

# Moments of K

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```
suppressWarnings(suppressPackageStartupMessages(library(GDFMM)))
suppressWarnings(suppressPackageStartupMessages(library(ACutils)))
suppressWarnings(suppressPackageStartupMessages(library(tidyverse)))
suppressWarnings(suppressPackageStartupMessages(library(RColorBrewer)))
suppressWarnings(suppressPackageStartupMessages(library(salso)))
suppressWarnings(suppressPackageStartupMessages(library(wesanderson)))
suppressWarnings(suppressPackageStartupMessages(library(mcclust.ext)))
```

Moments of the Number of Clusters  $K$  a priori

## One level

I first consider the case where  $d = 1$  and  $n = 100$

```
n_j = 100
Npoints1 = 30
Npoints2 = 10
gamma_grid = seq(0.01,2,length.out = Npoints1)
Lambda_grid = seq(1,10,length.out = Npoints2)
Exp_val = matrix(nrow = Npoints1,ncol = Npoints2)
Var = matrix(nrow = Npoints1,ncol = Npoints2)

for(i in 1:Npoints1){
  for(j in 1:Npoints2){
    Moments = Expected_prior(n_j = n_j, gamma = gamma_grid[i], type = "distinct",
                           prior = "Poisson", lambda = Lambda_grid[j])
    Exp_val[i,j] = Moments$Mean
    Var[i,j] = Moments$Variance
  }
}

Lambda_names = unlist( lapply( as.list(as.character(round(Lambda_grid, digits = 1))),
                              FUN = function(x){ paste0( "Lambda", "=",x) }
                            )
                     )

gamma_names = unlist( lapply( as.list(as.character(round(gamma_grid, digits = 2))),
                              FUN = function(x){ paste0( "gamma", "=",x) }
                            )
                     )
```

```
rownames(Exp_val) <- rownames(Var) <- gamma_names
colnames(Exp_val) <- colnames(Var) <- Lambda_names
```

Guardo la media

```
#Exp_val
round(Exp_val, digits = 3)
```

##	Lambda=1	Lambda=2	Lambda=3	Lambda=4	Lambda=5	Lambda=6	Lambda=7
## gamma=0.01	1.050	1.100	1.149	1.199	1.247	1.296	1.345
## gamma=0.08	1.320	1.630	1.931	2.223	2.509	2.788	3.061
## gamma=0.15	1.500	1.979	2.441	2.887	3.320	3.740	4.149
## gamma=0.22	1.625	2.222	2.796	3.349	3.884	4.404	4.909
## gamma=0.28	1.715	2.397	3.052	3.684	4.296	4.889	5.467
## gamma=0.35	1.780	2.525	3.243	3.935	4.606	5.257	5.891
## gamma=0.42	1.828	2.622	3.387	4.127	4.844	5.542	6.222
## gamma=0.49	1.864	2.696	3.498	4.276	5.032	5.767	6.484
## gamma=0.56	1.892	2.753	3.586	4.395	5.181	5.948	6.697
## gamma=0.63	1.913	2.797	3.655	4.489	5.302	6.095	6.870
## gamma=0.7	1.930	2.833	3.711	4.566	5.401	6.216	7.013
## gamma=0.76	1.943	2.861	3.755	4.629	5.482	6.316	7.133
## gamma=0.83	1.953	2.884	3.792	4.680	5.549	6.400	7.234
## gamma=0.9	1.962	2.902	3.822	4.723	5.606	6.471	7.319
## gamma=0.97	1.968	2.917	3.847	4.759	5.654	6.531	7.392
## gamma=1.04	1.974	2.930	3.868	4.789	5.694	6.583	7.455
## gamma=1.11	1.978	2.940	3.885	4.815	5.729	6.627	7.510
## gamma=1.18	1.981	2.948	3.900	4.837	5.759	6.665	7.557
## gamma=1.25	1.984	2.955	3.913	4.856	5.784	6.699	7.598
## gamma=1.31	1.987	2.961	3.923	4.872	5.807	6.728	7.634
## gamma=1.38	1.989	2.967	3.932	4.886	5.826	6.753	7.666
## gamma=1.45	1.990	2.971	3.940	4.898	5.843	6.775	7.695
## gamma=1.52	1.992	2.974	3.947	4.908	5.858	6.795	7.720
## gamma=1.59	1.993	2.978	3.953	4.917	5.871	6.813	7.742
## gamma=1.66	1.994	2.980	3.958	4.925	5.882	6.828	7.762
## gamma=1.73	1.995	2.983	3.962	4.933	5.893	6.842	7.780
## gamma=1.79	1.995	2.985	3.966	4.939	5.902	6.854	7.796
## gamma=1.86	1.996	2.986	3.969	4.944	5.910	6.866	7.811
## gamma=1.93	1.997	2.988	3.972	4.949	5.917	6.876	7.824
## gamma=2	1.997	2.989	3.975	4.954	5.924	6.885	7.836
##	Lambda=8	Lambda=9	Lambda=10				
## gamma=0.01	1.393	1.441	1.488				
## gamma=0.08	3.329	3.591	3.848				
## gamma=0.15	4.549	4.939	5.320				
## gamma=0.22	5.401	5.882	6.351				
## gamma=0.28	6.029	6.578	7.115				
## gamma=0.35	6.509	7.112	7.702				
## gamma=0.42	6.885	7.532	8.165				
## gamma=0.49	7.184	7.868	8.538				
## gamma=0.56	7.428	8.142	8.842				
## gamma=0.63	7.627	8.369	9.094				
## gamma=0.7	7.793	8.557	9.306				
## gamma=0.76	7.933	8.716	9.485				
## gamma=0.83	8.051	8.852	9.637				
## gamma=0.9	8.151	8.968	9.769				

