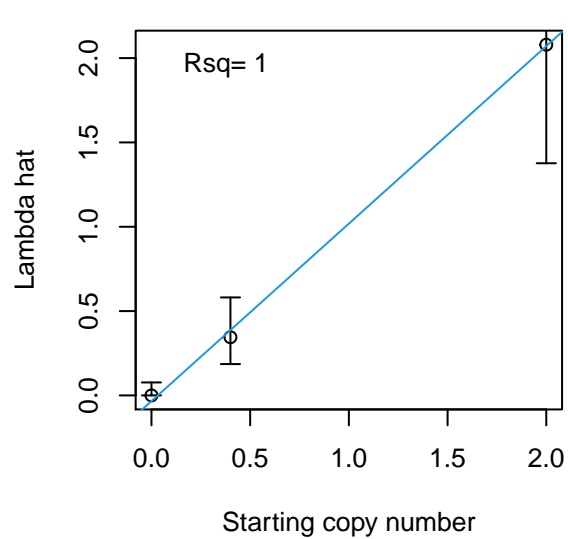
Monroe, Goby



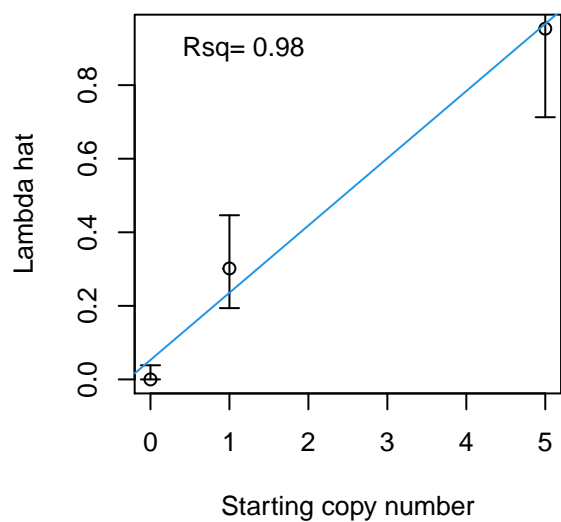
Monroe, SC4



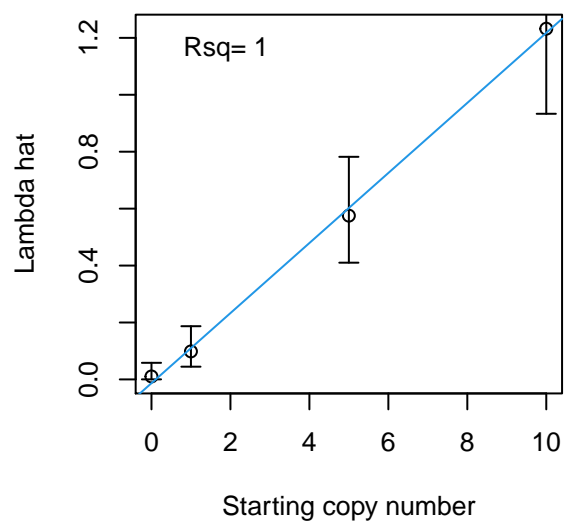
Monroe, SC5



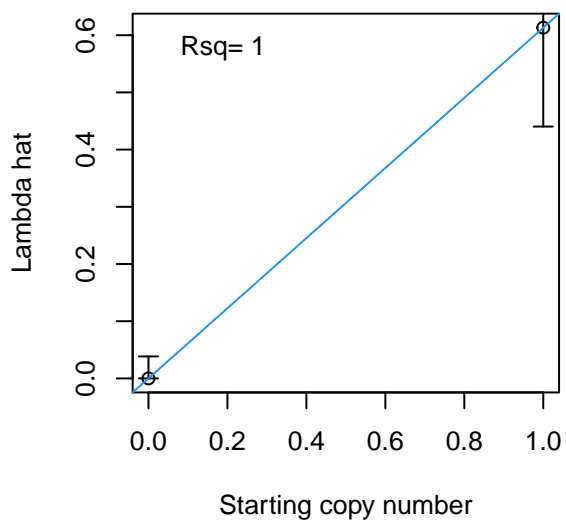
UMESC, BHC



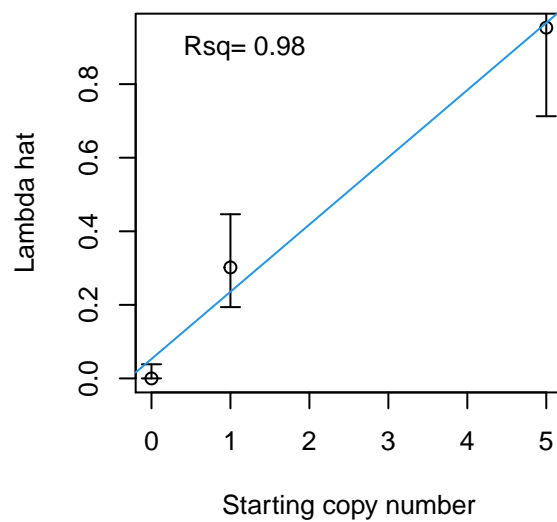
UMESC, Dre16s



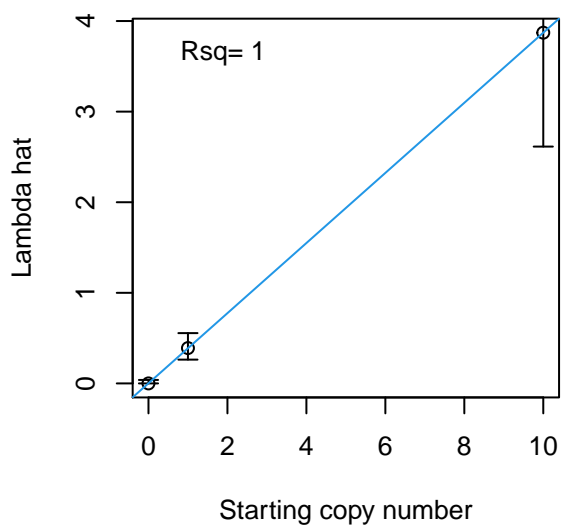
UMESC, SS



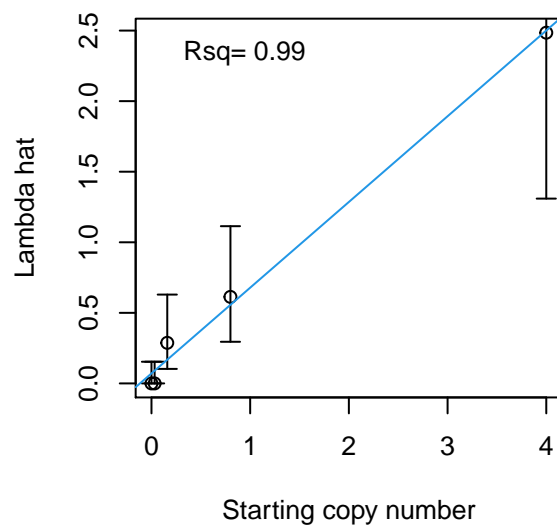
UMESC, SVC



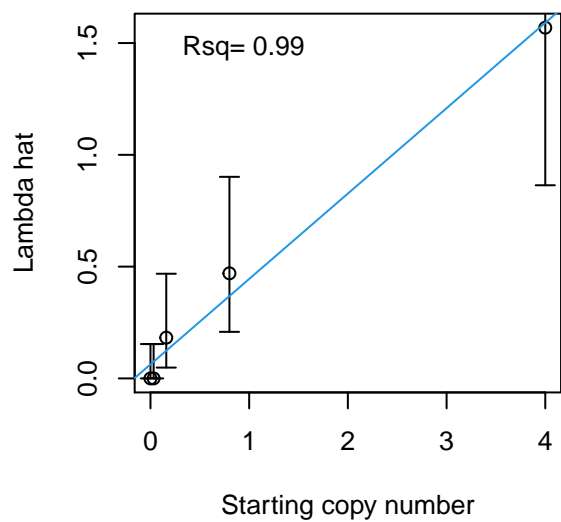
UMESC, YPC



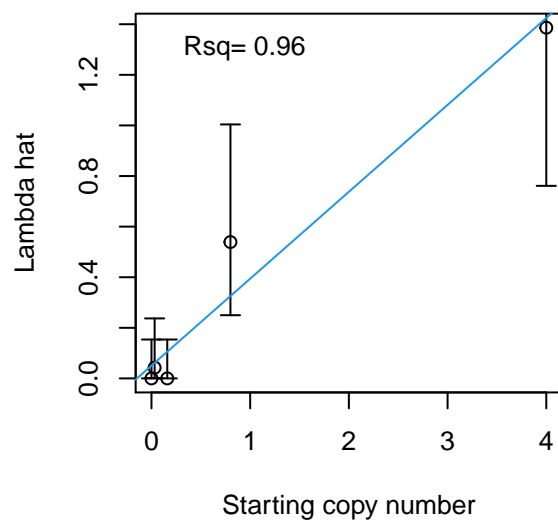
UVIC, eASMO9



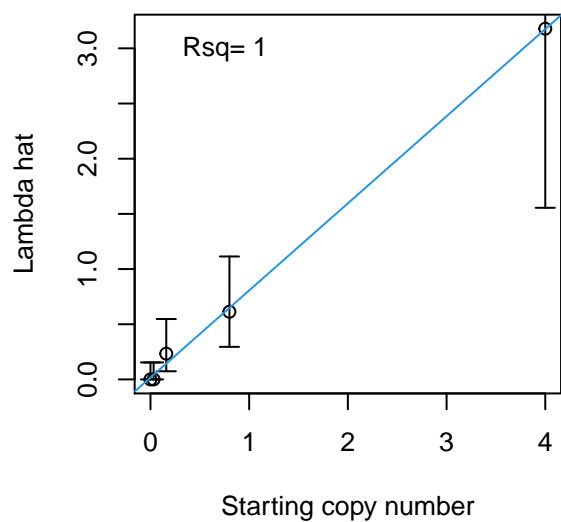
UVIC, eASTR4



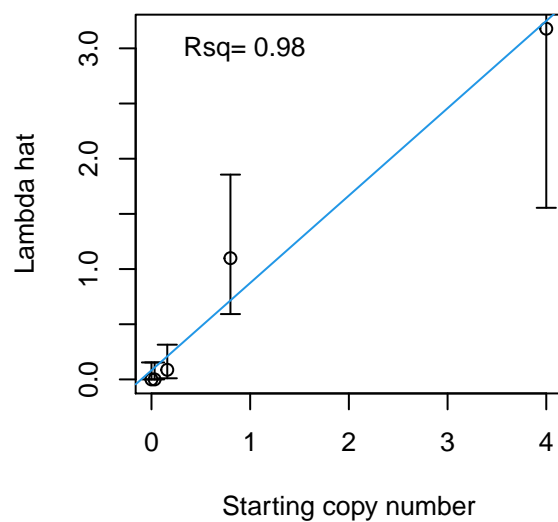
UVIC, eFISH1



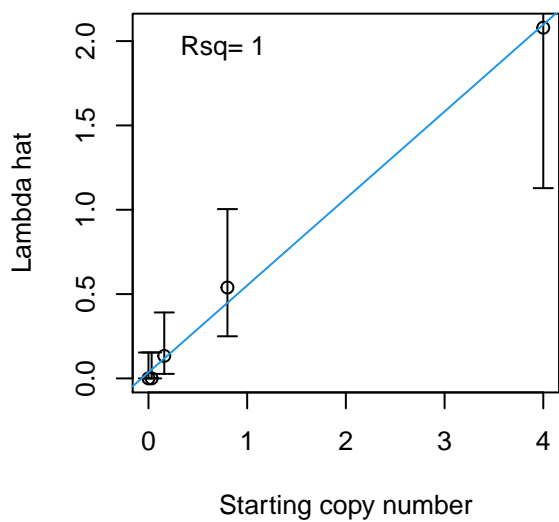
UVIC, eLIP11



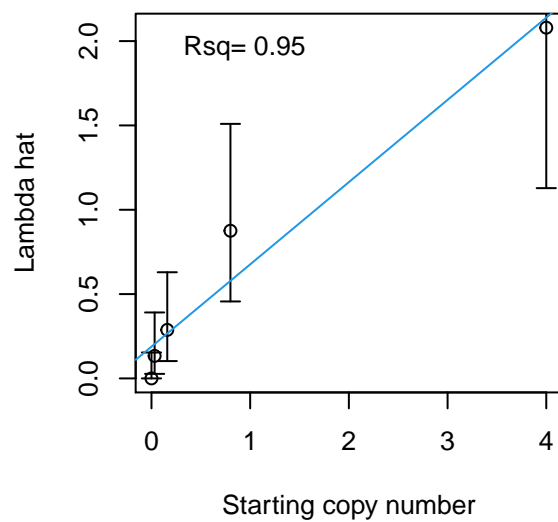
UVIC, eMIDO1



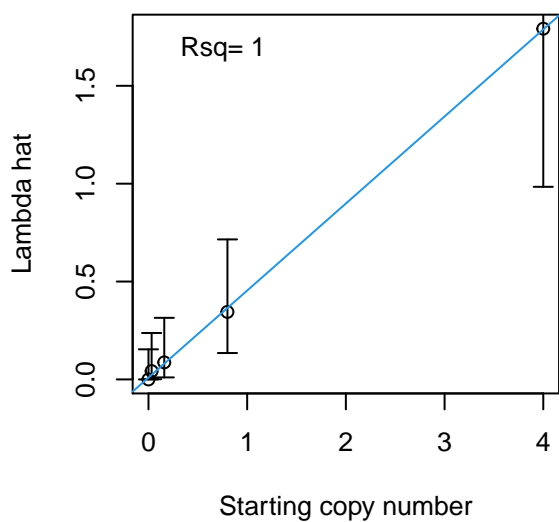
UVIC, eMISA2



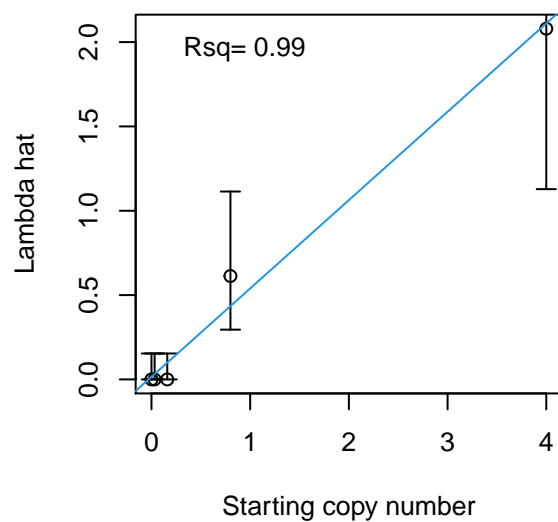
UVIC, eONKI4



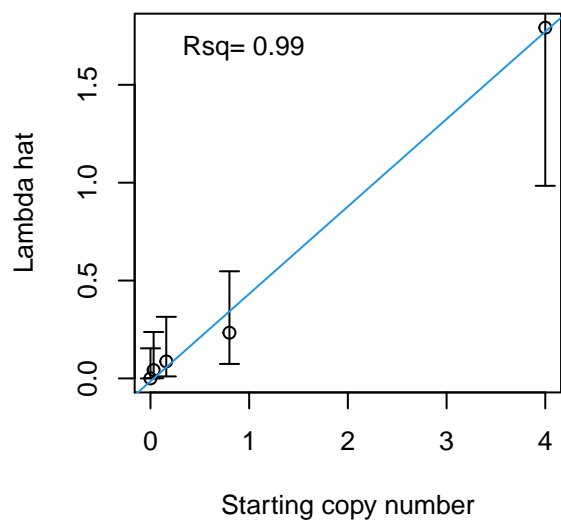
UVIC, eRAAU1



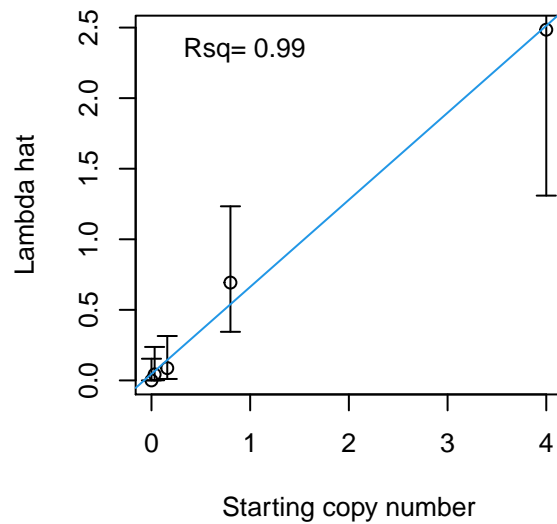
UVIC, eRACA2



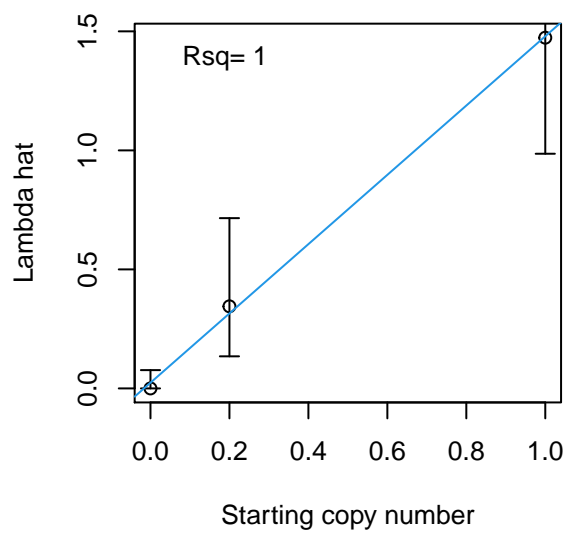
UVIC, eRALU2



UVIC, eRAPR2



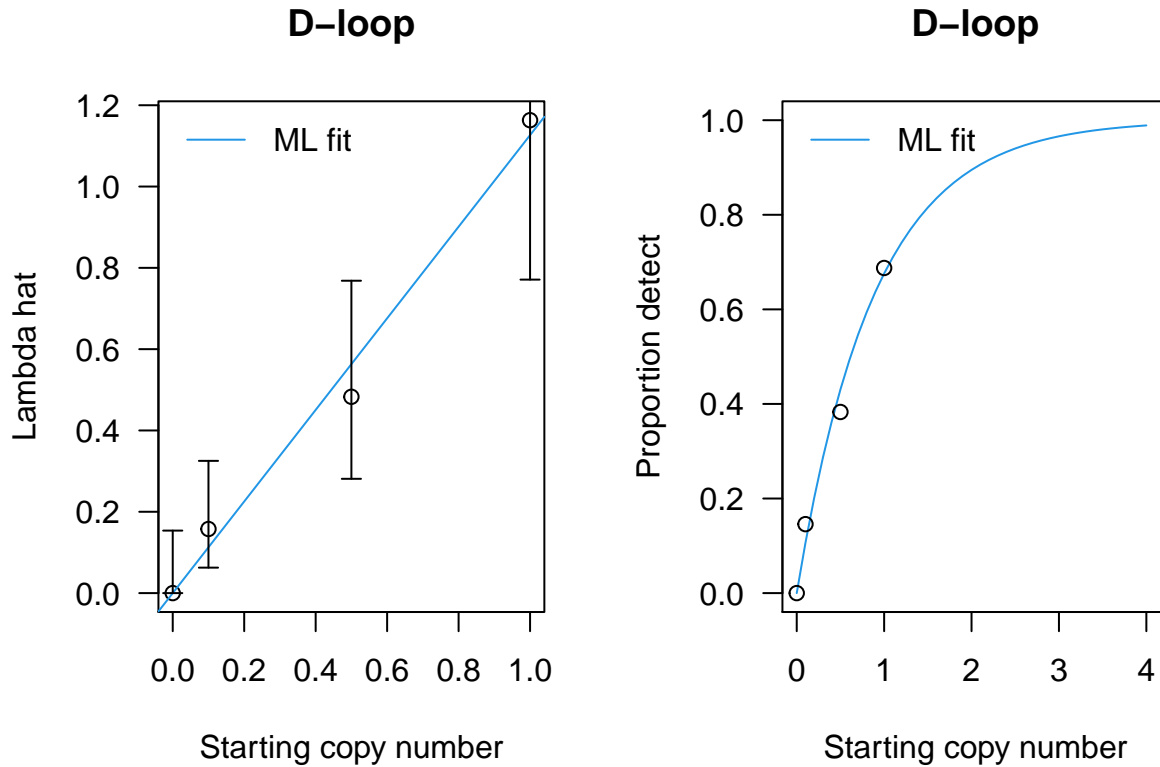
Wilson, Hno



Supplemental Material

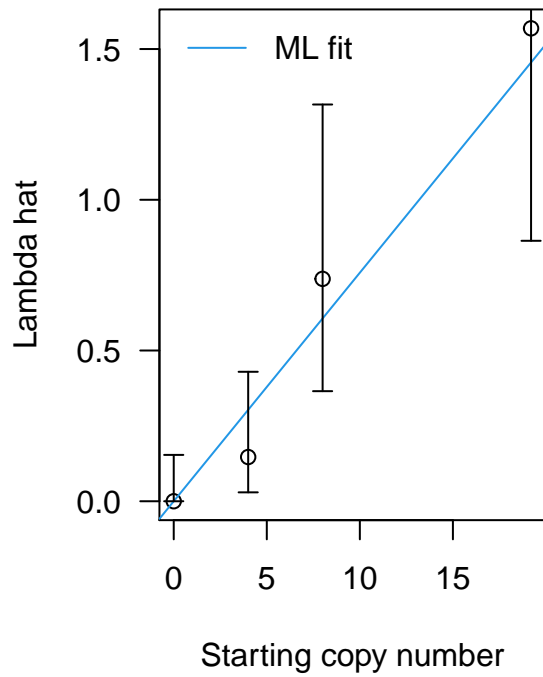
This file contains the outputs from all 29 data sets from Klymus et al. (2019) whose data are summarized in Table 4 of the manuscript. Results from both no intercept and intercept models are presented.

Estimate Poisson models - no intercept model

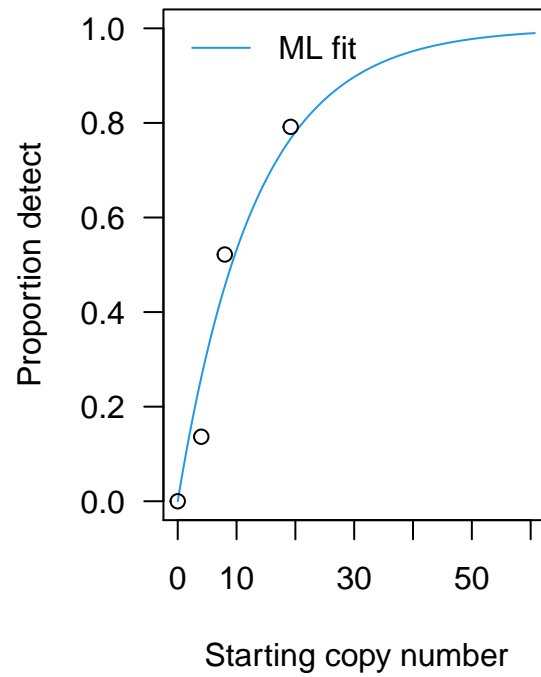


```
##
##
##
##
## D-loop
## Convergence= 0
##      Estimate Std.Err Z value  Pr(>z)
## beta    1.126   0.153    7.37 1.8e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.17757 , df= 3 , p-value= 0.7583889
```

AC1_CID_1

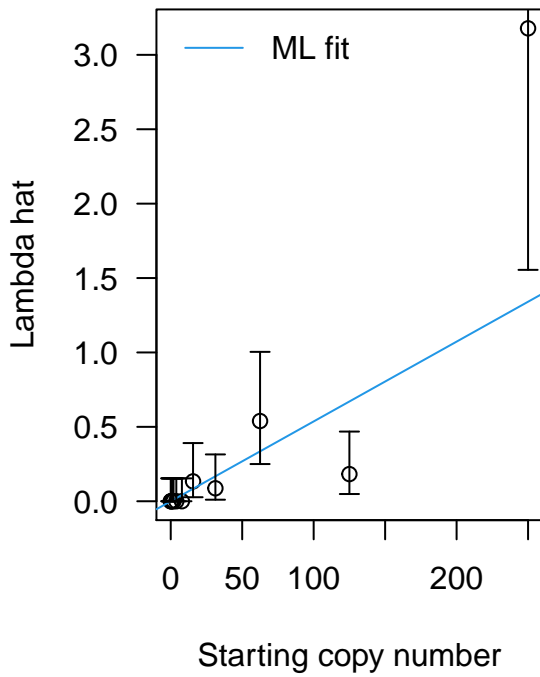


AC1_CID_1

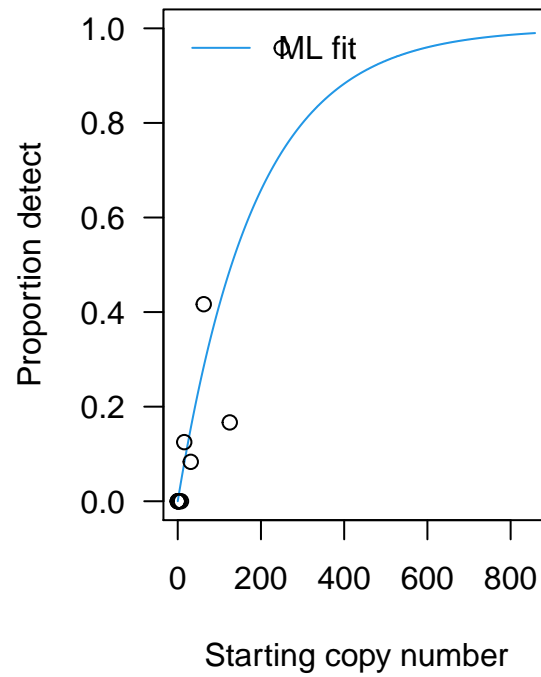


```
##
##
##
##
## AC1_CID_1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta  0.0758  0.0137    5.53 3.2e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.543745 , df= 3 , p-value= 0.4674373
##
## Too few values for CID
```

MYPI-6

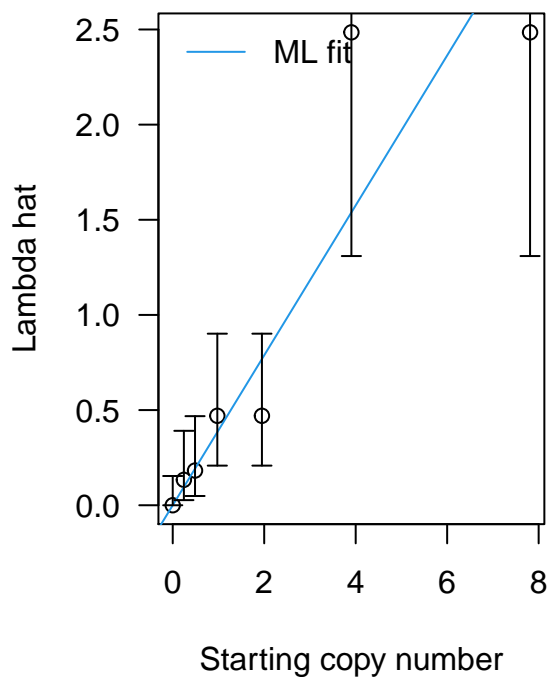


MYPI-6

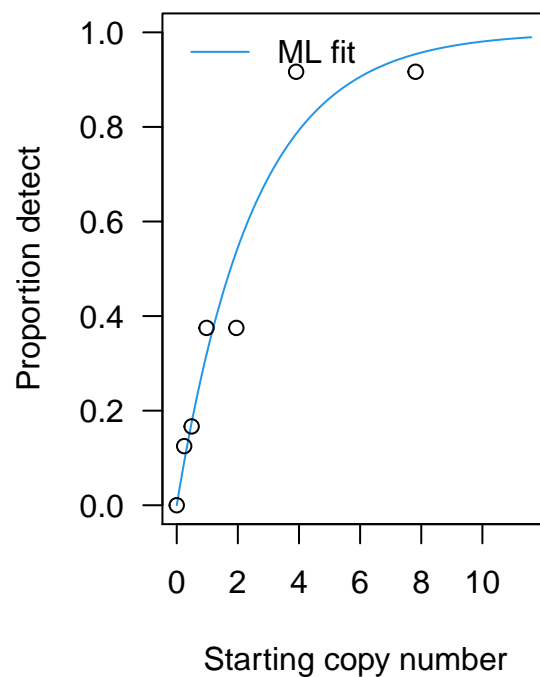


```
##
##
##
##
## MYPI-6
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta 0.005363 0.000847   6.33 2.4e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 26.74636 , df= 11 , p-value= 0.005018208
```

MYPI-6t

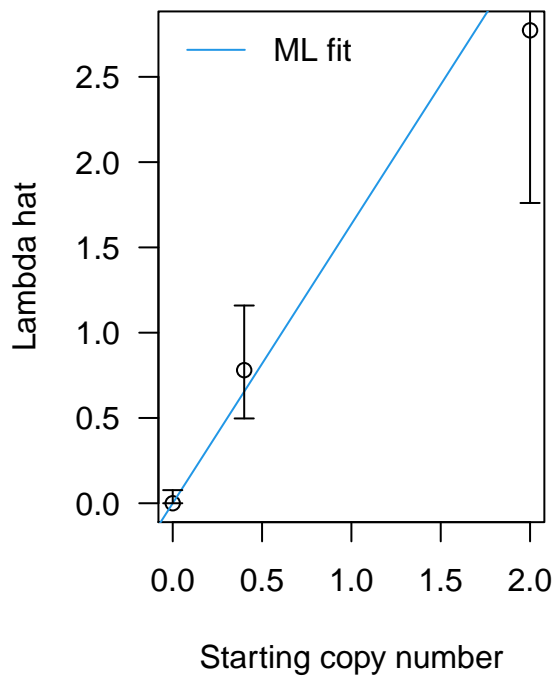


MYPI-6t

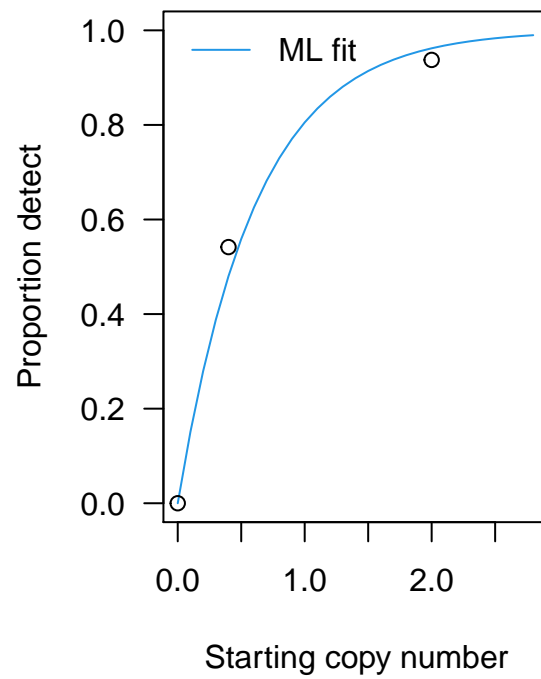


```
##
##
##
##
## MYPI-6t
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.394   0.054   7.29 3.1e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.787625 , df= 6 , p-value= 0.340935
##
## Too few values for  NZMS
## Too few values for  SAFO
## Too few values for  Elod-2-NU
## Too few values for  MYPI
```

AC1

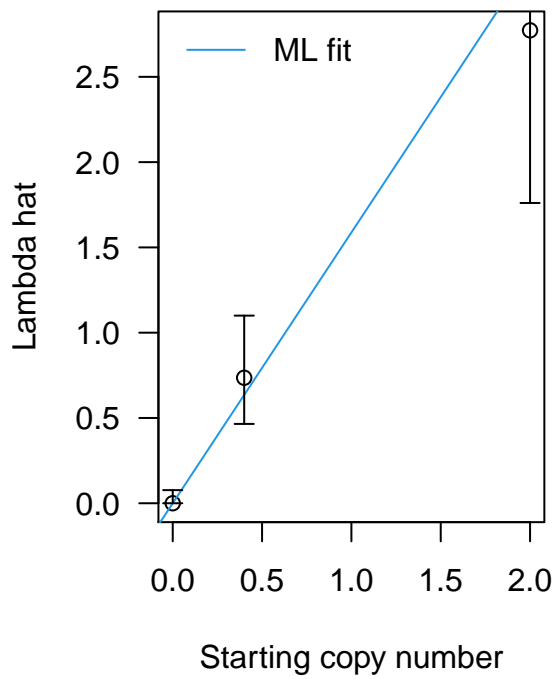


AC1

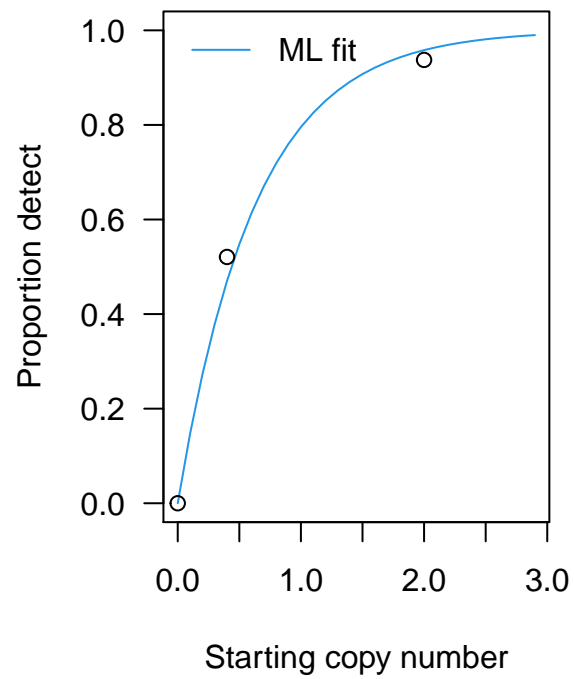


```
##
##
##
##
## AC1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    1.636   0.244    6.7 2.1e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.393097 , df= 2 , p-value= 0.4983022
```

AC3

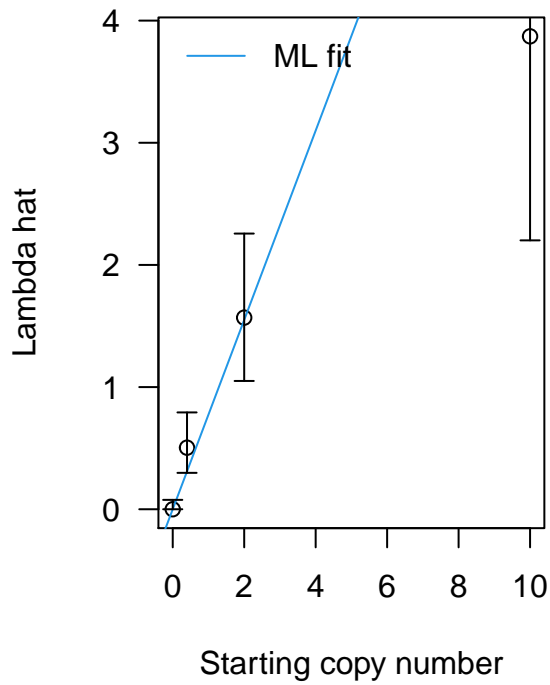


AC3

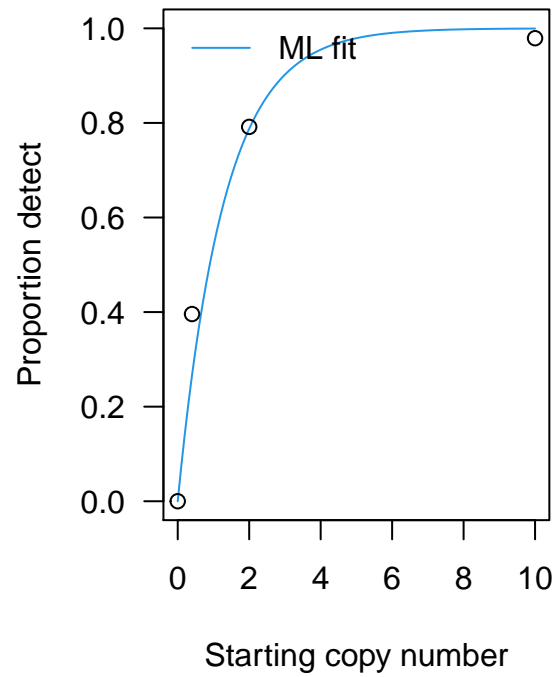


```
##
##
##
##
## AC3
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    1.587   0.237    6.7 2.2e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.9438297 , df= 2 , p-value= 0.6238066
```


BH1

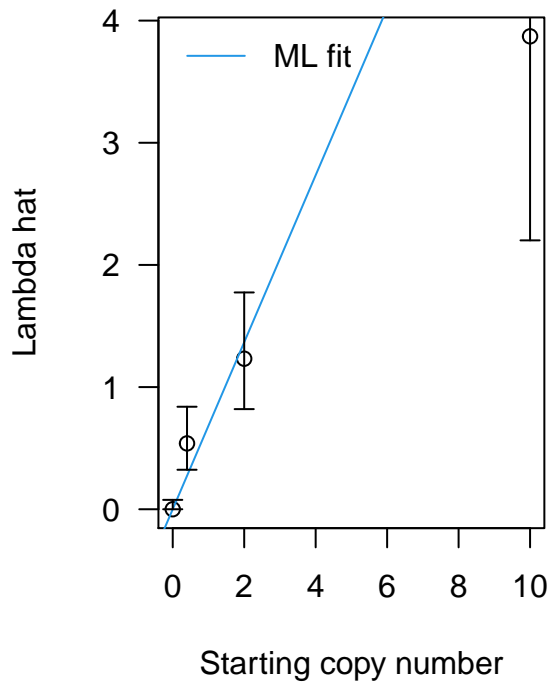


BH1

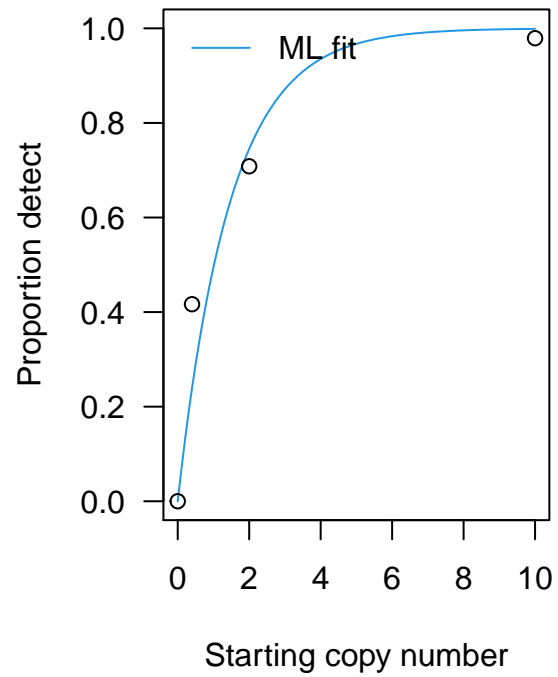


```
##
##
##
##
## BH1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.775   0.108   7.16  8e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 9.602954 , df= 3 , p-value= 0.02226096
```

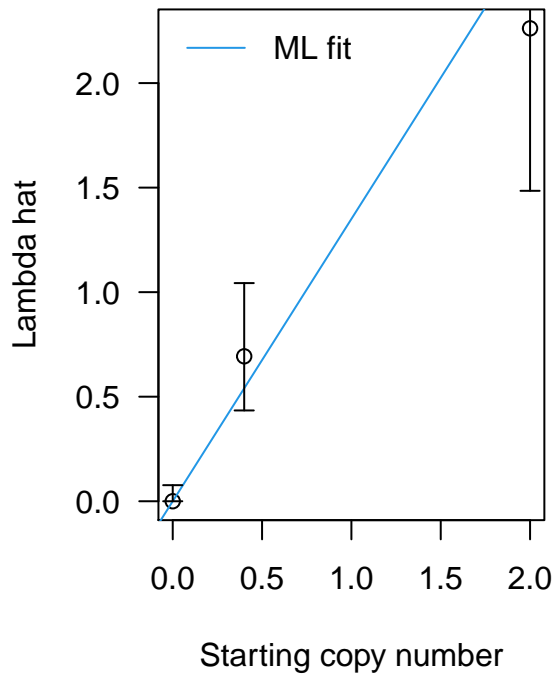
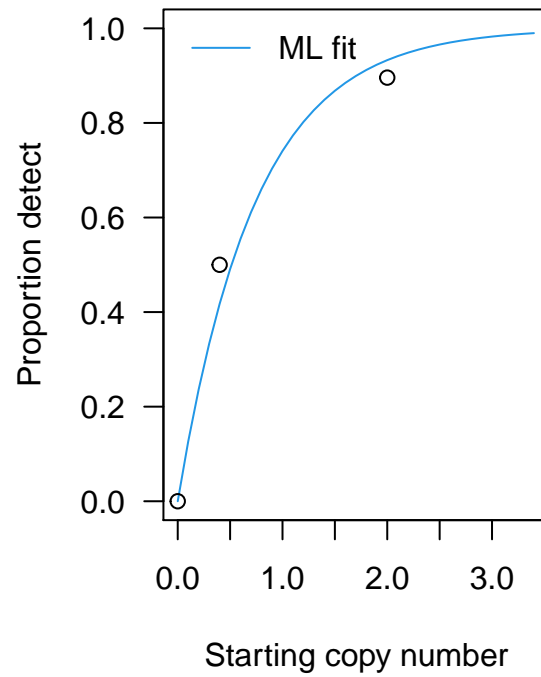
BH2



BH2

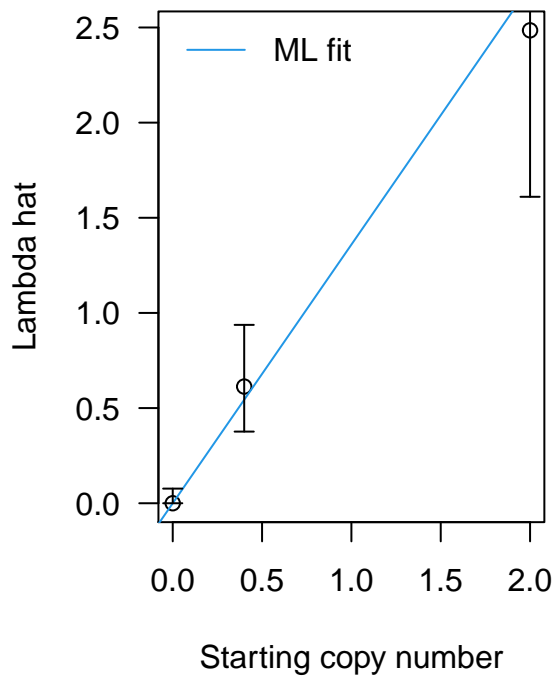


```
##
##
##
##
## BH2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.6836  0.0953   7.17 7.5e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 11.70607 , df= 3 , p-value= 0.008460941
```

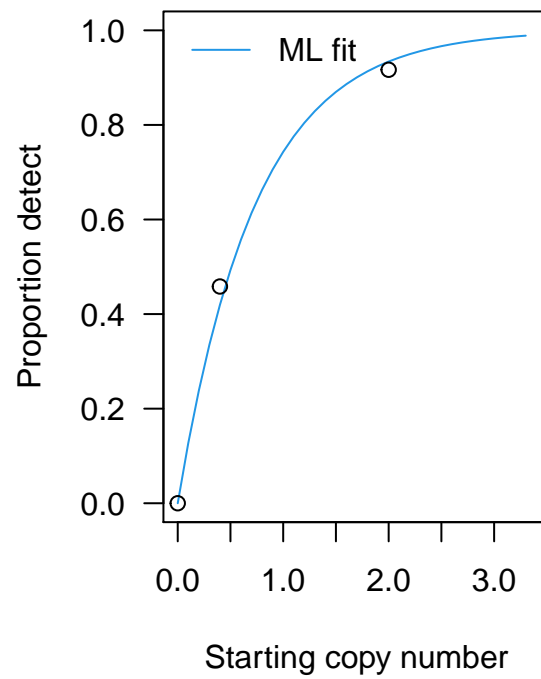
GCTM10**GCTM10**

```
##
##
##
##
## GCTM10
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    1.350   0.196    6.9 5.1e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.23836 , df= 2 , p-value= 0.3265475
```

GCTM22

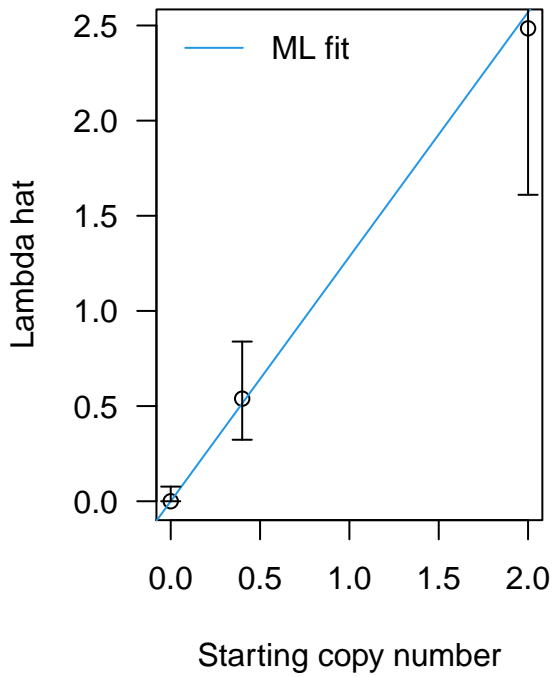


GCTM22

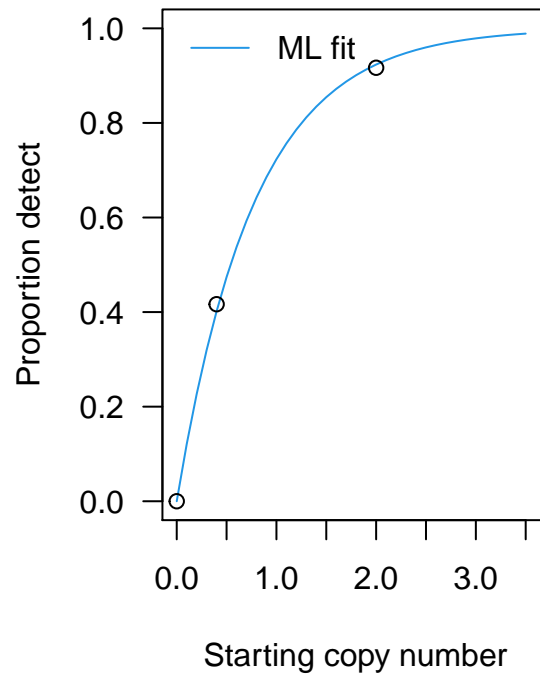


```
##
##
##
##
## GCTM22
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta      1.36    0.20   6.79 1.1e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.5143054 , df= 2 , p-value= 0.7732501
```

GCTM32

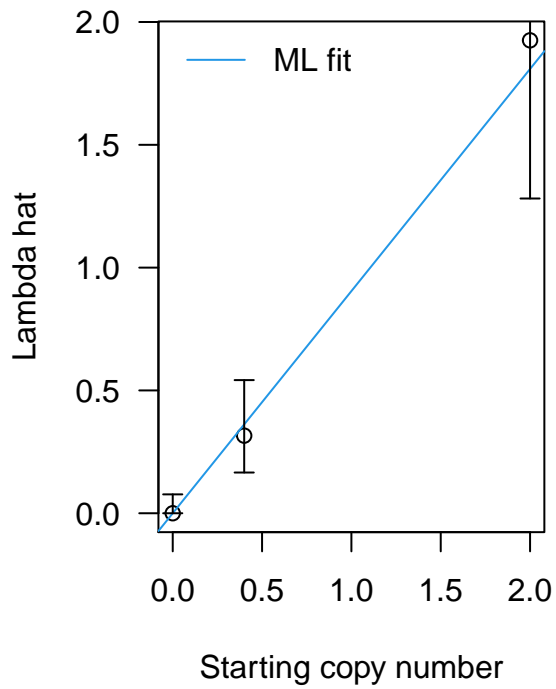


GCTM32

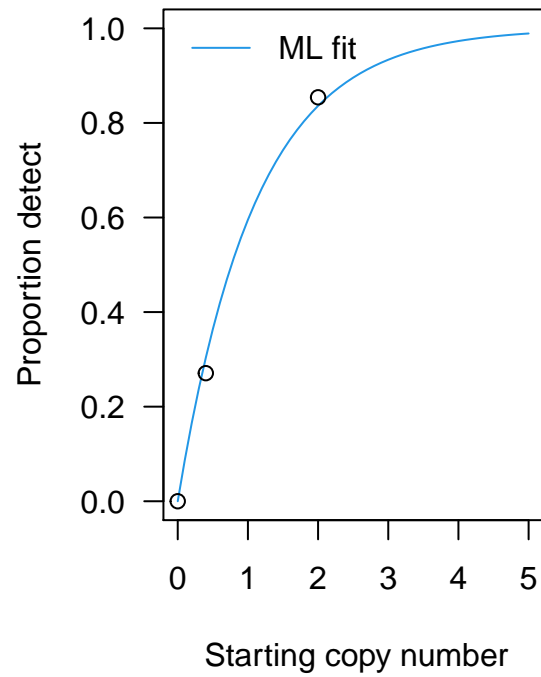


```
##
##
##
##
## GCTM32
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta      1.28    0.19   6.76 1.4e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.07393266 , df= 2 , p-value= 0.9637086
##
## Too few values for Goby
```