```
## gamma=0.97      8.238      9.067      9.882
## gamma=1.04      8.312      9.154      9.981
## gamma=1.11      8.377      9.230     10.068
## gamma=1.18      8.434      9.296     10.144
## gamma=1.25      8.484      9.355     10.212
## gamma=1.31      8.527      9.406     10.272
## gamma=1.38      8.566      9.452     10.325
## gamma=1.45      8.601      9.493     10.373
## gamma=1.52      8.631      9.530     10.416
## gamma=1.59      8.659      9.563     10.455
## gamma=1.66      8.684      9.593     10.490
## gamma=1.73      8.706      9.620     10.522
## gamma=1.79      8.726      9.644     10.551
## gamma=1.86      8.744      9.667     10.577
## gamma=1.93      8.761      9.687     10.601
## gamma=2         8.776      9.705     10.623
```

- Valori piccoli di  $\gamma$  hanno un certo impatto, che però svanisce al crescere di  $\gamma$ .
- La soglia da cui in poi  $\gamma$  non impatta più, dipende da  $\Lambda$ , più cresce  $\Lambda$ , più valori di  $\gamma$  hanno effetto
- Da quella soglia in poi,  $E[K]$  è determinato solo da  $\Lambda$

```
#Var
round(Var, digits = 3)
```

```
##          Lambda=1 Lambda=2 Lambda=3 Lambda=4 Lambda=5 Lambda=6 Lambda=7
## gamma=0.01    0.050    0.099    0.148    0.196    0.244    0.291    0.337
## gamma=0.08    0.313    0.605    0.879    1.138    1.384    1.619    1.844
## gamma=0.15    0.484    0.922    1.324    1.699    2.050    2.382    2.699
## gamma=0.22    0.601    1.138    1.628    2.080    2.503    2.901    3.279
## gamma=0.28    0.685    1.295    1.850    2.361    2.839    3.288    3.714
## gamma=0.35    0.748    1.414    2.020    2.580    3.103    3.594    4.059
## gamma=0.42    0.795    1.507    2.156    2.757    3.318    3.845    4.343
## gamma=0.49    0.832    1.581    2.267    2.903    3.497    4.056    4.583
## gamma=0.56    0.861    1.641    2.359    3.026    3.650    4.236    4.790
## gamma=0.63    0.884    1.691    2.437    3.131    3.781    4.393    4.970
## gamma=0.7     0.903    1.732    2.502    3.221    3.895    4.530    5.130
## gamma=0.76    0.918    1.767    2.558    3.299    3.996    4.651    5.271
## gamma=0.83    0.930    1.796    2.607    3.368    4.084    4.759    5.397
## gamma=0.9     0.940    1.821    2.649    3.427    4.162    4.855    5.510
## gamma=0.97    0.949    1.843    2.685    3.480    4.231    4.941    5.612
## gamma=1.04    0.956    1.861    2.717    3.527    4.293    5.018    5.705
## gamma=1.11    0.962    1.877    2.745    