SC4

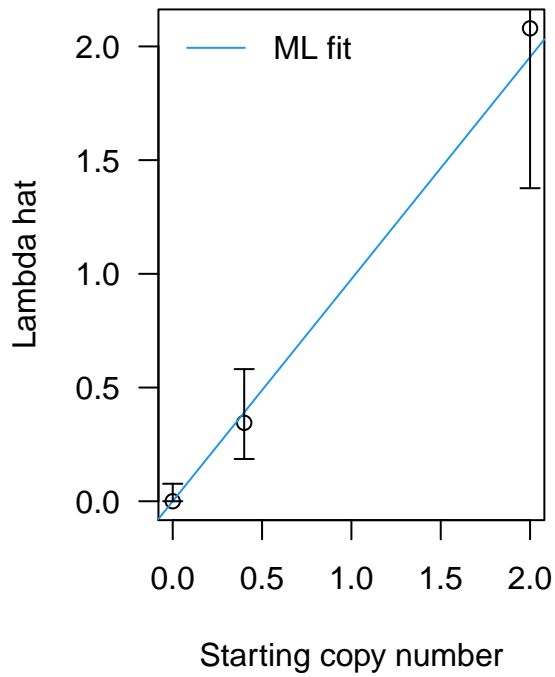


SC4

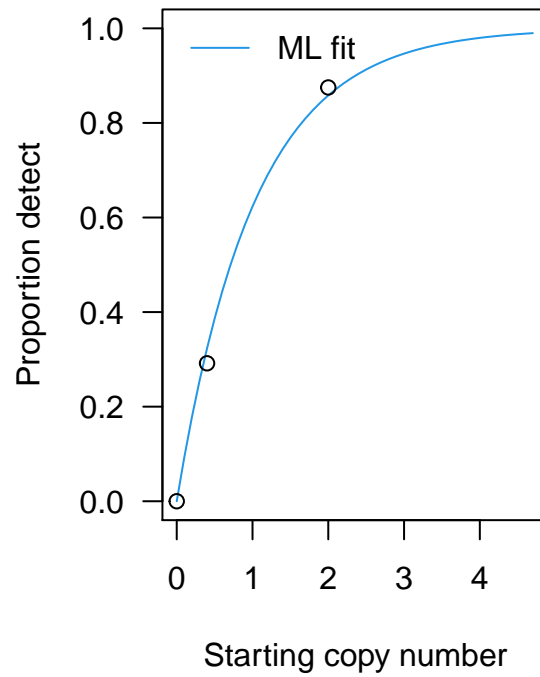


```
##
##
##
##
## SC4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.904   0.136   6.66 2.8e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.365632 , df= 2 , p-value= 0.8329214
```

SC5

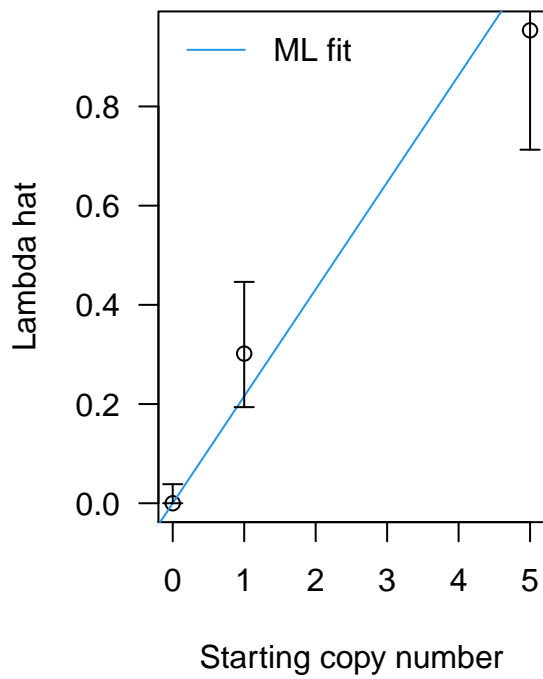


SC5

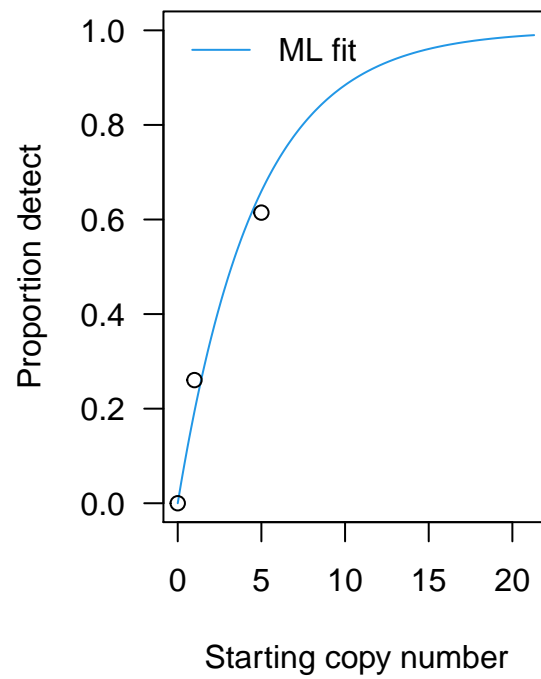


```
##
##
##
##
## SC5
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.977   0.146   6.68 2.3e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.3401212 , df= 2 , p-value= 0.8436137
```

BHC

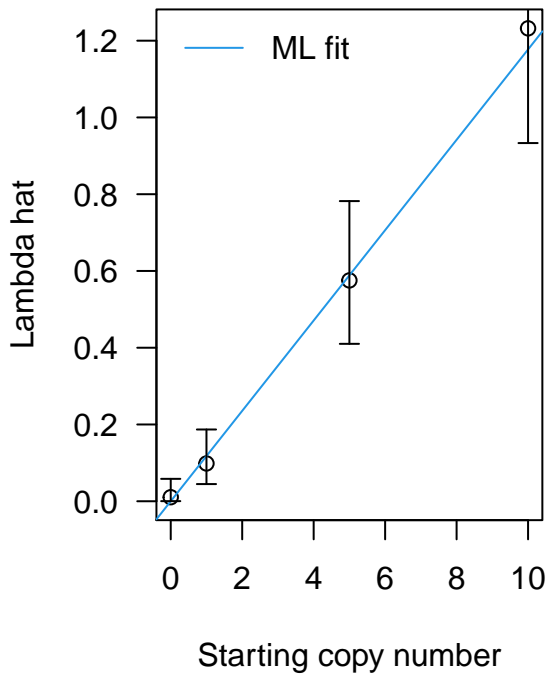


BHC

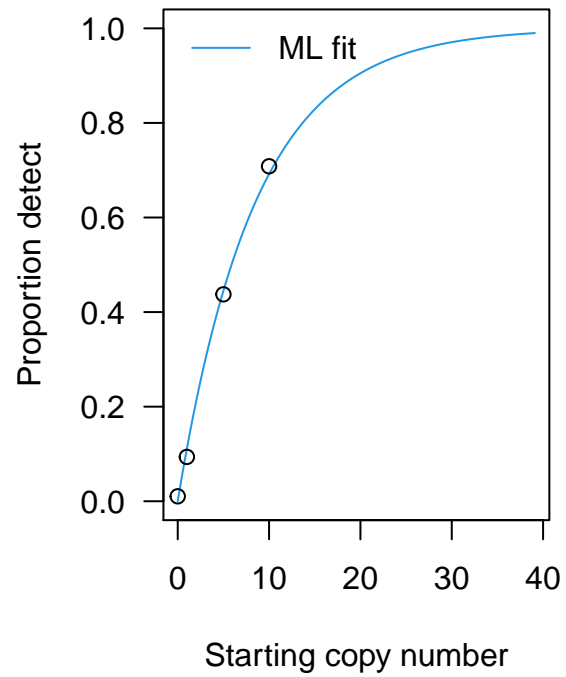


```
##
##
##
##
## BHC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.2156  0.0243   8.86 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 3.368959 , df= 2 , p-value= 0.1855409
```


Dre16s

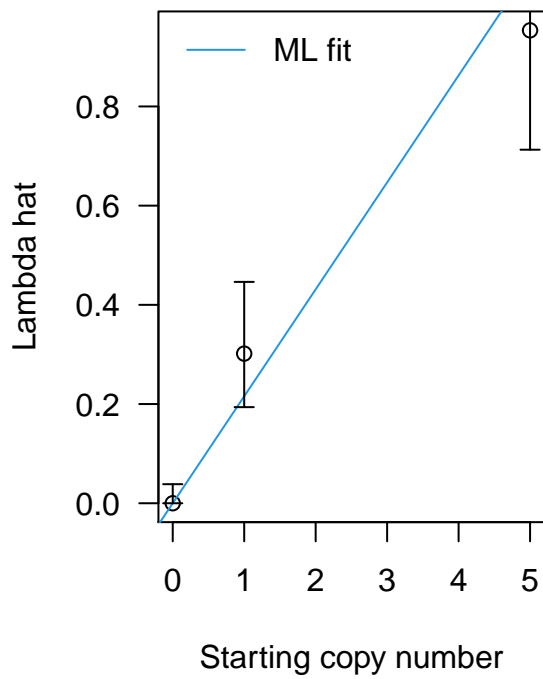


Dre16s

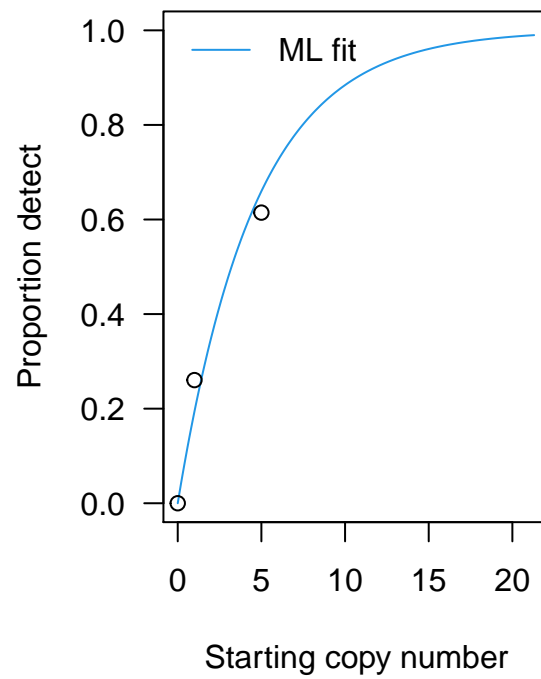


```
##
##
##
##
## Dre16s
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.1177  0.0112   10.5 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.450349 , df= 3 , p-value= 0.929656
##
## Too few values for SS
```

SVC

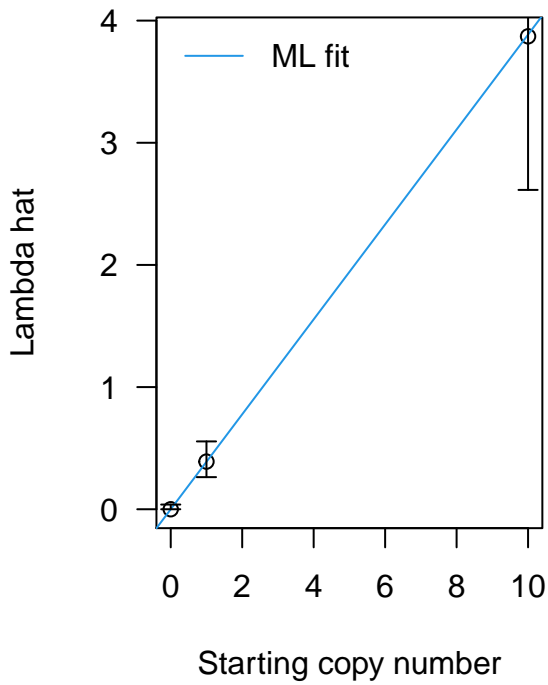


SVC

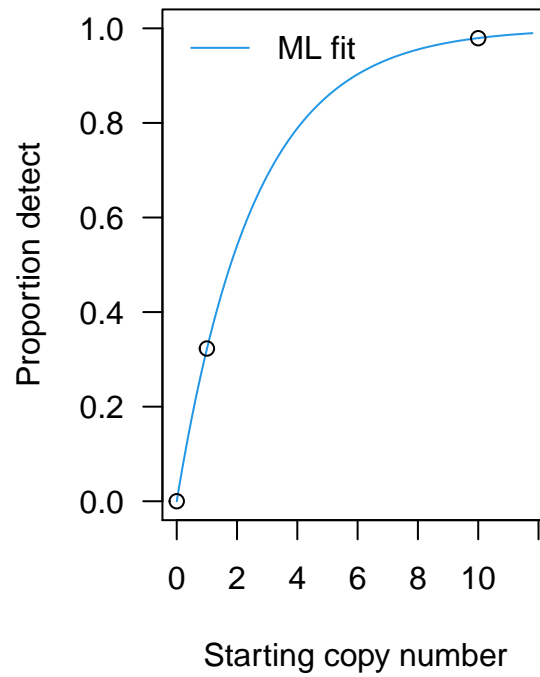


```
##
##
##
##
## SVC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.2156  0.0243   8.86 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 3.368959 , df= 2 , p-value= 0.1855409
```

YPC

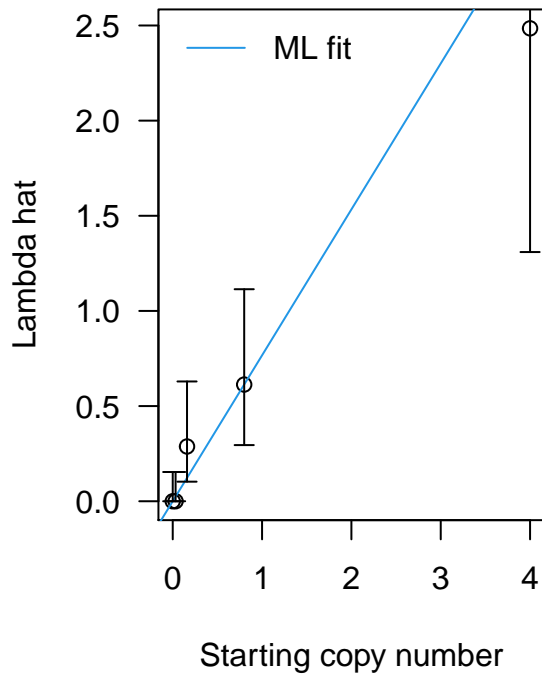


YPC

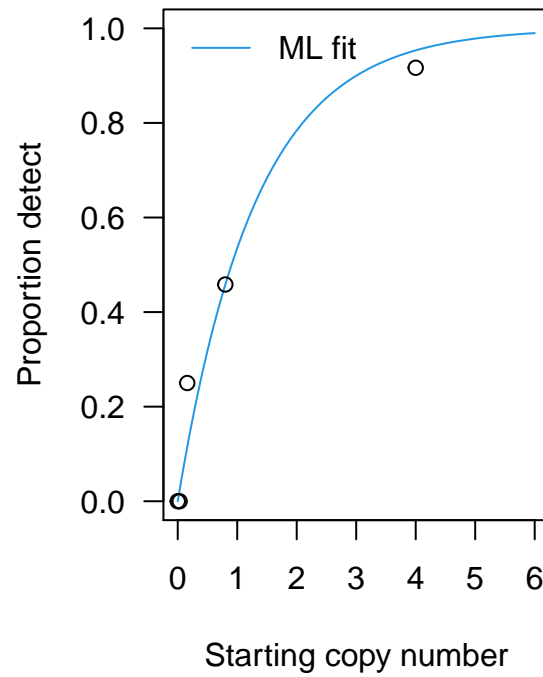


```
##
##
##
##
## YPC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.3885  0.0497   7.81 5.7e-15 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.0008171872 , df= 2 , p-value= 0.9995915
```

eASMO9

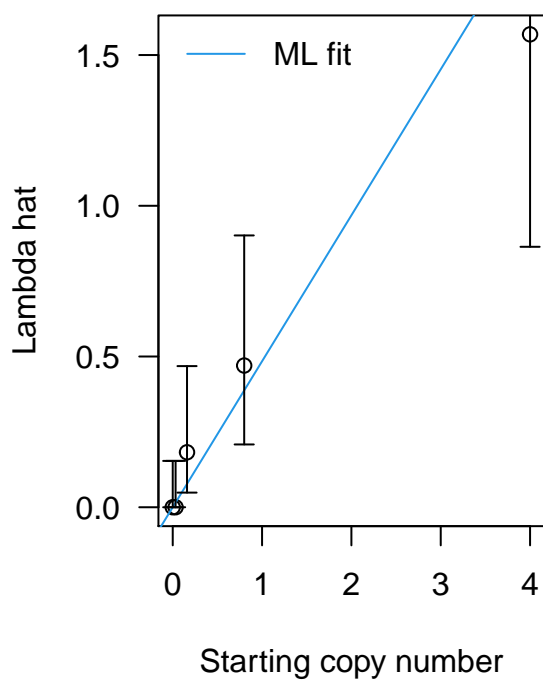


eASMO9

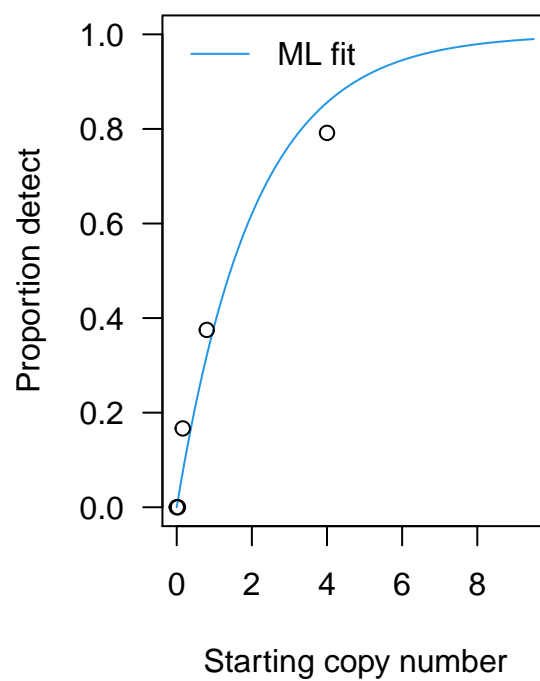


```
##
##
##
##
## eASMO9
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.767    0.147    5.22 1.8e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 5.106559 , df= 4 , p-value= 0.2765376
```

eASTR4

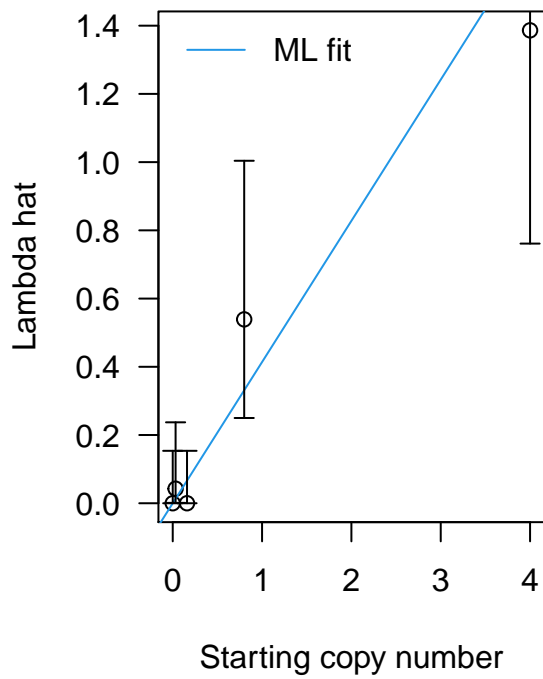


eASTR4

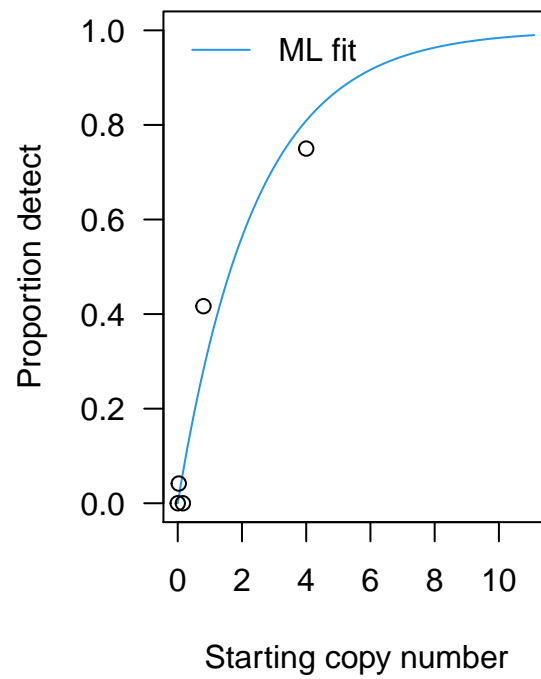


```
##
##
##
##
## eASTR4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.4842  0.0933   5.19 2.1e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 4.017301 , df= 4 , p-value= 0.4036694
```

eFISH1

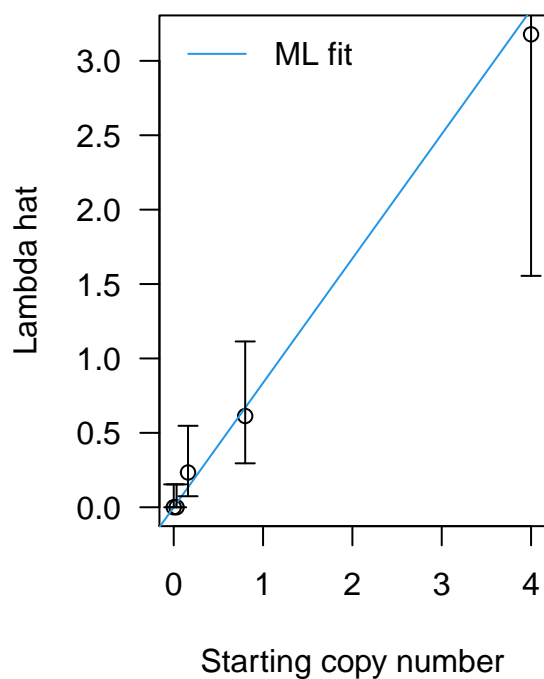


eFISH1

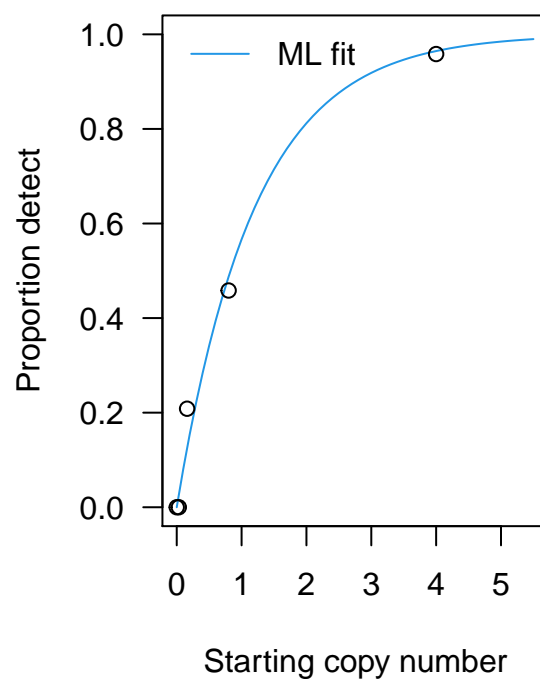


```
##
##
##
##
## eFISH1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.4139  0.0823   5.03 4.9e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.63572 , df= 4 , p-value= 0.1564373
```

eLIP11

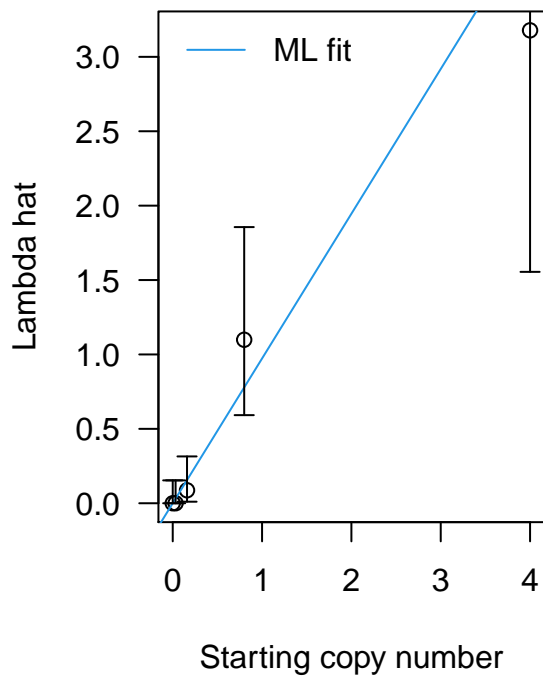


eLIP11

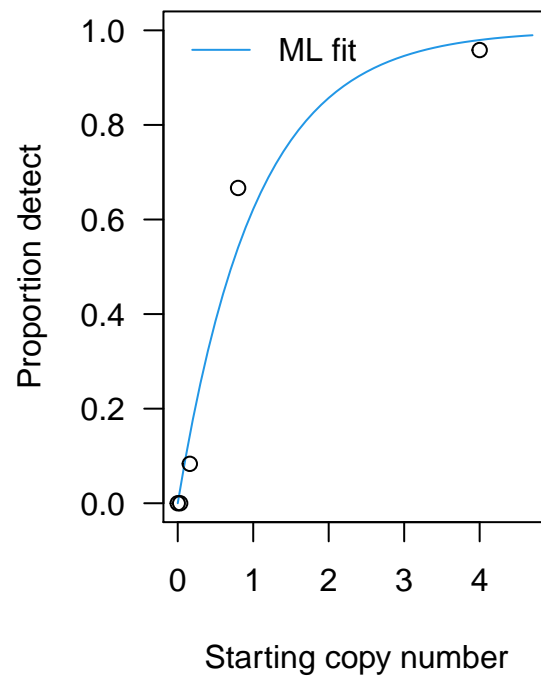


```
##
##
##
##
## eLIP11
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.836   0.166   5.03 4.8e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.691805 , df= 4 , p-value= 0.6106493
```

eMIDO1

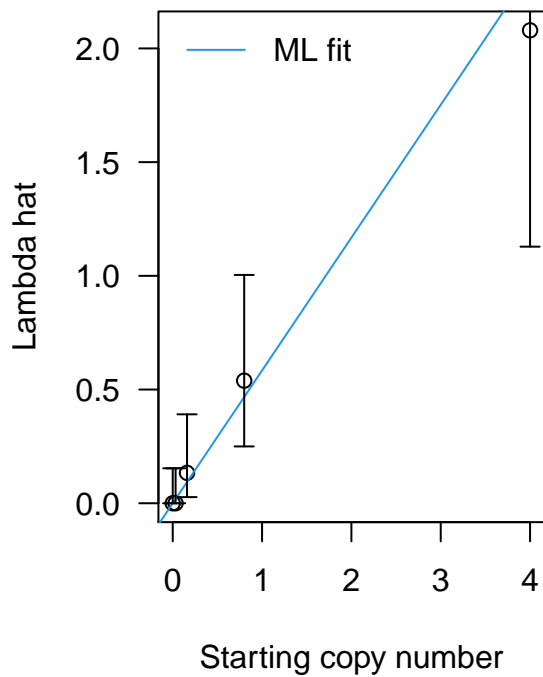


eMIDO1

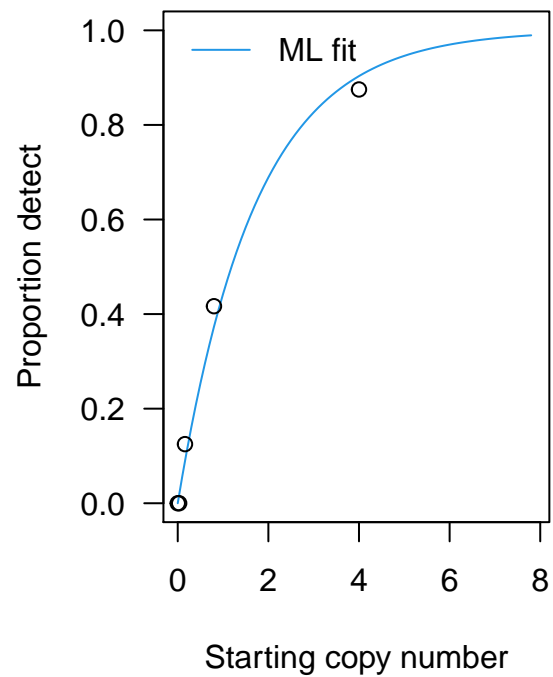


```
##
##
##
##
## eMIDO1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.973   0.196   4.96  7e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 4.310673 , df= 4 , p-value= 0.3655905
```


eMISA2

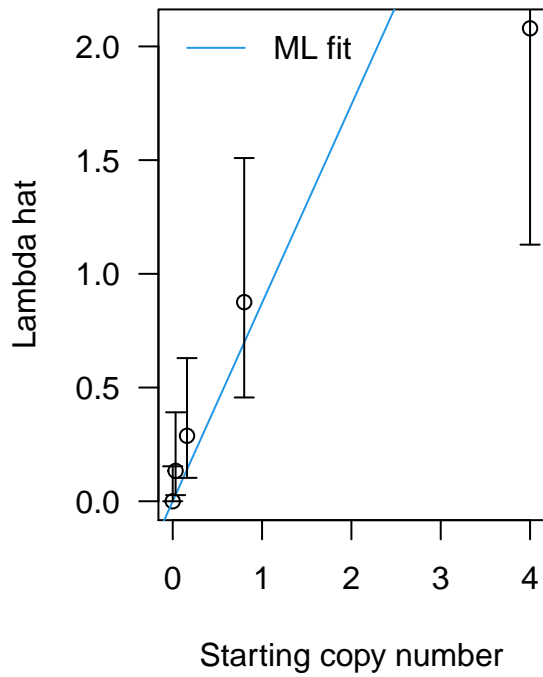


eMISA2

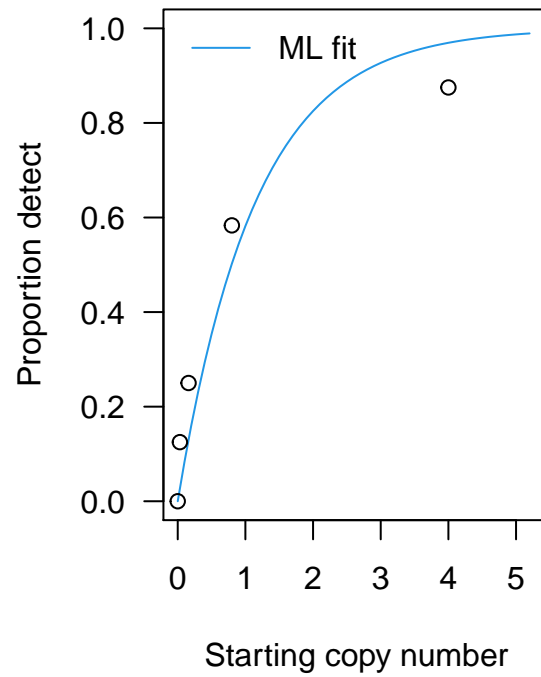


```
##
##
##
##
## eMISA2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.584   0.114   5.14 2.8e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.631095 , df= 4 , p-value= 0.8031928
```

eONKI4

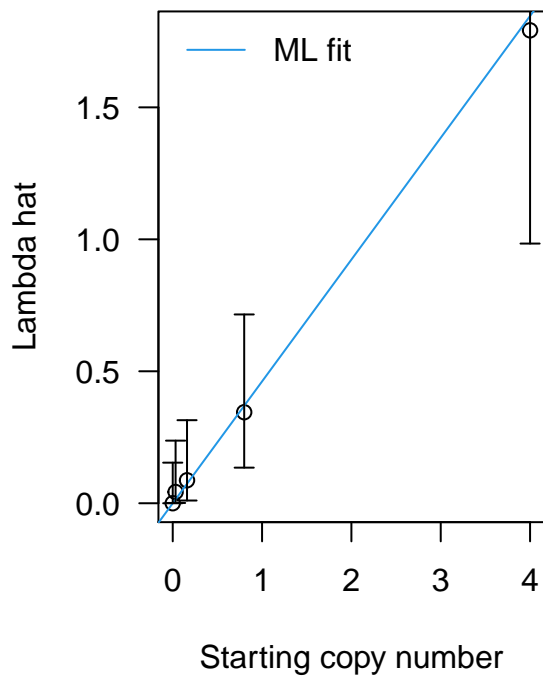


eONKI4

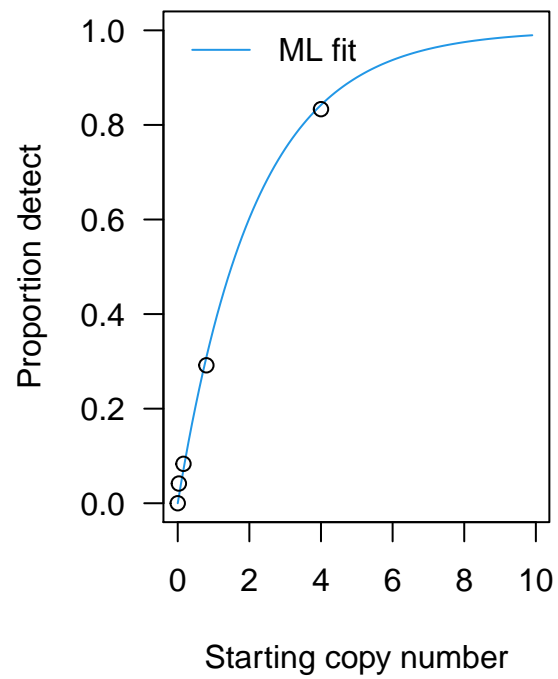


```
##
##
##
##
## eONKI4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.872   0.157   5.54 2.9e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 11.91839 , df= 4 , p-value= 0.01796854
```

eRAAU1

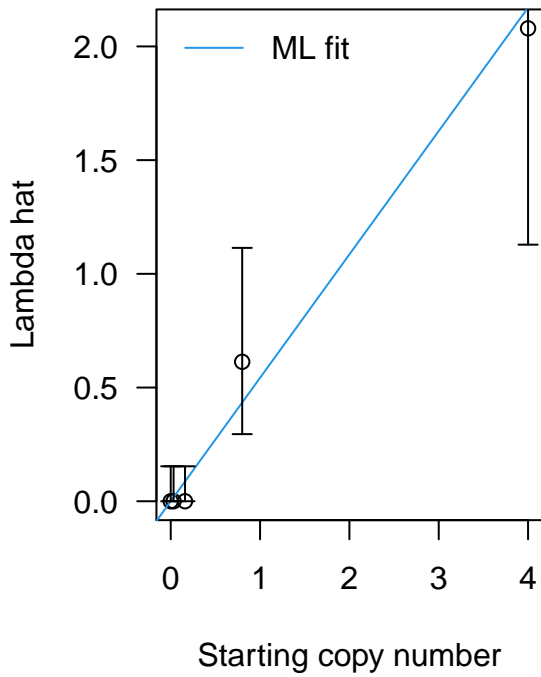


eRAAU1

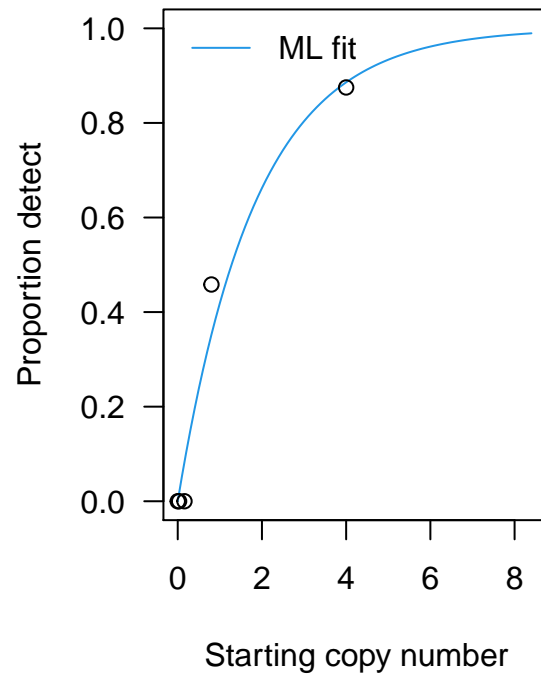


```
##
##
##
##
## eRAAU1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.4616  0.0921   5.01 5.5e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.9086316 , df= 4 , p-value= 0.9233192
```

eRACA2

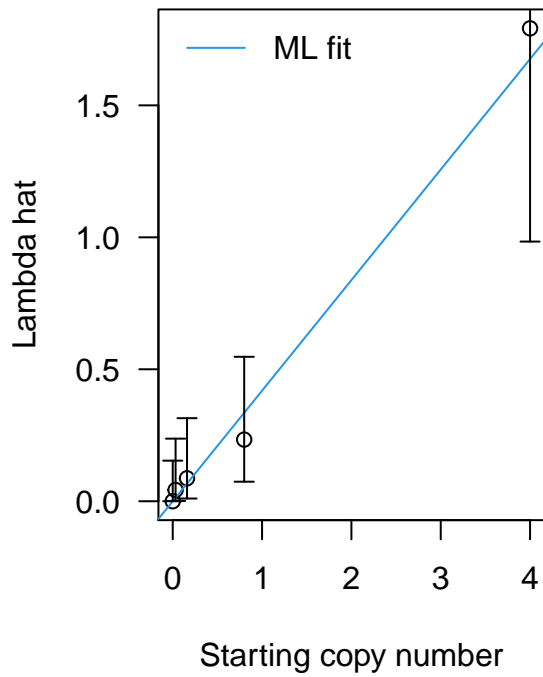


eRACA2

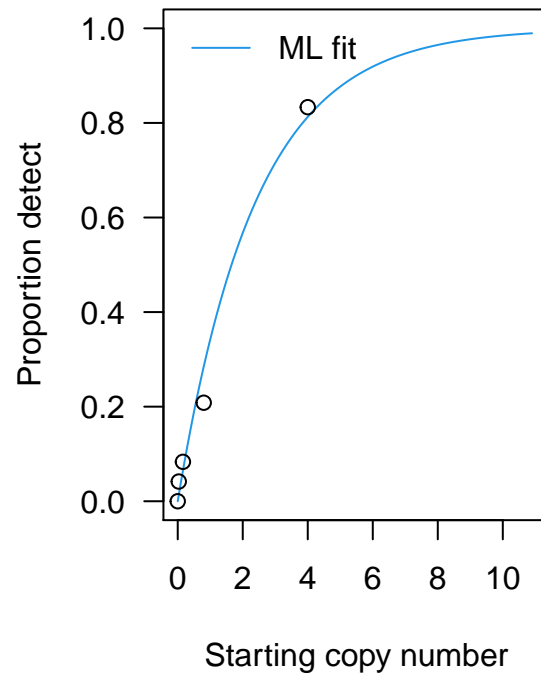


```
##
##
##
##
## eRACA2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.543   0.108   5.02 5.1e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.171783 , df= 4 , p-value= 0.1866815
```

eRALU2

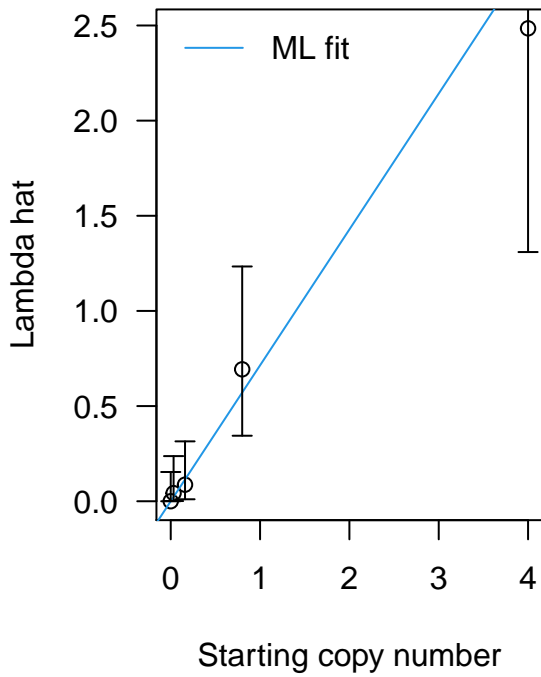


eRALU2

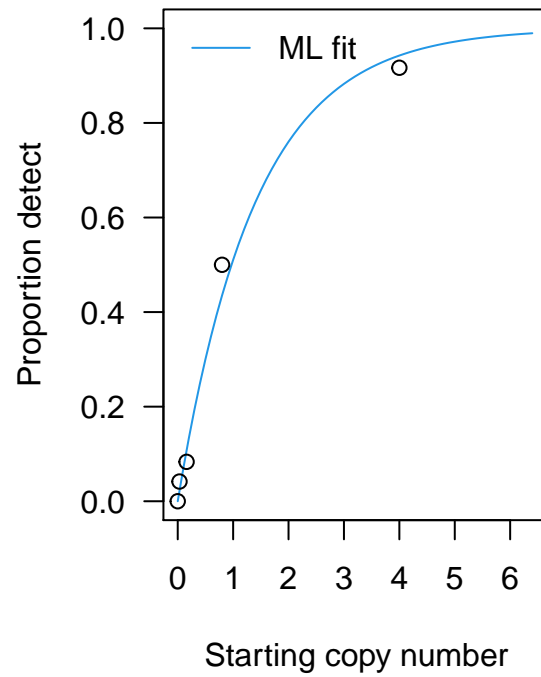


```
##
##
##
##
## eRALU2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta   0.4189  0.0858   4.89  1e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.866509 , df= 4 , p-value= 0.7602944
```

eRAPR2

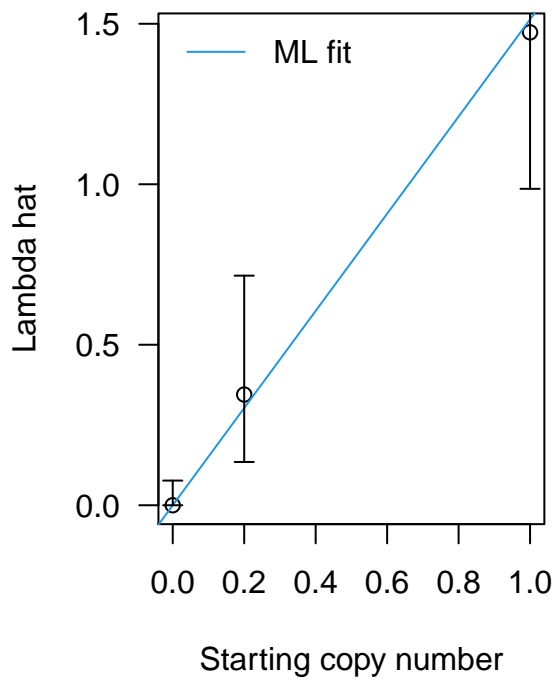


eRAPR2

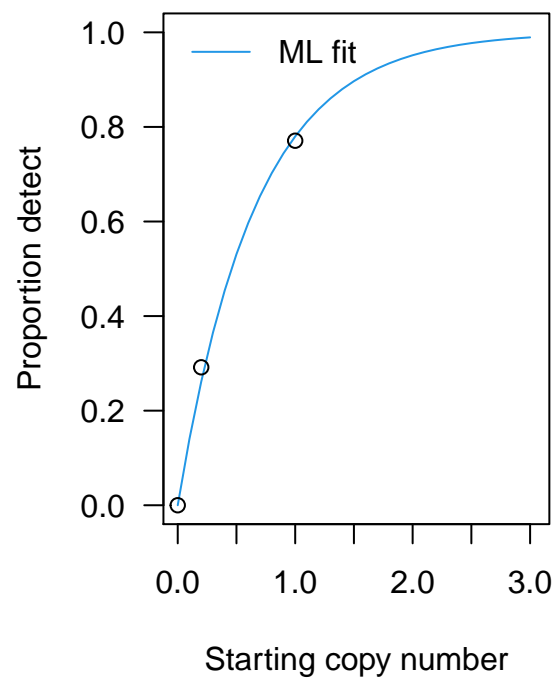


```
##
##
##
##
## eRAPR2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    0.714    0.139    5.13 2.9e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.148873 , df= 4 , p-value= 0.8864424
```

Hno



Hno

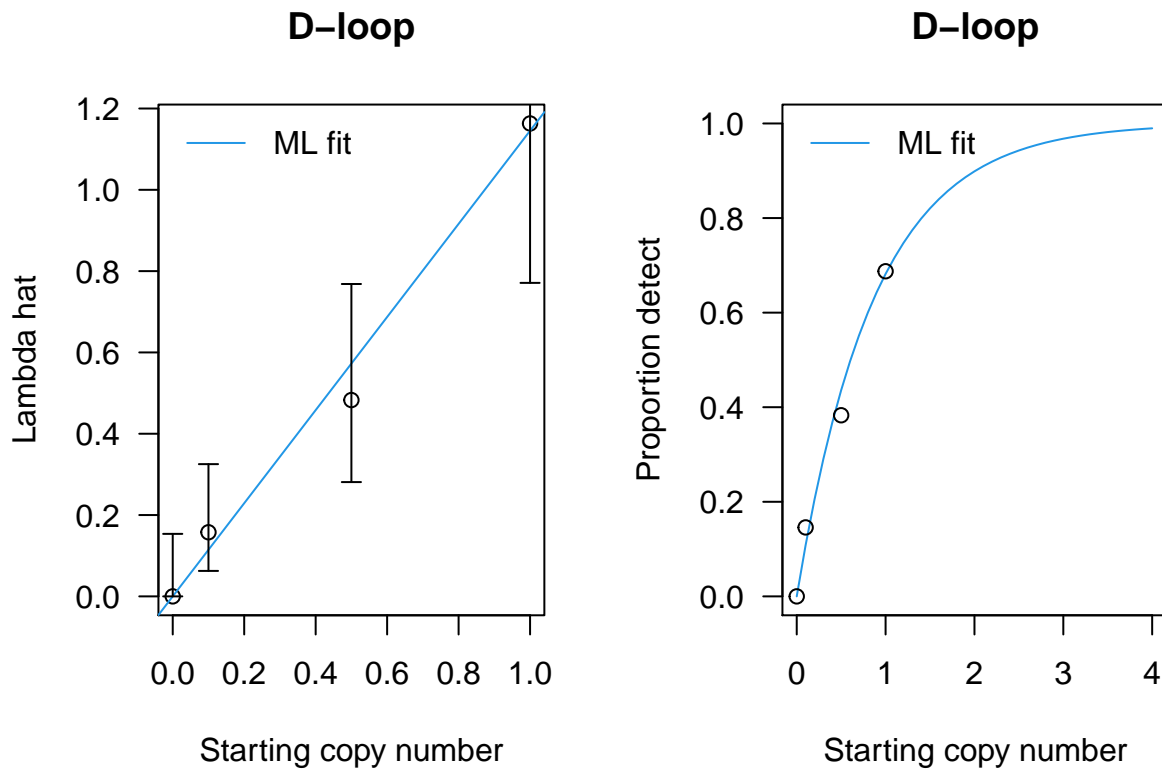


```
##
##
##
##
## Hno
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta    1.514    0.247    6.13 8.6e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.1354573 , df= 2 , p-value= 0.934514
```

Estimate predicted Sq given number detects and technical replicates - no intercept (not shown)

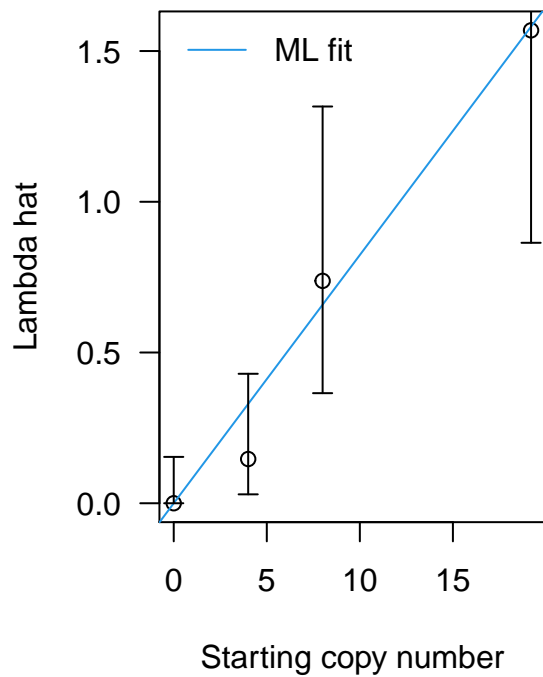
The estimated SQ is easily obtained for given new values of nn0=number of replicates, nd0=number detected and the estimated slope betaS as: $\text{Shat} \leftarrow -(\log((\text{nn0} - \text{nd0})/\text{nn0})) / \text{betaS}[1]$ The standard errors are obtained from the Hessian matrix (or via the function `CalibS0Or.ddLLik()`) .