3.568    4.349    5.088    5.789
## gamma=1.18    0.967    1.891    2.770    3.605    4.399    5.151    5.865
## gamma=1.25    0.971    1.902    2.791    3.638    4.444    5.209    5.935
## gamma=1.31    0.975    1.913    2.811    3.668    4.484    5.261    5.999
## gamma=1.38    0.978    1.922    2.828    3.694    4.521    5.309    6.057
## gamma=1.45    0.981    1.930    2.843    3.718    4.555    5.352    6.111
## gamma=1.52    0.983    1.937    2.857    3.740    4.585    5.392    6.160
## gamma=1.59    0.985    1.943    2.869    3.759    4.613    5.428    6.206
## gamma=1.66    0.987    1.949    2.880    3.777    4.638    5.462    6.248
## gamma=1.73    0.988    1.953    2.890    3.793    4.661    5.493    6.287
## gamma=1.79    0.990    1.958    2.898    3.807    4.682    5.521    6.323
## gamma=1.86    0.991    1.962    2.906    3.821    4.702    5.547    6.357
## gamma=1.93    0.992    1.965    2.914    3.833    4.720    5.572    6.388
## gamma=2       0.993    1.968    2.920    3.844    4.736    5.594    6.417
```

##	Lambda=8	Lambda=9	Lambda=10
## gamma=0.01	0.384	0.429	0.474
## gamma=0.08	2.061	2.270	2.472
## gamma=0.15	3.001	3.290	3.569
## gamma=0.22	3.639	3.983	4.314
## gamma=0.28	4.118	4.504	4.873
## gamma=0.35	4.499	4.919	5.320
## gamma=0.42	4.815	5.264	5.692
## gamma=0.49	5.082	5.557	6.008
## gamma=0.56	5.314	5.811	6.284
## gamma=0.63	5.517	6.035	6.527
## gamma=0.7	5.696	6.234	6.744
## gamma=0.76	5.857	6.412	6.938
## gamma=0.83	6.000	6.572	7.114
## gamma=0.9	6.130	6.717	7.274
## gamma=0.97	6.248	6.849	7.420
## gamma=1.04	6.355	6.970	7.553
## gamma=1.11	6.452	7.081	7.676
## gamma=1.18	6.541	7.182	7.789
## gamma=1.25	6.623	7.276	7.894
## gamma=1.31	6.698	7.362	7.990
## gamma=1.38	6.768	7.442	8.080
## gamma=1.45	6.832	7.515	8.164
## gamma=1.52	6.891	7.584	8.242
## gamma=1.59	6.945	7.648	8.314
## gamma=1.66	6.996	7.707	8.382
## gamma=1.73	7.044	7.763	8.446
## gamma=1.79	7.087	7.815	8.505
## gamma=1.86	7.128	7.863	8.561
## gamma=1.93	7.167	7.909	8.613
## gamma=2	7.203	7.951	8.663

- Per quanto riguarda la varianza, l'impatto di  $\gamma$  sembra trascurabile rispetto a  $\Lambda$ , soprattutto per  $\gamma > 1$ .
- In più, la varianza sembra essere sempre simile alla media, cosa che può dato che  $\Lambda$  è distribuita con una Poisson. Potrebbe tuttavia essere una limitazione.

## Distribution for $\Lambda$ fixed

Fisso  $\Lambda = 4$  e guardo la distribuzione del numero di cluster al variare di  $\gamma$