Estimate Poisson models - intercept model

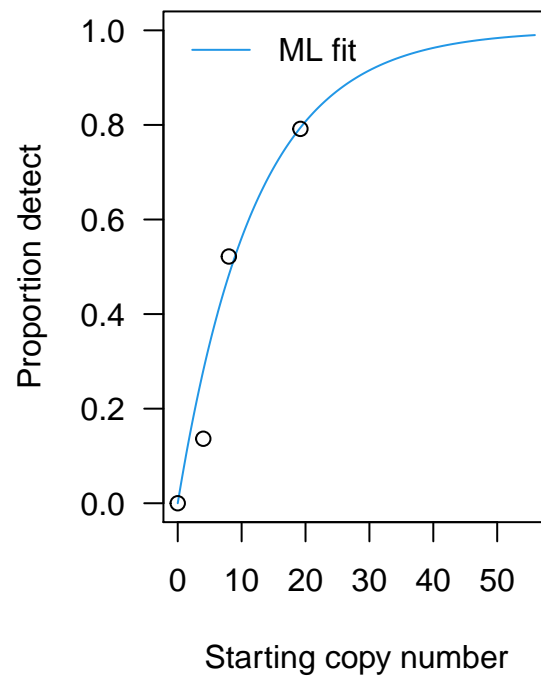


```
##
##
##
##
## D-loop
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 3.33e-09 5.31e-02   0.00    1
## beta  1.15e+00 2.04e-01   5.61 2e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.192445 , df= 2 , p-value= 0.5508887
```

AC1_CID_1

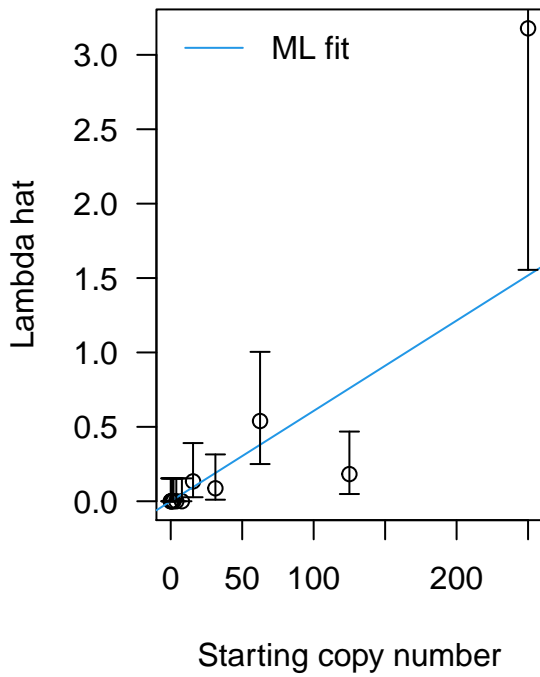


AC1_CID_1

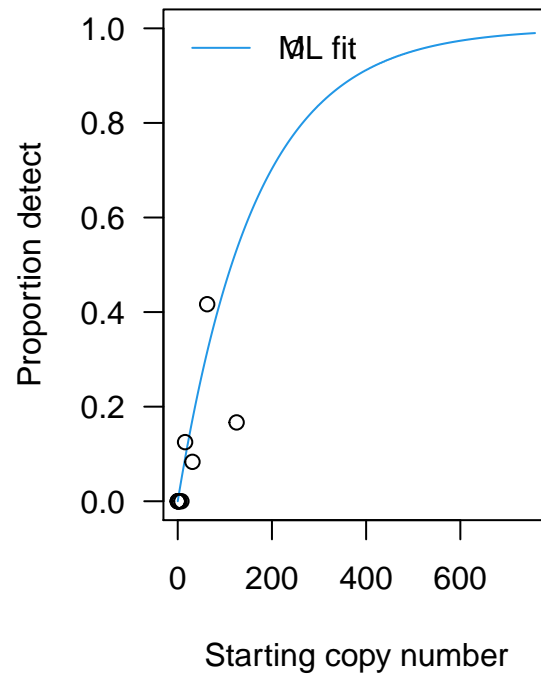


```
##
##
##
##
## AC1_CID_1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.96e-08 2.49e-01    0.00 1.0000
## beta  8.23e-02 2.90e-02    2.84 0.0045 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.757685 , df= 2 , p-value= 0.2518699
##
## Too few values for CID
```

MYPI-6

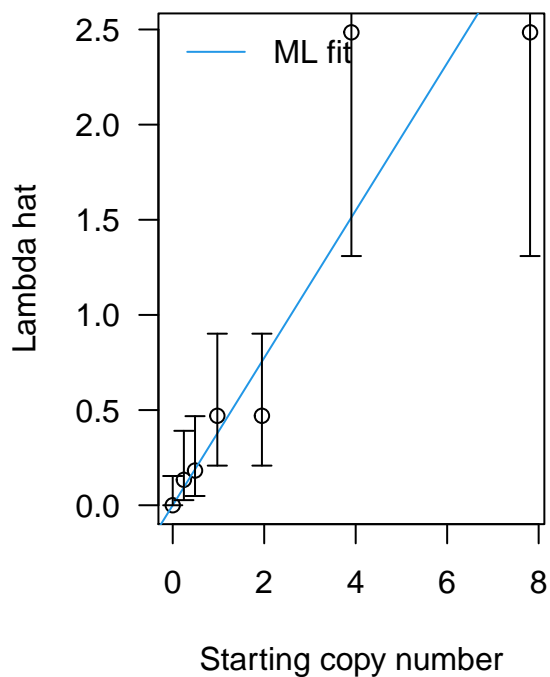


MYPI-6

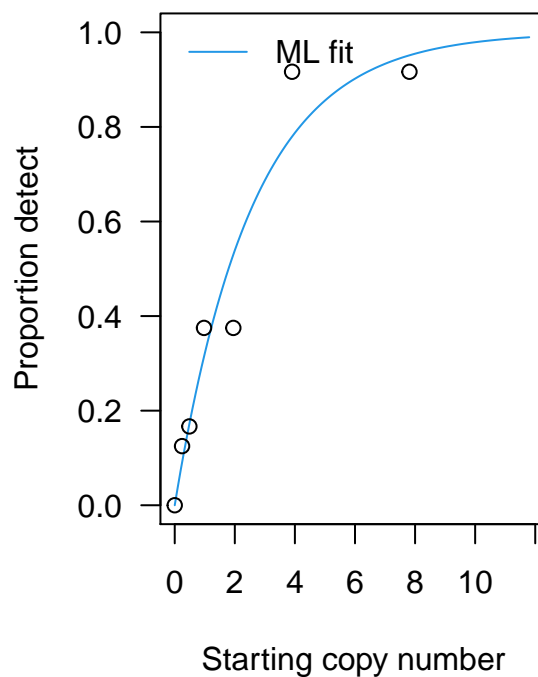


```
##
##
##
##
## MYPI-6
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 2.78e-11 6.00e-02   0.00    1
## beta  6.07e-03 1.27e-03   4.78 1.7e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 27.35988 , df= 10 , p-value= 0.002283695
```

MYPI-6t

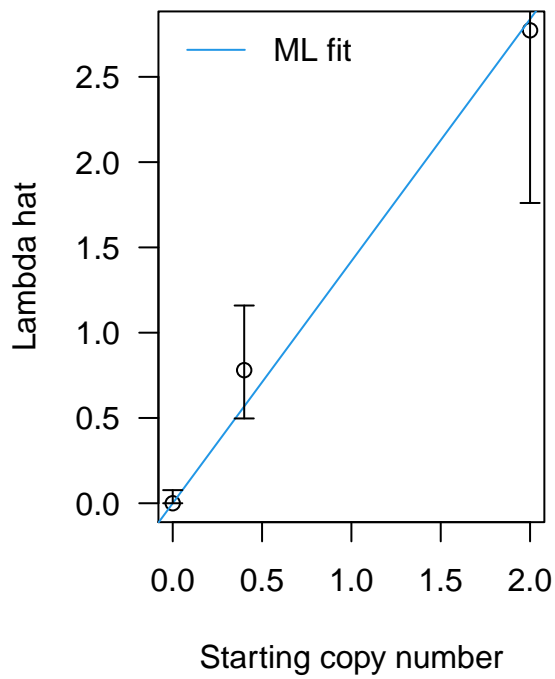


MYPI-6t

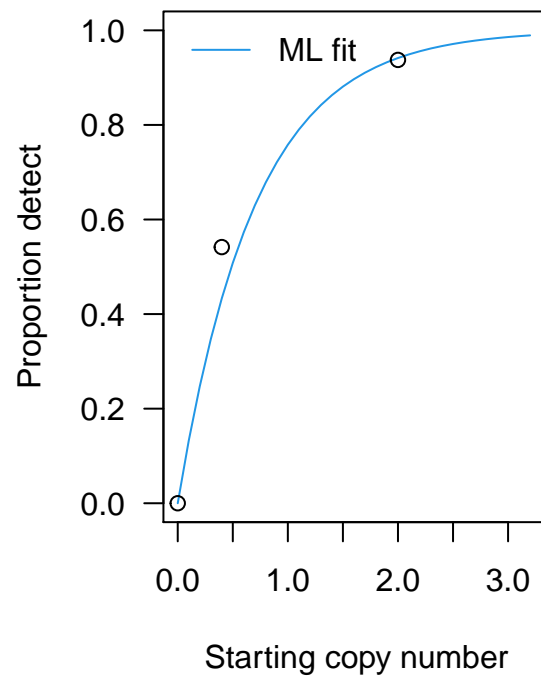


```
##
##
##
##
## MYPI-6t
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.86e-08 5.46e-02   0.00    1
## beta  3.87e-01 6.69e-02   5.79 7.1e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.803394 , df= 5 , p-value= 0.2356775
##
## Too few values for  NZMS
## Too few values for  SAFO
## Too few values for  Elod-2-NU
## Too few values for  MYPI
```

AC1

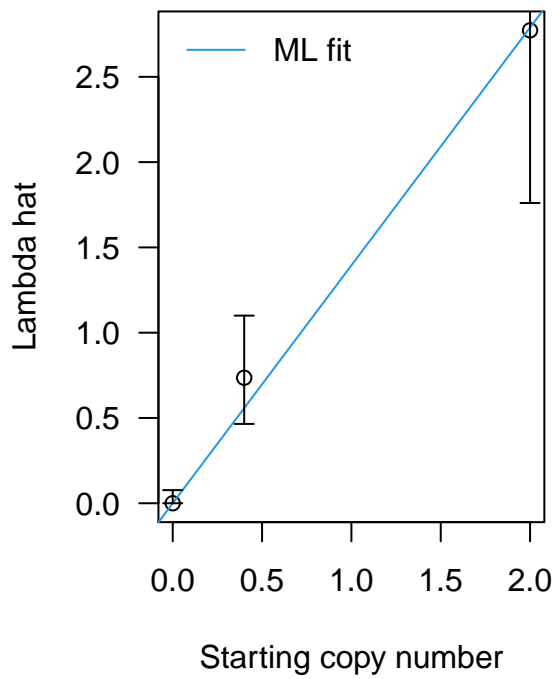


AC1

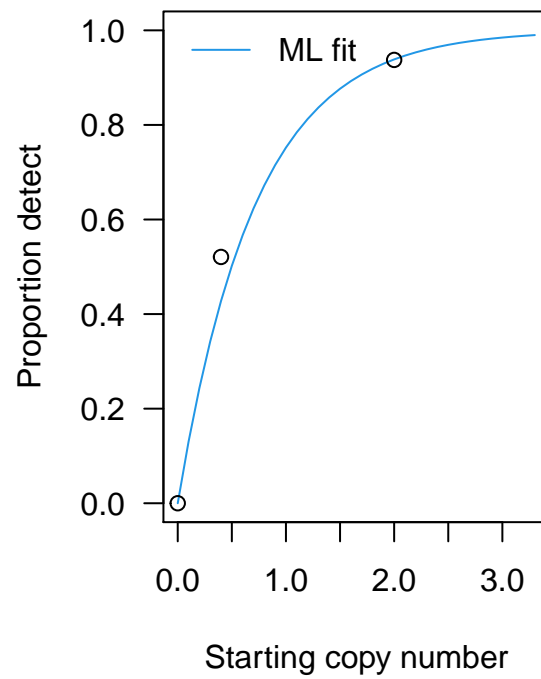


```
##
##
##
##
## AC1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 5.57e-10 2.02e-01    0.00 1.00000
## beta  1.42e+00 3.69e-01    3.84 0.00012 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.288843 , df= 1 , p-value= 0.1303071
```

AC3

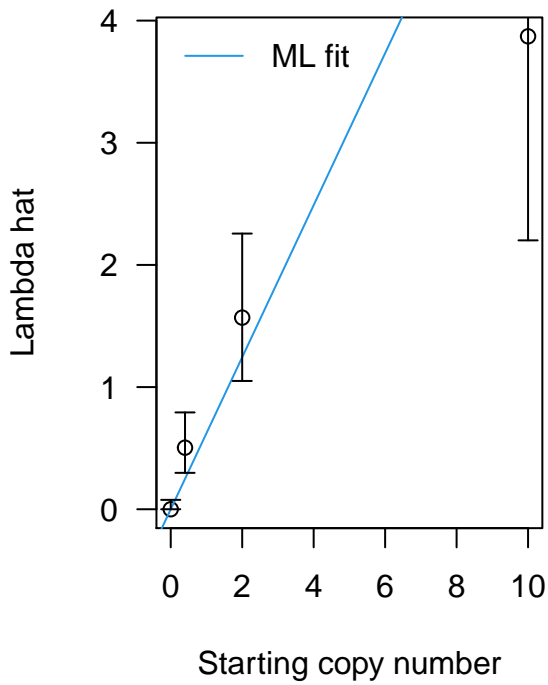


AC3

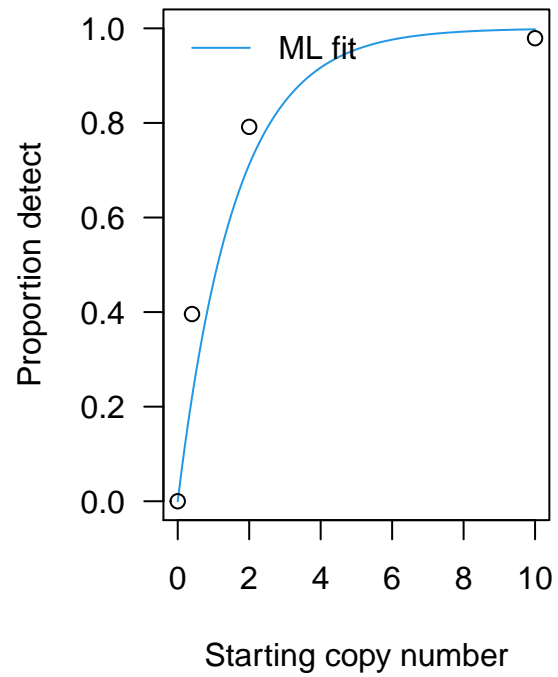


```
##
##
##
##
## AC3
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.83e-08 2.00e-01    0.00 1.00000
## beta  1.39e+00 3.60e-01    3.88 0.00011 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.685805 , df= 1 , p-value= 0.1941549
```

BH1

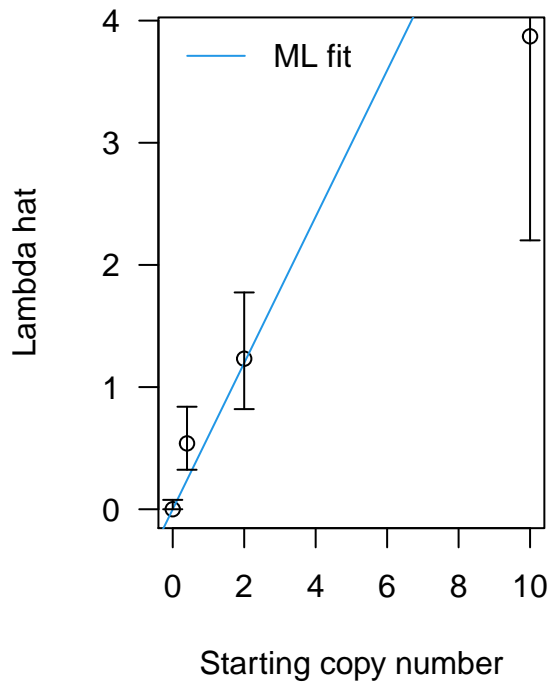


BH1

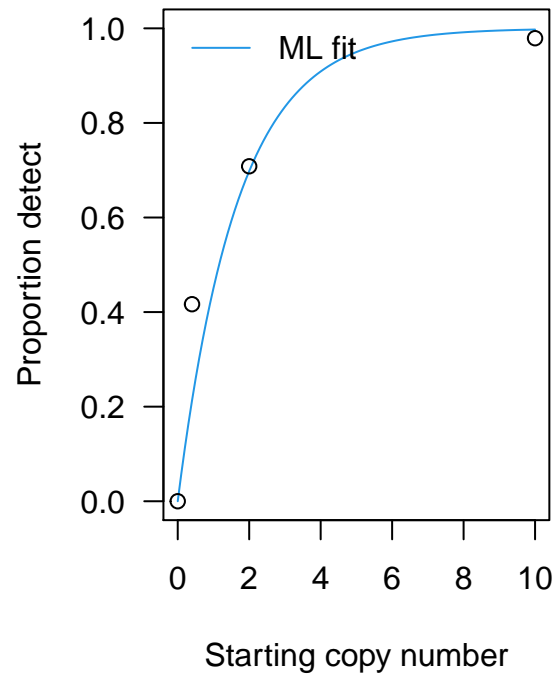


```
##
##
##
##
## BH1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 9.77e-09 8.58e-02   0.00    1
## beta  6.22e-01 1.29e-01   4.82 1.4e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 11.96976 , df= 2 , p-value= 0.002516509
```

BH2

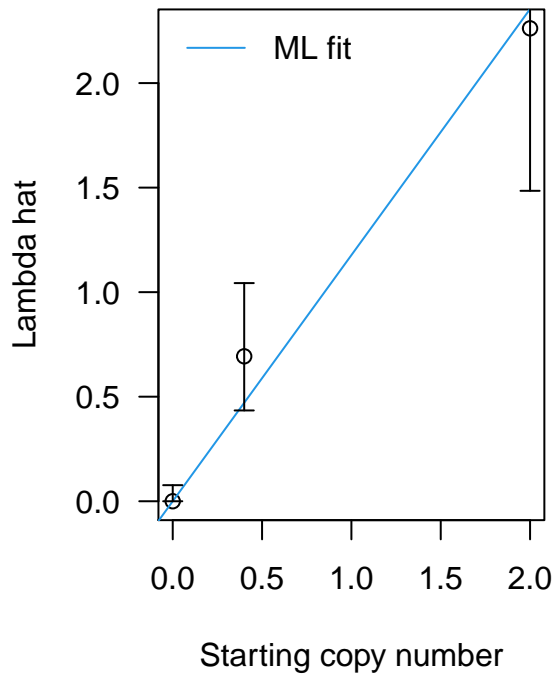


BH2

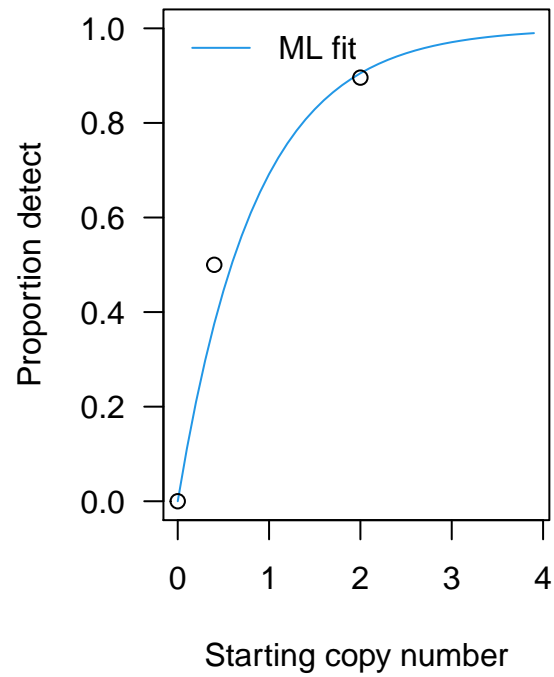


```
##
##
##
##
## BH2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 7.41e-09 8.18e-02   0.0      1
## beta  5.99e-01 1.27e-01   4.7 2.6e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 12.58791 , df= 2 , p-value= 0.001847435
```


GCTM10

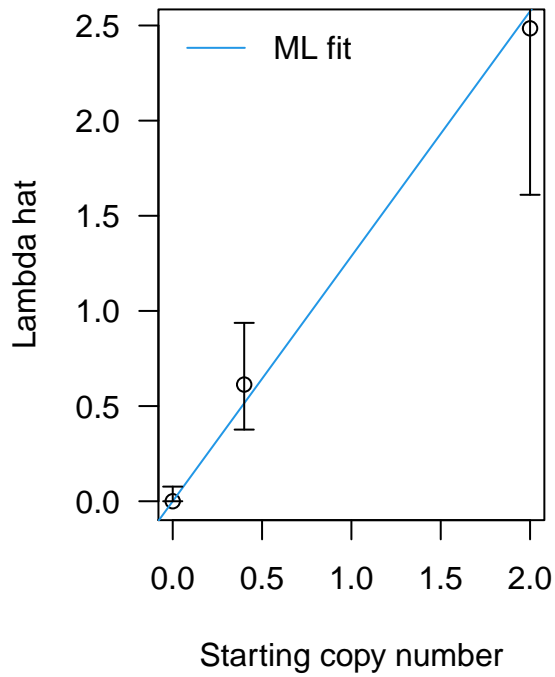


GCTM10

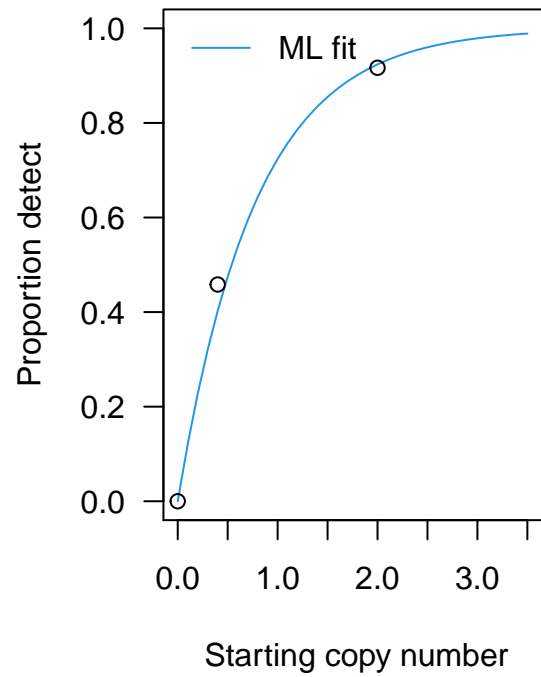


```
##
##
##
##
## GCTM10
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 3.46e-09 1.65e-01   0.00    1
## beta  1.18e+00 2.86e-01   4.11 4e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 3.116952 , df= 1 , p-value= 0.07748159
```

GCTM22

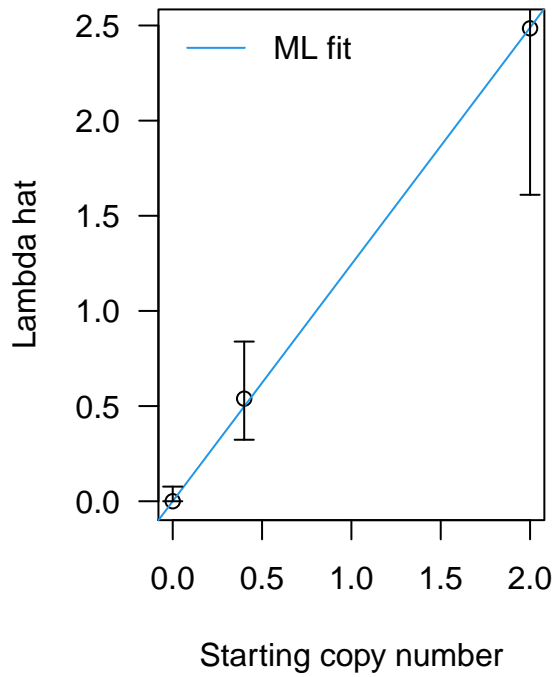


GCTM22

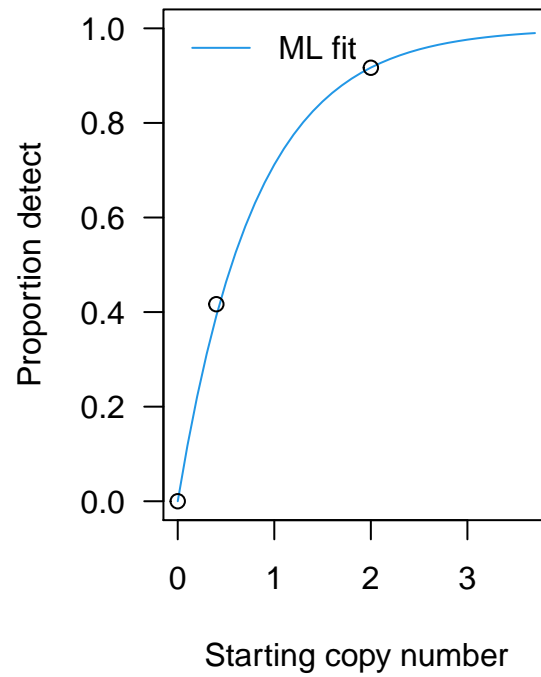


```
##
##
##
##
## GCTM22
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.38e-08 1.88e-01   0.00      1
## beta  1.29e+00 3.23e-01   3.99 6.7e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.6489803 , df= 1 , p-value= 0.4204775
```

GCTM32

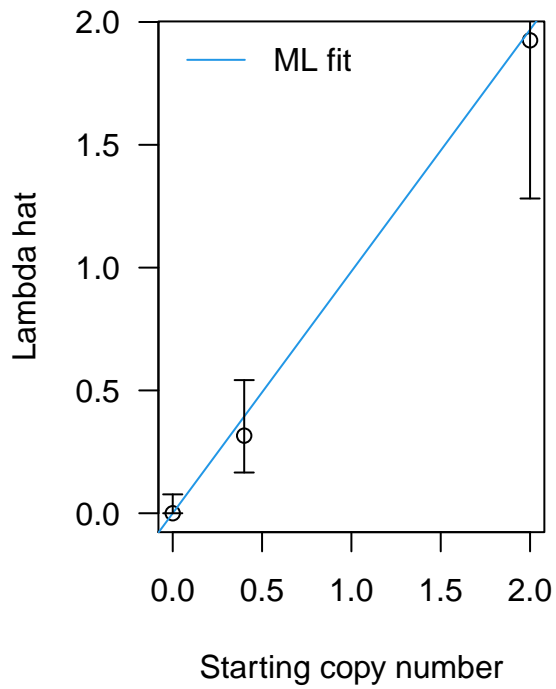


GCTM32

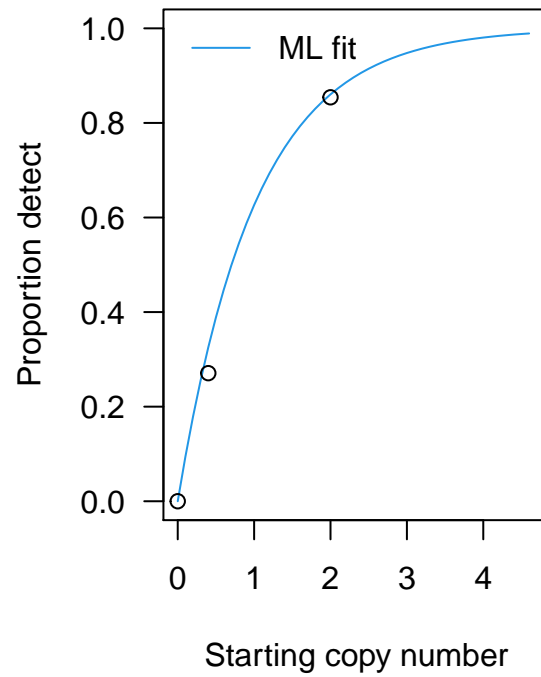


```
##
##
##
##
## GCTM32
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.91e-09 1.85e-01   0.00     1
## beta  1.24e+00 3.08e-01   4.04 5.3e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.1206682 , df= 1 , p-value= 0.7283109
##
## Too few values for Goby
```

SC4

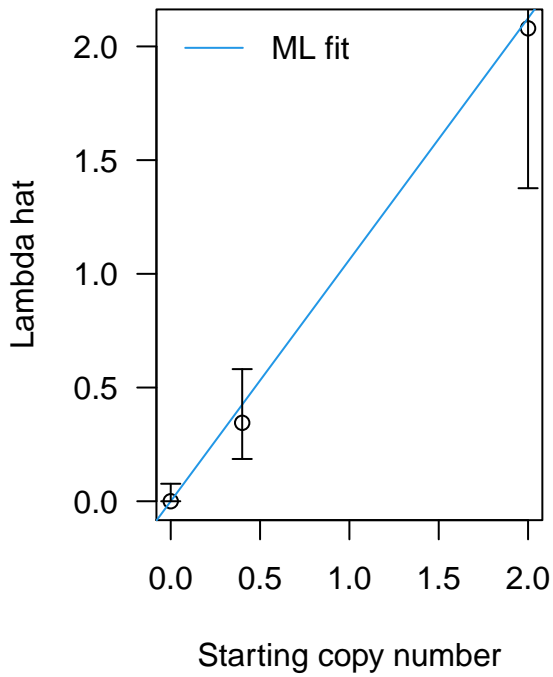


SC4

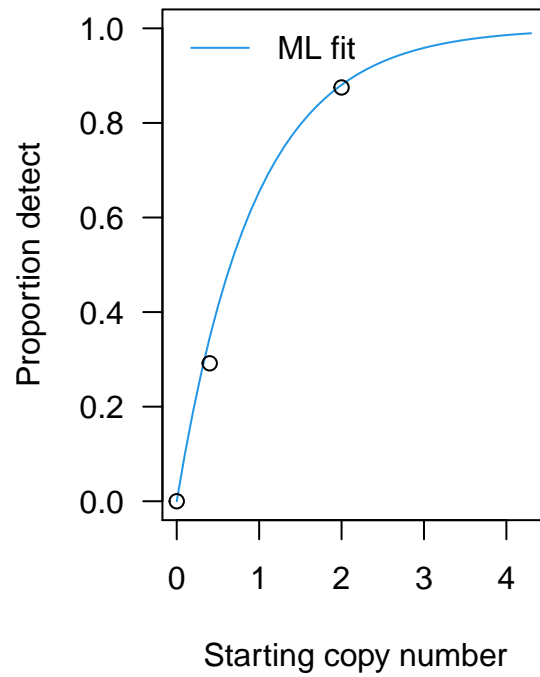


```
##
##
##
##
## SC4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 2.54e-10 1.64e-01   0.00    1
## beta  9.84e-01 2.35e-01   4.19 2.8e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.6901617 , df= 1 , p-value= 0.4061094
```

SC5

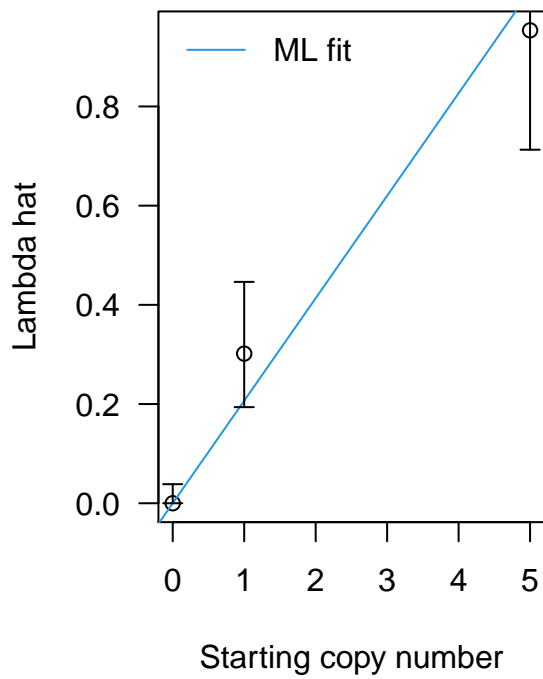


SC5

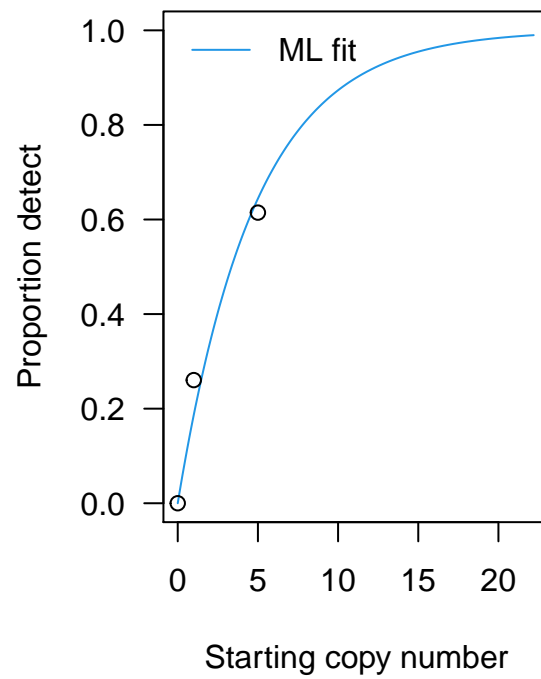


```
##
##
##
##
## SC5
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.46e-09 1.73e-01   0.00     1
## beta  1.06e+00 2.56e-01   4.15 3.3e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.6602861 , df= 1 , p-value= 0.416459
```

BHC

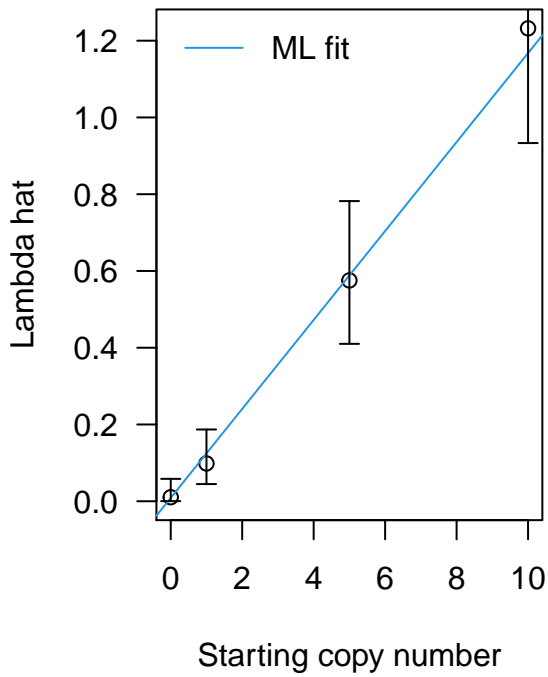


BHC

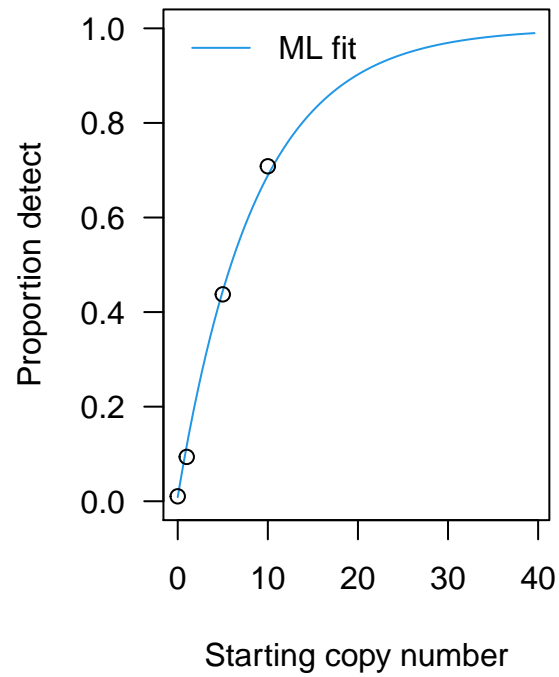


```
##
##
##
##
## BHC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 6.31e-09 6.26e-02   0.00    1
## beta  2.07e-01 3.66e-02   5.64 1.7e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 3.509393 , df= 1 , p-value= 0.06102183
```

Dre16s

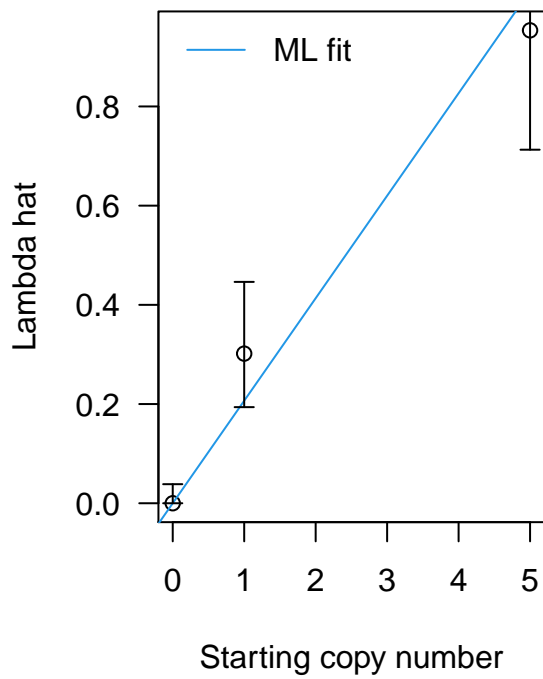


Dre16s

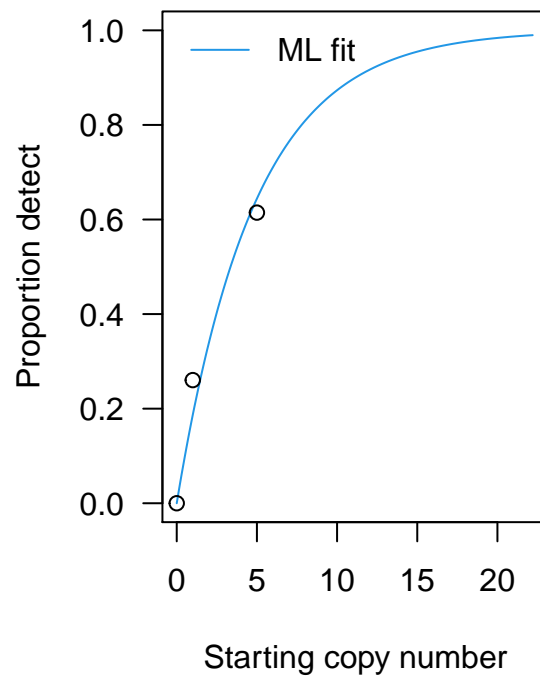


```
##
##
##
##
## Dre16s
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha  0.00880 0.00861   1.02   0.31
## beta   0.11590 0.01133  10.23 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.7640605 , df= 2 , p-value= 0.6824744
##
## Too few values for SS
```

SVC

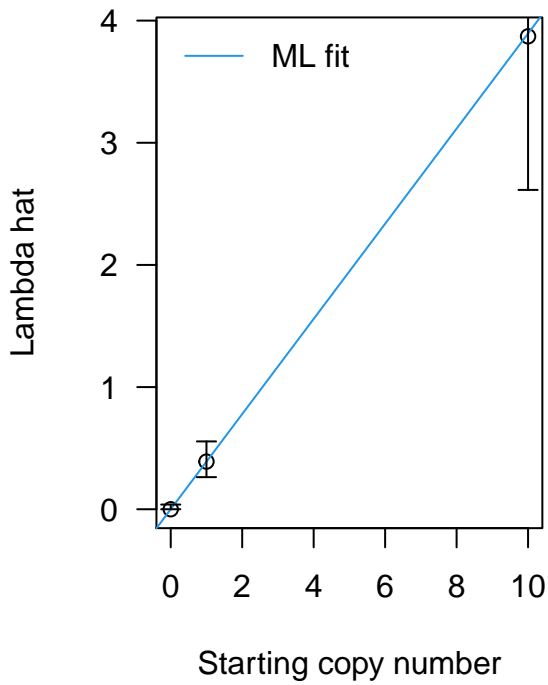


SVC

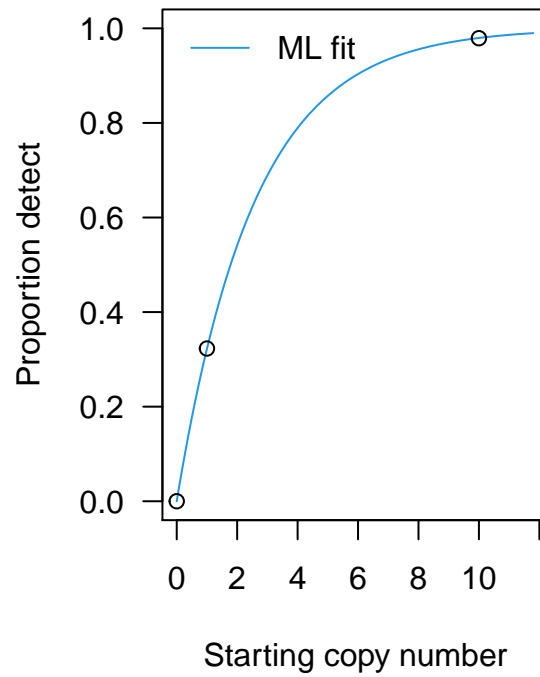


```
##
##
##
##
## SVC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 6.31e-09 6.26e-02   0.00    1
## beta  2.07e-01 3.66e-02   5.64 1.7e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 3.509393 , df= 1 , p-value= 0.06102183
```


YPC

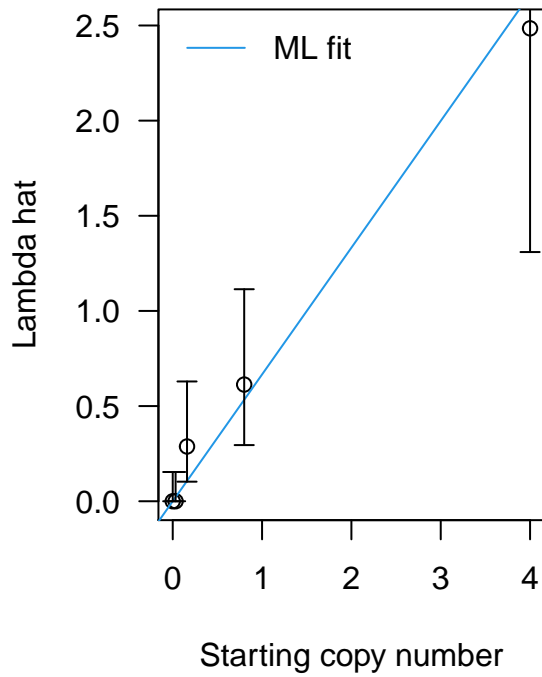


YPC

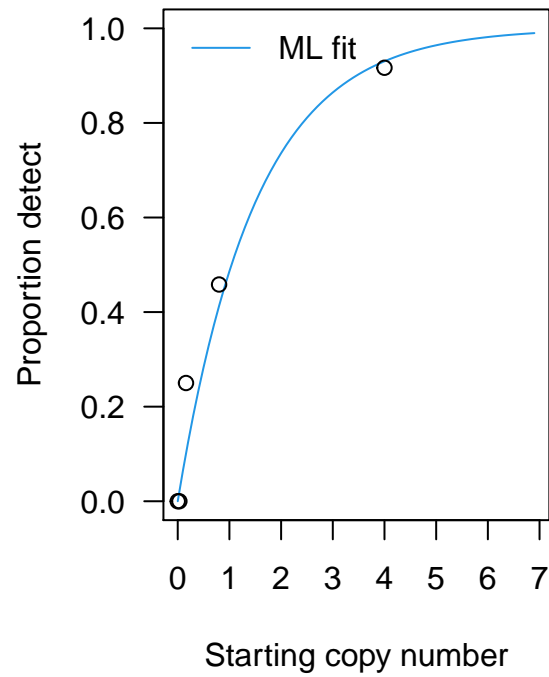


```
##
##
##
##
## YPC
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 4.99e-09 1.11e-01   0.00    1
## beta  3.89e-01 7.91e-02   4.92 8.5e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.001161691 , df= 1 , p-value= 0.9728105
```

eASMO9

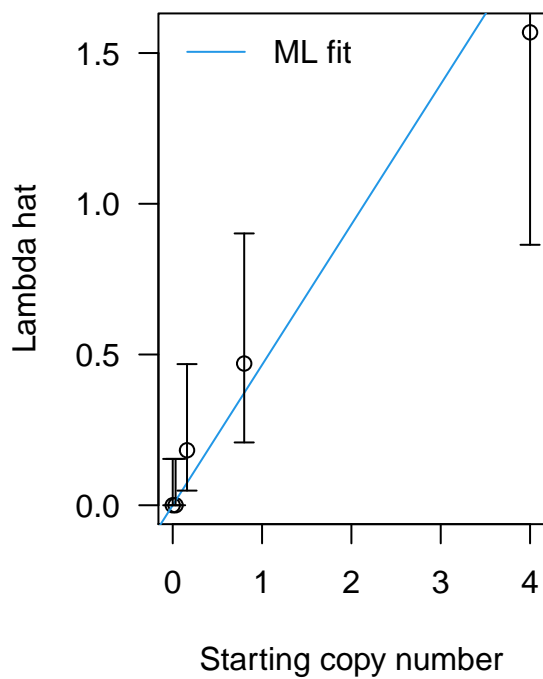


eASMO9

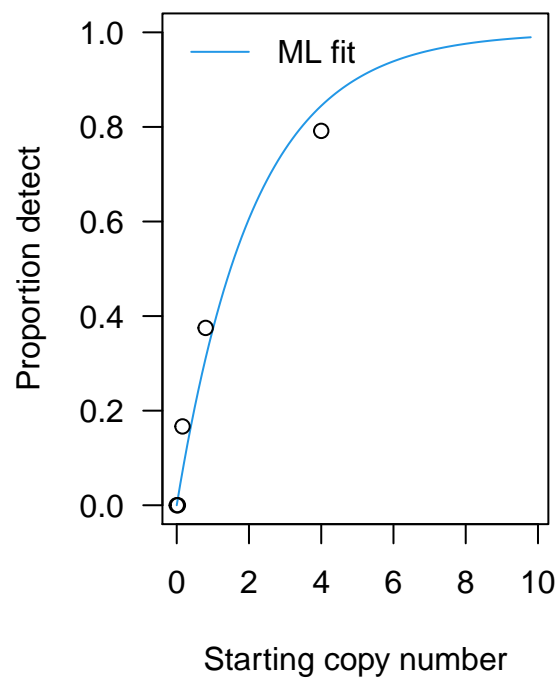


```
##
##
##
##
## eASMO9
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 2.90e-09 5.40e-02   0.00    1
## beta  6.66e-01 1.58e-01   4.21 2.6e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 5.634289 , df= 3 , p-value= 0.1308237
```

eASTR4

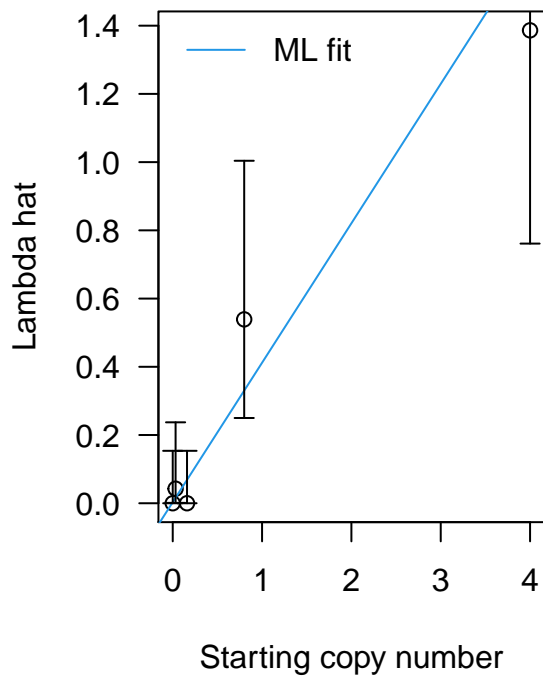


eASTR4

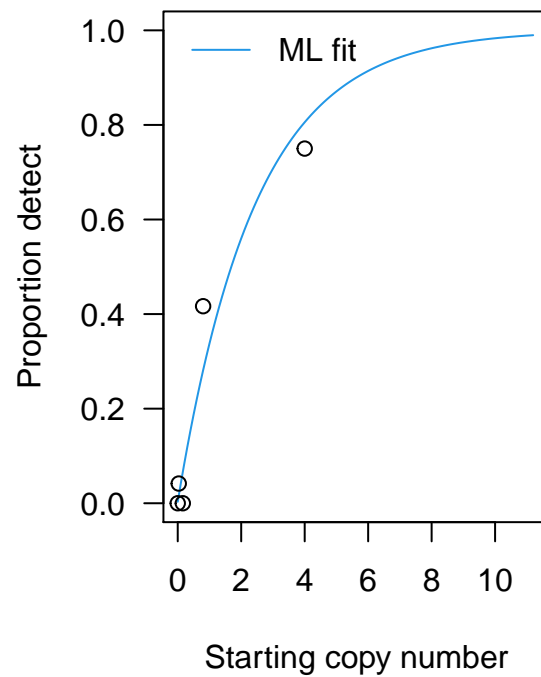


```
##
##
##
##
## eASTR4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 2.64e-09 4.38e-02   0.00     1
## beta  4.66e-01 1.10e-01   4.25 2.2e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 4.058386 , df= 3 , p-value= 0.2552281
```

eFISH1

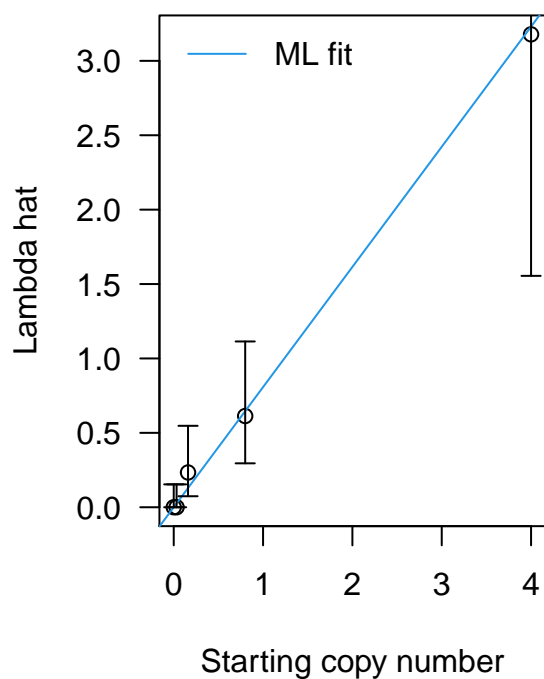


eFISH1

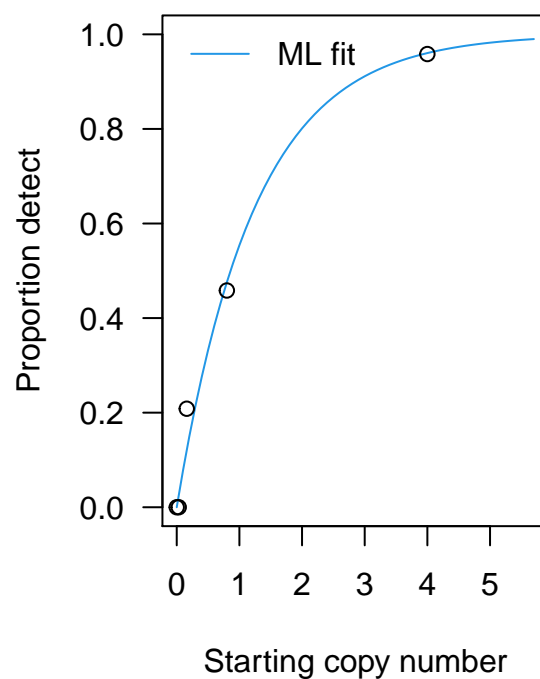


```
##
##
##
##
## eFISH1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha  0.00316 0.01669   0.19   0.85
## beta   0.40873 0.08563   4.77 1.8e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.595206 , df= 3 , p-value= 0.08598249
```

eLIP11

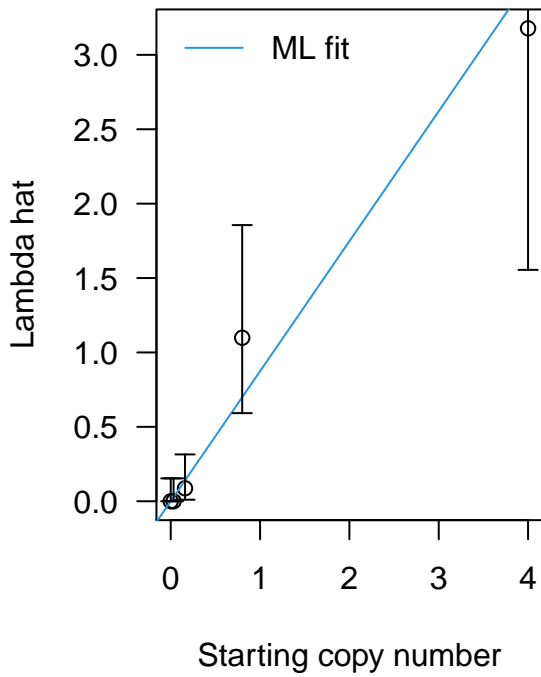


eLIP11

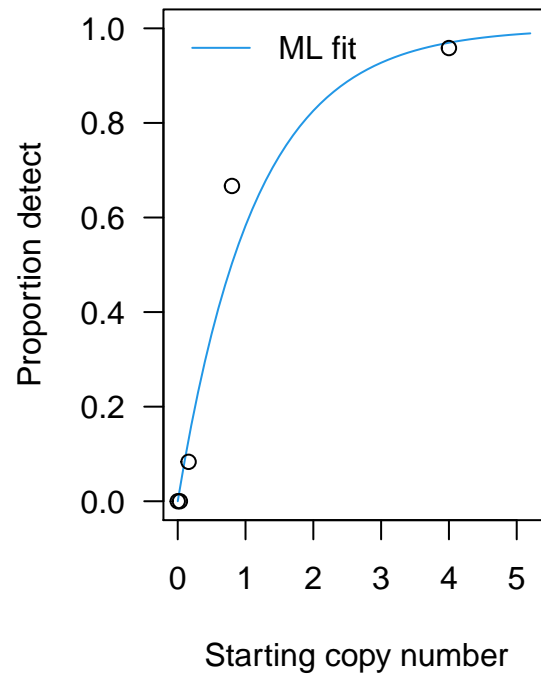


```
##
##
##
##
## eLIP11
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.03e-09 7.17e-02   0.00    1
## beta  8.08e-01 2.05e-01   3.93 8.4e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 2.721685 , df= 3 , p-value= 0.4365547
```

eMIDO1

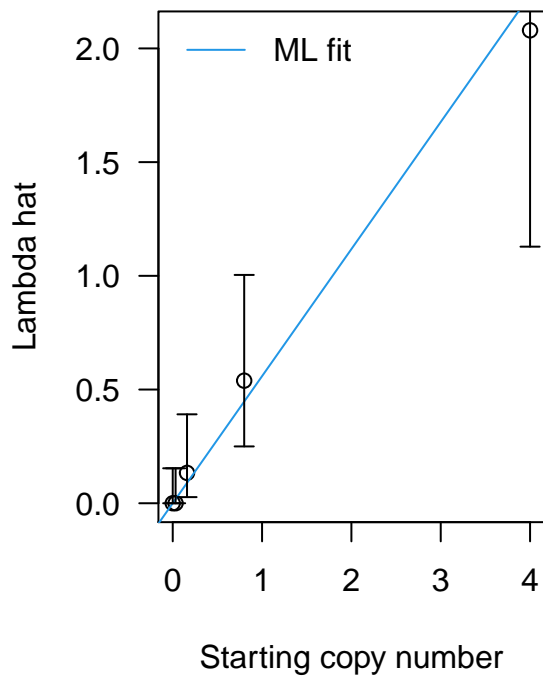


eMIDO1

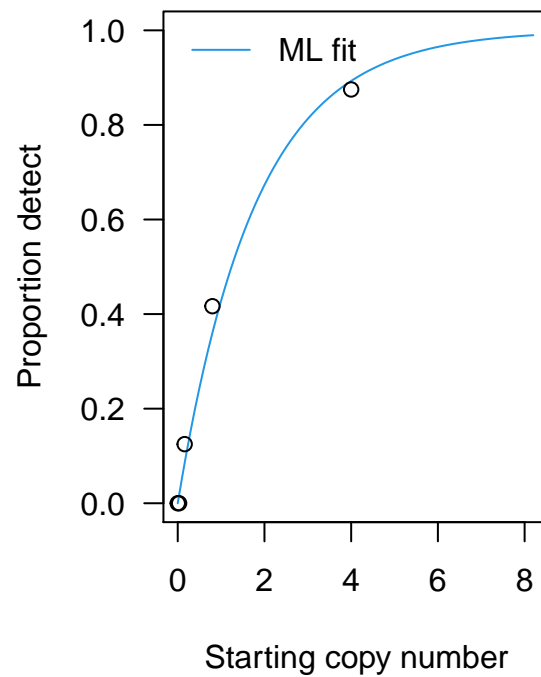


```
##
##
##
##
## eMIDO1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 5.97e-10 1.14e-01    0.0      1
## beta  8.74e-01 2.24e-01    3.9 9.7e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 4.59117 , df= 3 , p-value= 0.2043009
```

eMISA2

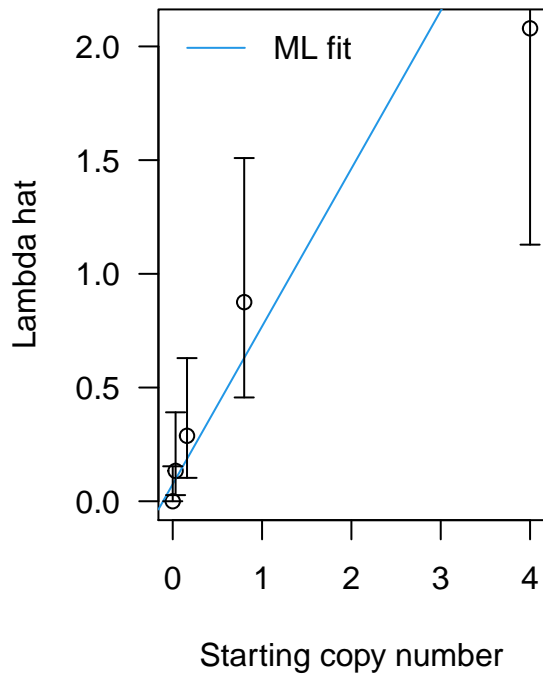


eMISA2

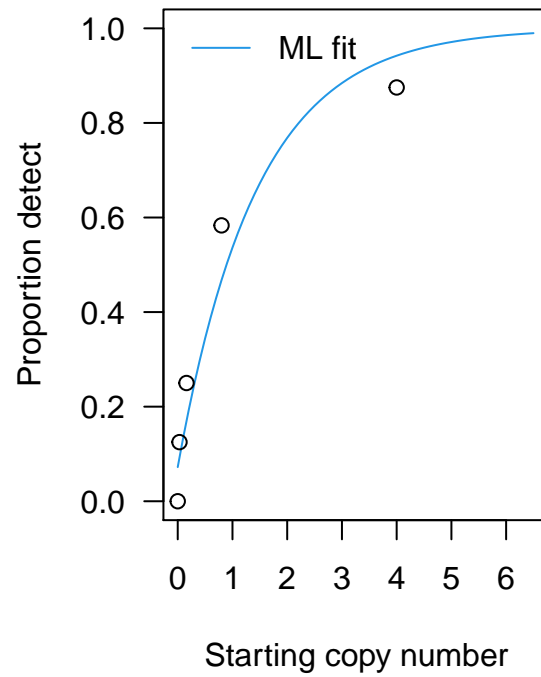


```
##
##
##
##
## eMISA2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 5.64e-10 5.92e-02   0.00    1
## beta  5.59e-01 1.32e-01   4.24 2.3e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.682044 , df= 3 , p-value= 0.6409331
```

eONKI4

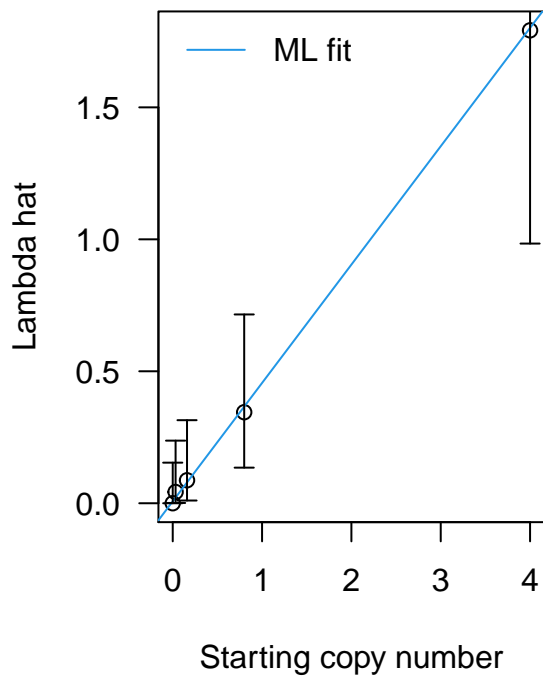


eONKI4

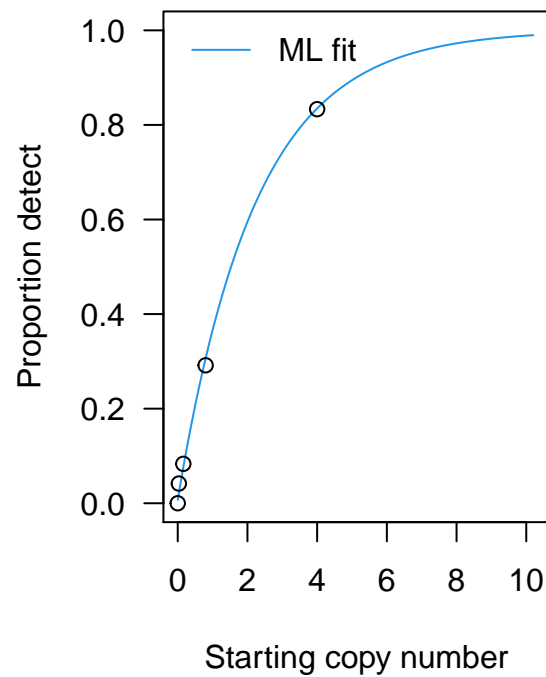


```
##
##
##
##
## eONKI4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha  0.0749  0.0486   1.54   0.12
## beta   0.6939  0.1601   4.33 1.5e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 7.671079 , df= 3 , p-value= 0.05332185
```


eRAAU1

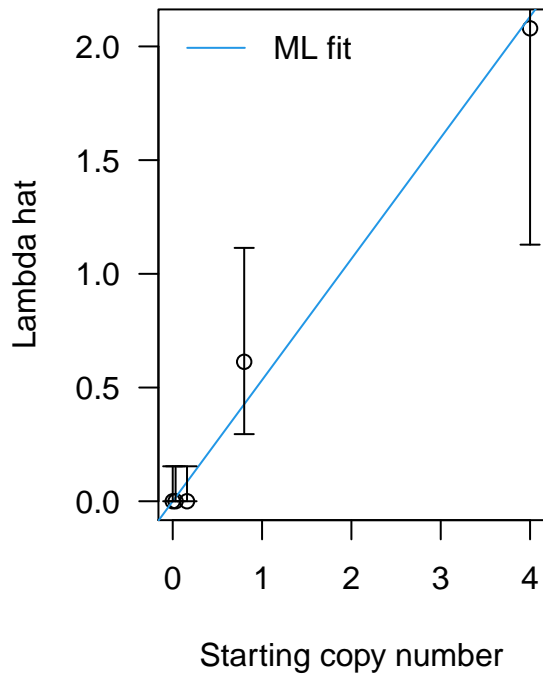


eRAAU1

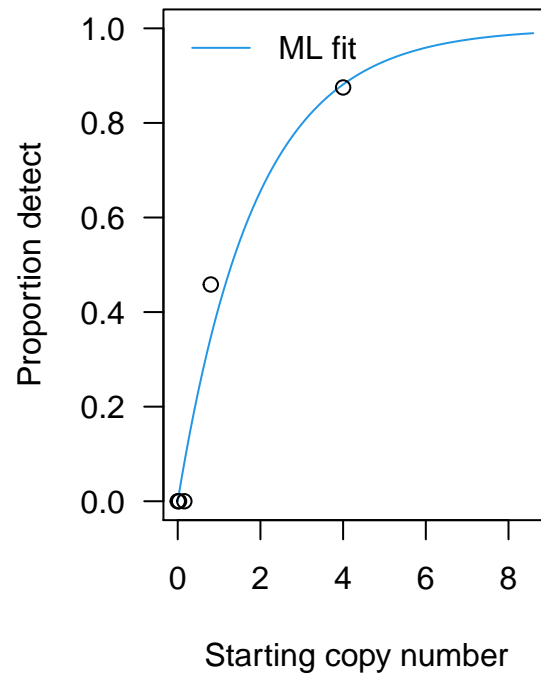


```
##
##
##
##
## eRAAU1
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha  0.00722 0.02104   0.34   0.73
## beta   0.44846 0.09693   4.63 3.7e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.7629684 , df= 3 , p-value= 0.8583024
```

eRACA2

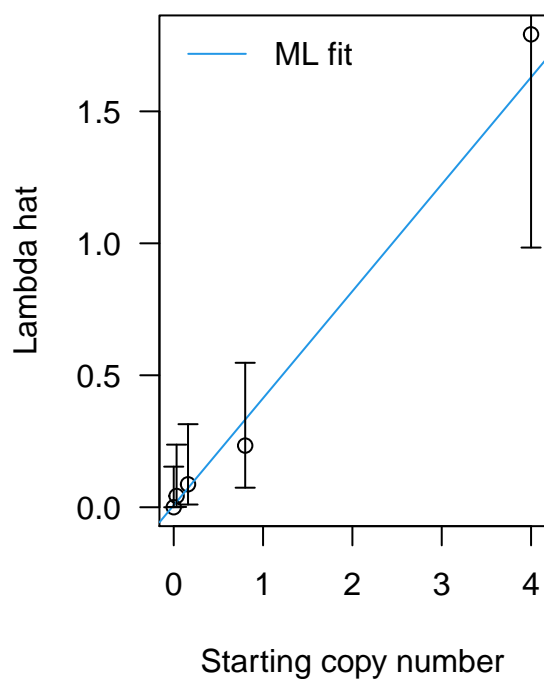


eRACA2

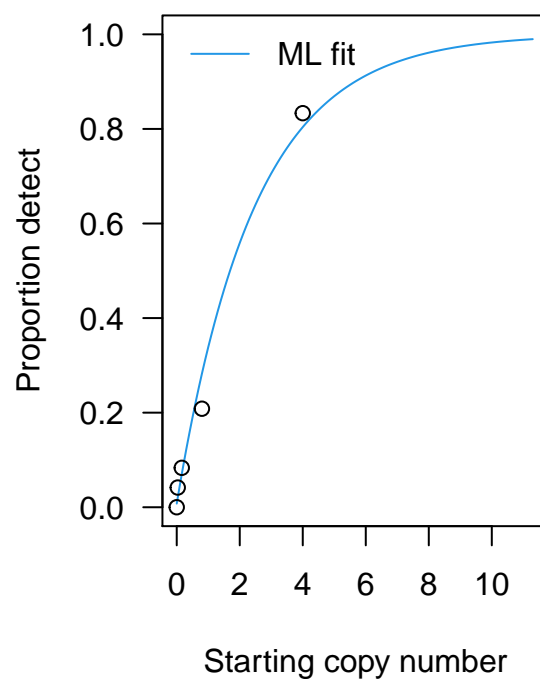


```
##
##
##
##
## eRACA2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 5.92e-09 2.14e-01    0.00 1.0000
## beta  5.33e-01 1.79e-01    2.98 0.0029 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.180137 , df= 3 , p-value= 0.1031676
```

eRALU2

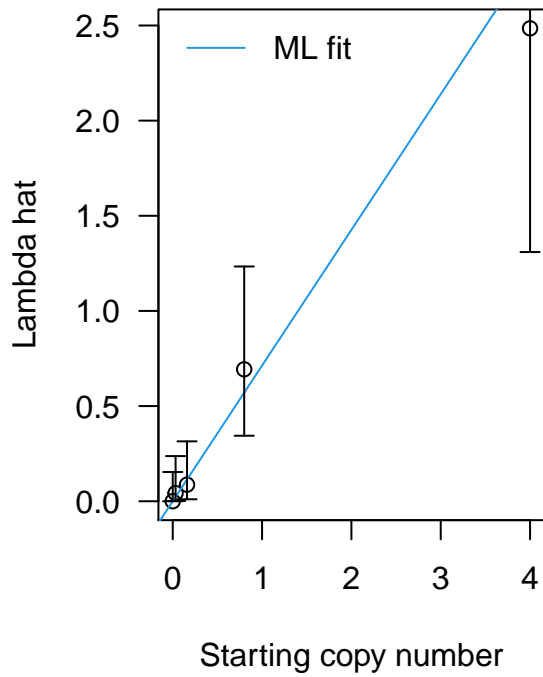


eRALU2

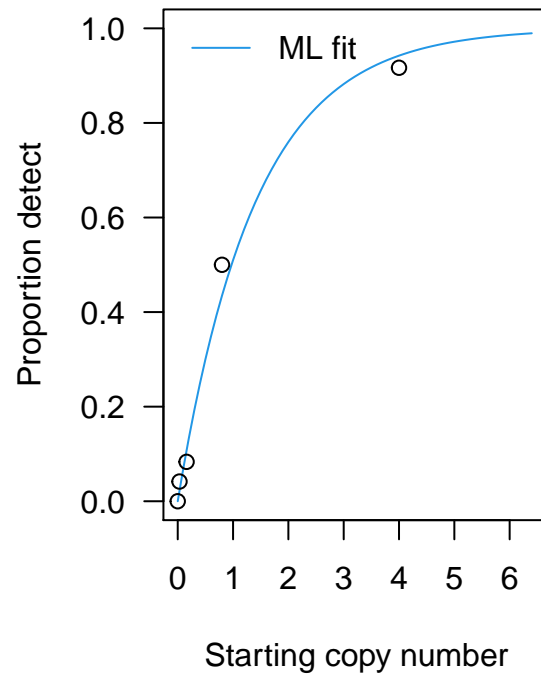


```
##
##
##
##
## eRALU2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha  0.00771 0.02002   0.39   0.7
## beta   0.40536 0.08975   4.52 6.3e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.677365 , df= 3 , p-value= 0.6419778
```

eRAPR2

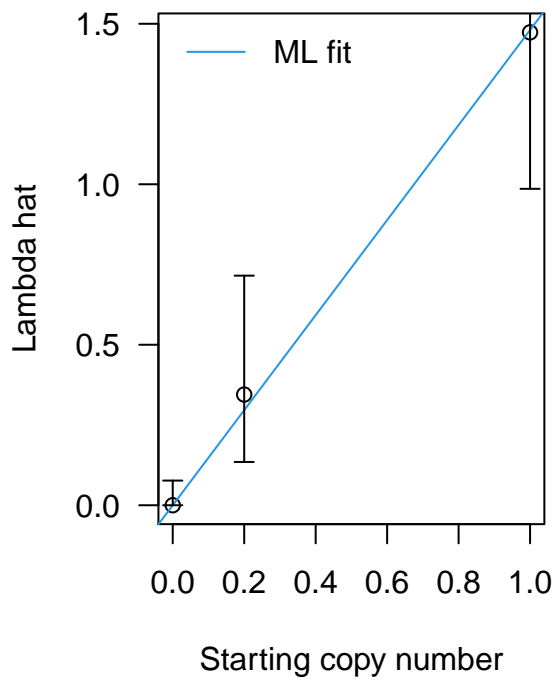


eRAPR2

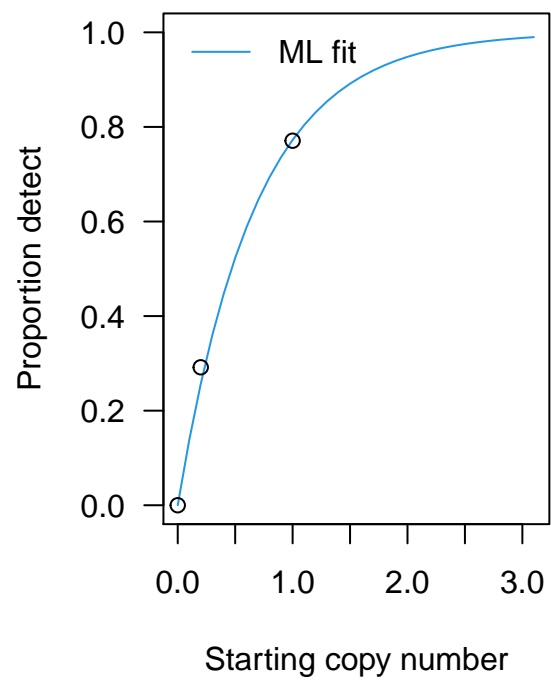


```
##
##
##
##
## eRAPR2
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 6.47e-08 2.33e-02   0.00     1
## beta  7.13e-01 1.49e-01   4.78 1.8e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 1.148937 , df= 3 , p-value= 0.7652756
```

Hno



Hno



```
##
##
##
##
## Hno
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha 1.15e-09 1.55e-01    0.0      1
## beta  1.48e+00 3.61e-01    4.1 4.2e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 0.1539404 , df= 1 , p-value= 0.6947979
```

Estimate predicted S_q given number detects and technical replicates - intercept model (not shown)

Determine Lc, Ld, Lq - no intercept model

Follows Lavagnini and Magno 2007, Mass Spectrometry Reviews *The notation in the paper was changed from the Lavagnini 2007 paper and is shown in brackets here.*

- Lc (*LOB Limit of blank*) = critical level is the assay signal above which a response is reliably attributed to the presence of analyte
- Ld (*Ld = expected number detects out of NN replicates at concentration LOD*) = signal corresponding to an analyte concentration x_d (*=LOD Limit of Detection*) level which may be a priori expected to be recognized
- Lq = quantification limit is a signal with a precision which satisfies an expected value (*=gamma_Q*)

Lc corresponds to a critical response level or a false positive rate, i.e. critical number of detects given NN replicates, above which we would reject the null hypothesis that the concentration/copy number is zero at $\alpha = \alpha_{Lc}$ (*=gamma_{FP}*). It is the critical response level corresponding to the false positive rate of α_{Lc} . Essentially, the test is positive if the $Y \sim \text{Binomial}(m, p) > Lc$. The False Positive Rate is $P(Y > Lc \mid S=0)$.

Ld is computed to correspond to the false negative rate, $\beta = \beta_{Ld}$ (*=gamma_{FN}*) here. It is computed so that the probability of observing a new (unknown concentration) response less than or equal to Lc is less than or equal to β_{Ld} . The probability of observing Lc or less detects if the concentration is x_d (*=LOD Limit of Detection*) or more is less than or equal to β_{Ld} . The values of Lc depend on the number of replicates, NN, so x_d does as well. Ld is the expected number of detects at values x_d and NN replicates. False negative rate Ld computation: $P(Y \leq Lc \mid p_{x_d}) \leq \beta_{Ld}$, and solve for x_d .

Lq is less well defined. The literature suggests using $Lq = \beta_0 + 10 \text{ s.e.}(\beta_0)$, but this uses the normality assumption. Other literature suggests using the “analyte concentration x_q (*=LOQ Limit of Quantification*) for which the experimental relative standard deviation of the responses reaches a fixed level (*=gamma_Q*), for example, the level 0.1.” Lavagnini and Magno 2007. I interpret the term “relative standard deviation” to mean the coefficient of variation, $CV = sd/\text{mean}$.

In the exercise below, we use the fits from the ML models to estimate the Lc, Ld and Lq, for various values of NN replicates for a new observation, i.e. a new (unknown concentration) response number of detects.

```
## Too few values for CID
## Too few values for NZMS
## Too few values for SAFO
## Too few values for Elod-2-NU
## Too few values for MYPI
## Too few values for Goby
## Too few values for SS
```

Determine Lc, Ld, Lq - intercept model

```
## Too few values for CID
## Too few values for NZMS
## Too few values for SAFO
## Too few values for Elod-2-NU
## Too few values for MYPI
## Too few values for Goby
## Too few values for SS
```

Estimates, Lc, Ld, Lq for a given number of technical reps NN[NNi]

Choose the model (intercept versus no intercept) with the best LLR test fit.

Limits for best choice model for N= 8

##	InterModel	alpha	aSE	beta	bSE	Lc	SdLow	Sd	SdUp	SqLow	Sq
## D-loop	0	0.00	0.00	1.13	0.15	0	0.26	0.33	0.45	0.99	1.26
## AC1_CID_1	0	0.00	0.00	0.08	0.01	0	3.65	4.94	7.65	13.80	18.69
## MYPI-6	0	0.00	0.00	0.01	0.00	0	53.32	69.83	101.12	201.79	264.24
## MYPI-6t	0	0.00	0.00	0.39	0.05	0	0.75	0.95	1.30	2.83	3.60
## AC1	0	0.00	0.00	1.64	0.24	0	0.18	0.23	0.32	0.67	0.87
## AC3	0	0.00	0.00	1.59	0.24	0	0.18	0.24	0.33	0.69	0.89
## BH1	0	0.00	0.00	0.77	0.11	0	0.38	0.48	0.67	1.44	1.83
## BH2	0	0.00	0.00	0.68	0.10	0	0.43	0.55	0.75	1.63	2.07
## GCTM10	0	0.00	0.00	1.35	0.20	0	0.22	0.28	0.39	0.82	1.05
## GCTM22	0	0.00	0.00	1.36	0.20	0	0.21	0.28	0.39	0.81	1.04
## GCTM32	0	0.00	0.00	1.28	0.19	0	0.23	0.29	0.41	0.85	1.10
## SC4	0	0.00	0.00	0.90	0.14	0	0.32	0.41	0.59	1.21	1.57
## SC5	0	0.00	0.00	0.98	0.15	0	0.30	0.38	0.54	1.12	1.45
## BHC	0	0.00	0.00	0.22	0.02	0	1.42	1.74	2.23	5.38	6.57
## Dre16s	0	0.00	0.00	0.12	0.01	0	2.68	3.18	3.91	10.15	12.04
## SVC	0	0.00	0.00	0.22	0.02	0	1.42	1.74	2.23	5.38	6.57
## YPC	0	0.00	0.00	0.39	0.05	0	0.77	0.96	1.29	2.92	3.65
## eASMO9	0	0.00	0.00	0.77	0.15	0	0.36	0.49	0.78	1.34	1.85
## eASTR4	0	0.00	0.00	0.48	0.09	0	0.56	0.77	1.24	2.12	2.93
## eFISH1	0	0.00	0.00	0.41	0.08	0	0.65	0.90	1.48	2.46	3.42
## eLIPI1	0	0.00	0.00	0.84	0.17	0	0.32	0.45	0.73	1.22	1.70
## eMIDO1	0	0.00	0.00	0.97	0.20	0	0.28	0.38	0.64	1.04	1.46
## eMISA2	0	0.00	0.00	0.58	0.11	0	0.46	0.64	1.04	1.76	2.43
## eONKI4	1	0.07	0.05	0.69	0.16	2	0.74	1.21	2.41	1.24	1.93
## eRAAU1	0	0.00	0.00	0.46	0.09	0	0.58	0.81	1.33	2.21	3.07
## eRACA2	0	0.00	0.00	0.54	0.11	0	0.50	0.69	1.13	1.88	2.61
## eRALU2	0	0.00	0.00	0.42	0.09	0	0.64	0.89	1.49	2.41	3.38
## eRAPR2	0	0.00	0.00	0.71	0.14	0	0.38	0.52	0.85	1.44	1.98
## Hno	0	0.00	0.00	1.51	0.25	0	0.19	0.25	0.36	0.71	0.94
##	SqUp										
## D-loop	1.71										
## AC1_CID_1	28.95										
## MYPI-6	382.66										
## MYPI-6t	4.92										
## AC1	1.22										
## AC3	1.26										
## BH1	2.52										
## BH2	2.85										
## GCTM10	1.47										
## GCTM22	1.47										
## GCTM32	1.55										
## SC4	2.22										
## SC5	2.05										
## BHC	8.44										
## Dre16s	14.80										
## SVC	8.44										
## YPC	4.87										

## eASM09	2.96
## eASTR4	4.70
## eFISH1	5.61
## eLIP11	2.78
## eMID01	2.41
## eMISA2	3.92
## eONKI4	3.73
## eRAAU1	5.04
## eRACA2	4.28
## eRALU2	5.65
## eRAPR2	3.21
## Hno	1.38

Tables and Graphs for the manuscript

Include MYPI-6t and eASMO9, eFISH1, eONKI4 in paper.

Index into uTargets (5, 25, 27, 31)

##		alpha	aSE	beta	bSE	Lc	LODL	LOD	LODU	LOQL	LOQ	LOQU	K.LOD8
##	-----	-----	-----	-----	-----	---	-----	-----	-----	-----	-----	-----	-----
##	AC1	0.0	0	1.6	0.2	0	0.2	0.2	0.3	0.7	0.9	1.2	0.2
##	AC1_CID_1	0.0	0	0.1	0.0	0	3.6	4.9	7.7	13.8	18.7	29.0	5.6
##	AC3	0.0	0	1.6	0.2	0	0.2	0.2	0.3	0.7	0.9	1.3	0.2
##	BH1	0.0	0	0.8	0.1	0	0.4	0.5	0.7	1.4	1.8	2.5	0.3
##	BH2	0.0	0	0.7	0.1	0	0.4	0.5	0.8	1.6	2.1	2.9	0.2
##	BHC	0.0	0	0.2	0.0	0	1.4	1.7	2.2	5.4	6.6	8.4	1.9
##	D-loop	0.0	0	1.1	0.2	0	0.3	0.3	0.5	1.0	1.3	1.7	0.3
##	Dre16s	0.0	0	0.1	0.0	0	2.7	3.2	3.9	10.1	12.0	14.8	3.4
##	eASMO9	0.0	0	0.8	0.1	0	0.4	0.5	0.8	1.3	1.8	3.0	0.3
##	eASTR4	0.0	0	0.5	0.1	0	0.6	0.8	1.2	2.1	2.9	4.7	0.5
##	eFISH1	0.0	0	0.4	0.1	0	0.7	0.9	1.5	2.5	3.4	5.6	0.6
##	eLIPI1	0.0	0	0.8	0.2	0	0.3	0.4	0.7	1.2	1.7	2.8	0.3
##	eMID01	0.0	0	1.0	0.2	0	0.3	0.4	0.6	1.0	1.5	2.4	0.4
##	eMISA2	0.0	0	0.6	0.1	0	0.5	0.6	1.0	1.8	2.4	3.9	0.5
##	eONKI4	0.1	0	0.7	0.2	2	0.7	1.2	2.4	1.2	1.9	3.7	0.2
##	eRAAU1	0.0	0	0.5	0.1	0	0.6	0.8	1.3	2.2	3.1	5.0	0.8
##	eRACA2	0.0	0	0.5	0.1	0	0.5	0.7	1.1	1.9	2.6	4.3	0.6
##	eRALU2	0.0	0	0.4	0.1	0	0.6	0.9	1.5	2.4	3.4	5.6	0.7
##	eRAPR2	0.0	0	0.7	0.1	0	0.4	0.5	0.8	1.4	2.0	3.2	0.4
##	GCTM10	0.0	0	1.4	0.2	0	0.2	0.3	0.4	0.8	1.0	1.5	0.2
##	GCTM22	0.0	0	1.4	0.2	0	0.2	0.3	0.4	0.8	1.0	1.5	0.2
##	GCTM32	0.0	0	1.3	0.2	0	0.2	0.3	0.4	0.9	1.1	1.6	0.3
##	Hno	0.0	0	1.5	0.2	0	0.2	0.2	0.4	0.7	0.9	1.4	0.2
##	MYPI-6	0.0	0	0.0	0.0	0	53.3	69.8	101.1	201.8	264.2	382.7	78.0
##	MYPI-6t	0.0	0	0.4	0.1	0	0.7	1.0	1.3	2.8	3.6	4.9	1.0
##	SC4	0.0	0	0.9	0.1	0	0.3	0.4	0.6	1.2	1.6	2.2	0.4
##	SC5	0.0	0	1.0	0.1	0	0.3	0.4	0.5	1.1	1.5	2.1	0.4
##	SVC	0.0	0	0.2	0.0	0	1.4	1.7	2.2	5.4	6.6	8.4	1.9
##	YPC	0.0	0	0.4	0.0	0	0.8	1.0	1.3	2.9	3.6	4.9	1.0

Table: MYPI-6t
##

##	S	num.detect	n	p.tilde	lambda.tilde
##	-----	-----	---	-----	-----
##	0.000	0	24	0.000	0.000
##	0.244	3	24	0.125	0.134
##	0.488	4	24	0.167	0.182
##	0.975	9	24	0.375	0.470
##	1.950	9	24	0.375	0.470
##	3.906	22	24	0.917	2.485
##	7.812	22	24	0.917	2.485

##

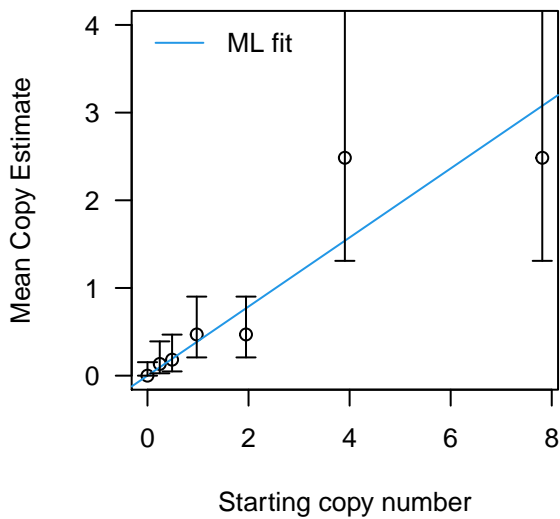
```

## Table: eASM09
##
##      S      num.detect      n      p.tilde      lambda.tilde
## -----
## 0.000          0      24      0.000          0.000
## 0.032          0      24      0.000          0.000
## 0.160          6      24      0.250          0.288
## 0.800         11      24      0.458          0.613
## 4.000         22      24      0.917          2.485
##
##
## Table: eFISH1
##
##      S      num.detect      n      p.tilde      lambda.tilde
## -----
## 0.000          0      24      0.000          0.000
## 0.032          1      24      0.042          0.043
## 0.160          0      24      0.000          0.000
## 0.800         10      24      0.417          0.539
## 4.000         18      24      0.750          1.386
##
##
## Table: eONKI4
##
##      S      num.detect      n      p.tilde      lambda.tilde
## -----
## 0.000          0      24      0.000          0.000
## 0.032          3      24      0.125          0.134
## 0.160          6      24      0.250          0.288
## 0.800         14      24      0.583          0.875
## 4.000         21      24      0.875          2.079
##
##
## MYPI-6t
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta      0.394   0.054    7.29 3.1e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.787625 , df= 6 , p-value= 0.340935
##
##
## eASM09
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## beta      0.767   0.147    5.22 1.8e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 5.106559 , df= 4 , p-value= 0.2765376
##
##
## eFISH1
## Convergence= 0

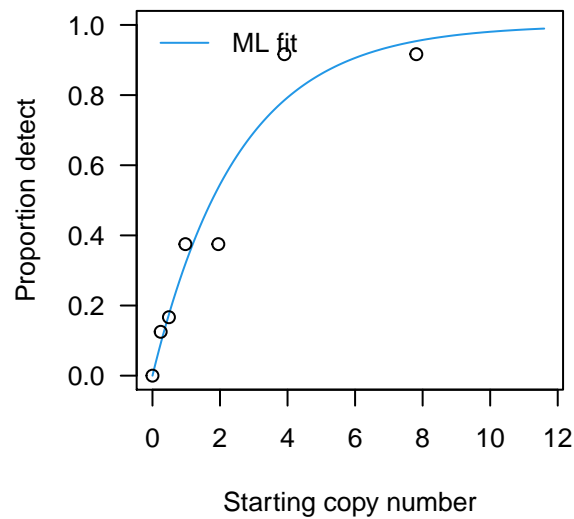
```

```
##      Estimate Std.Err Z value Pr(>z)
## beta   0.4139  0.0823    5.03 4.9e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 6.63572 , df= 4 , p-value= 0.1564373
##
##
## eONKI4
## Convergence= 0
##      Estimate Std.Err Z value Pr(>z)
## alpha   0.0749  0.0486    1.54  0.12
## beta    0.6939  0.1601    4.33 1.5e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## LLR test stat= 7.671079 , df= 3 , p-value= 0.05332185
```

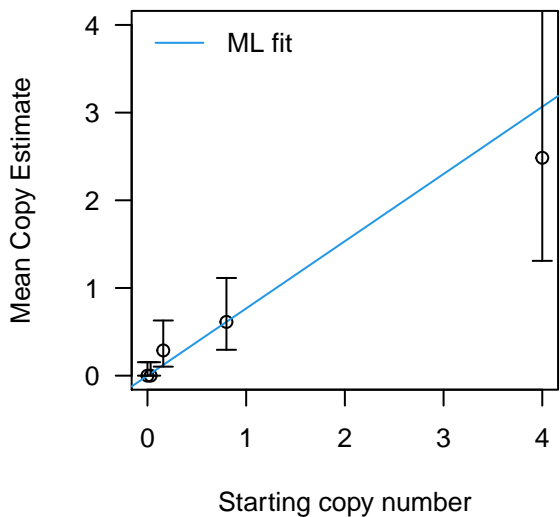
MYPI-6t



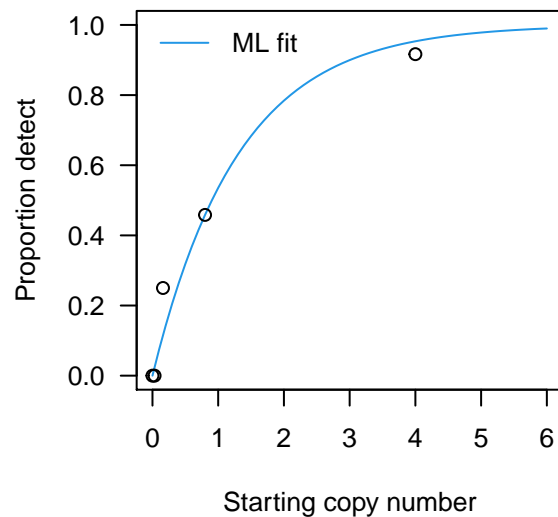
MYPI-6t



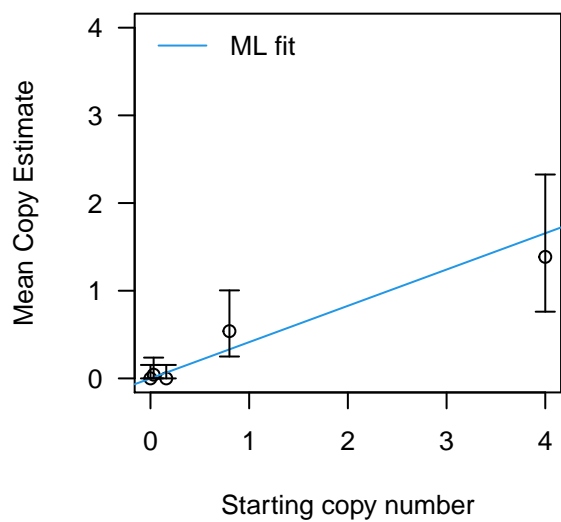
eASMO9



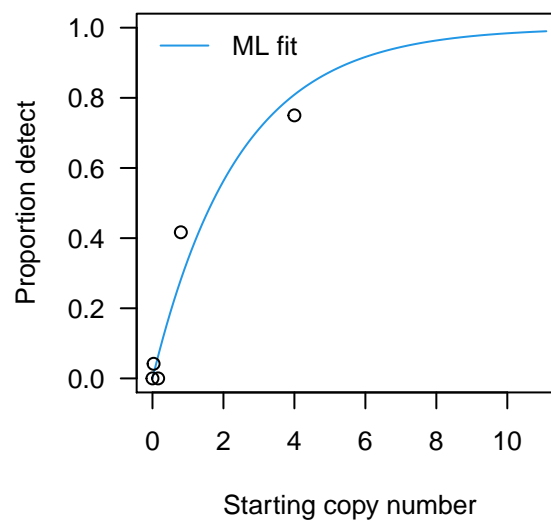
eASMO9



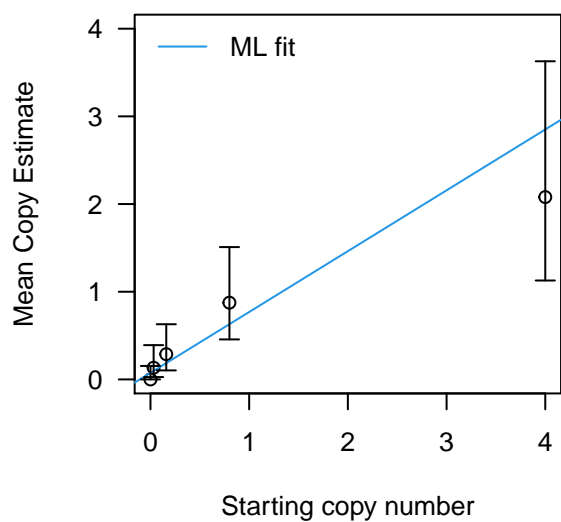
eFISH1



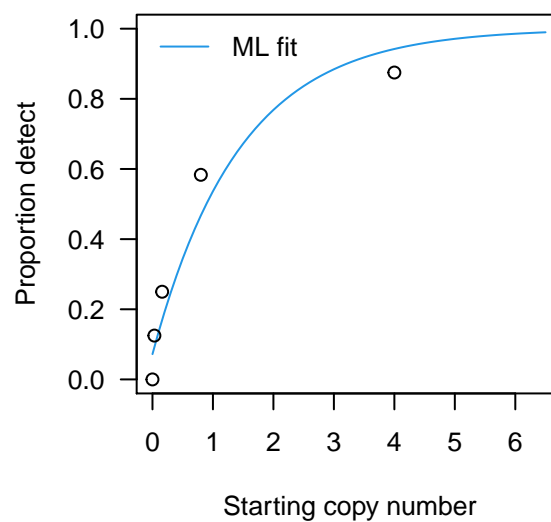
eFISH1

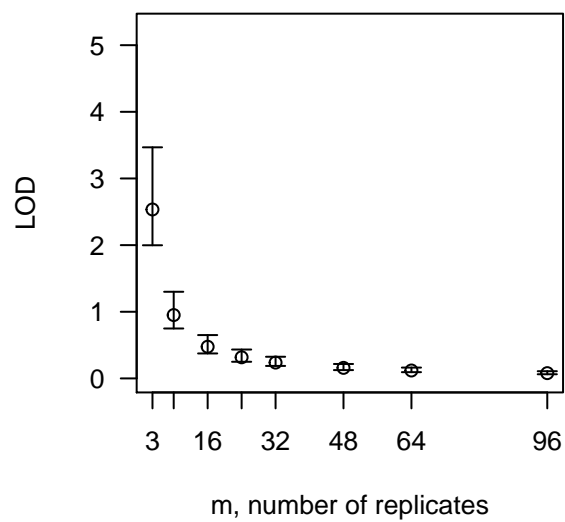
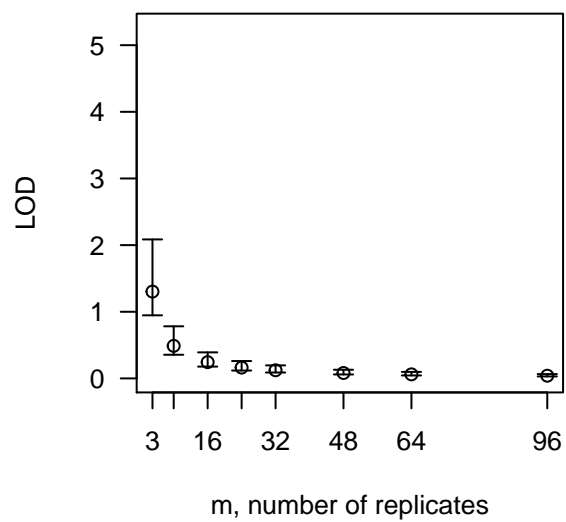
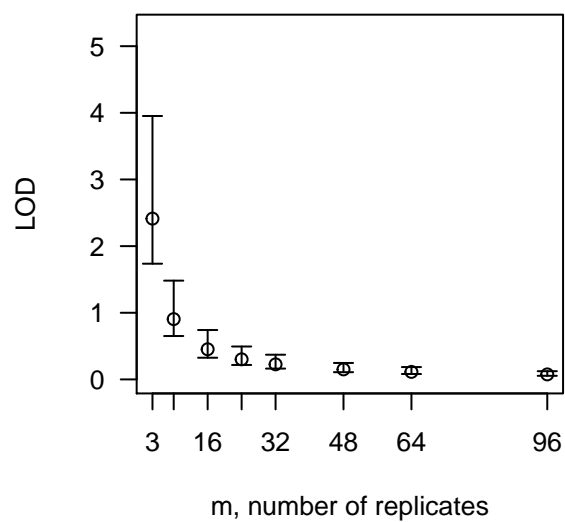
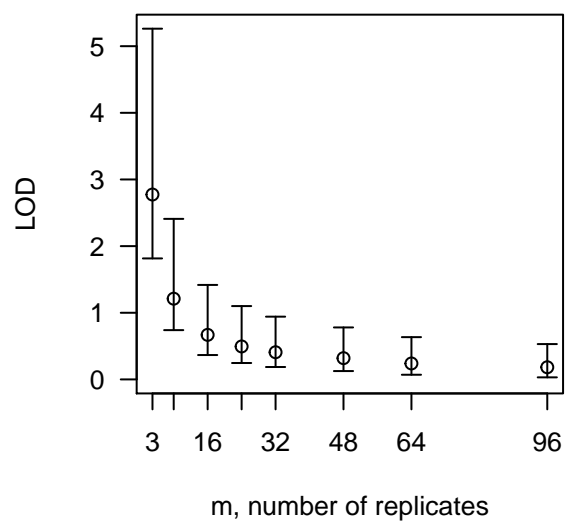


eONKI4

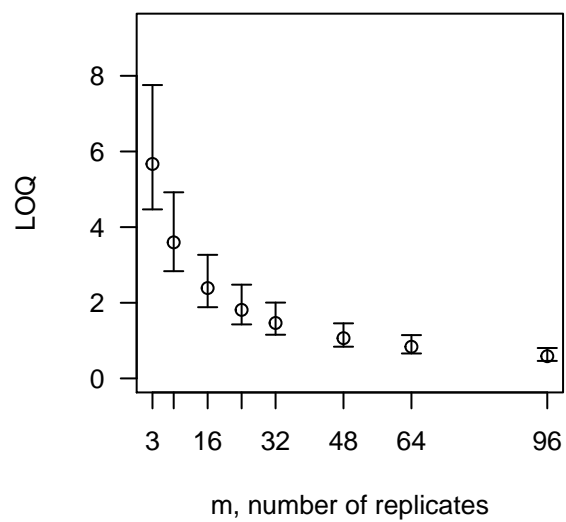


eONKI4

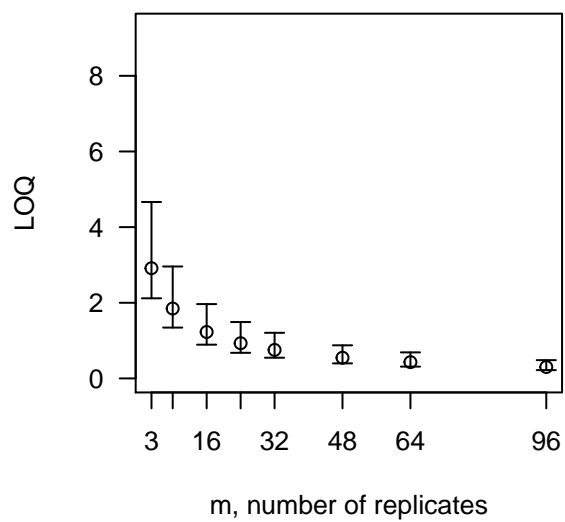


MYPI-6t**eASMO9****eFISH1****eONKI4**

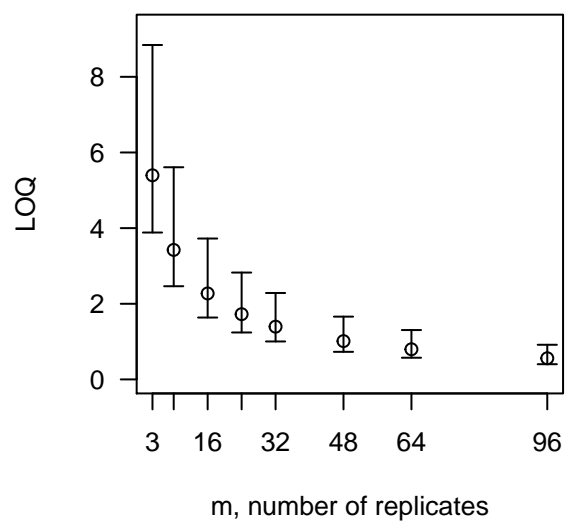
MYPI-6t



eASMO9



eFISH1



eONKI4