```
Npoints = 100
Lambda = 4
gamma_grid = seq(0.03,3,length.out = Npoints)
prob_k = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){

  #threshold = 0.995
  for(kk in 1:n_j){
    prob_k[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = gamma_grid[i],
                                     prior = "Poisson",
                                     lambda = Lambda)

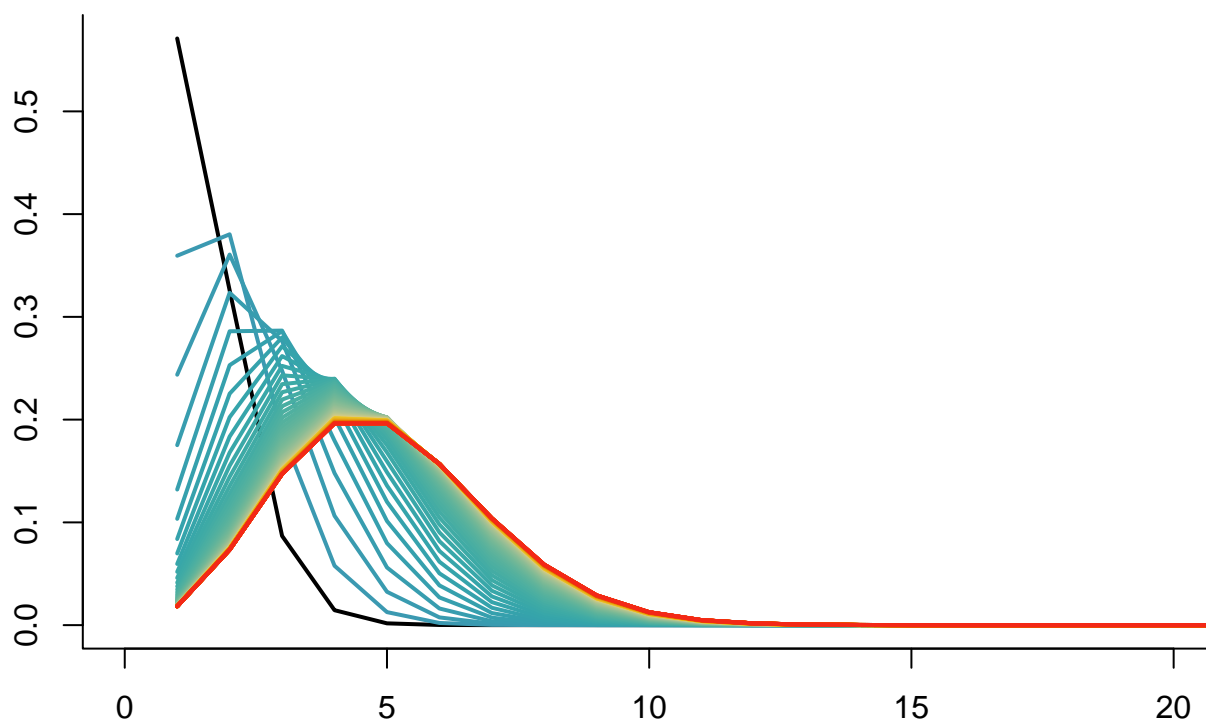
    #if(sum(prob_k)>=threshold)
    #break
  }
}
```

```
}
}
```

La curva nera è la prima, quella con  $\gamma$  minore

```
par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k), type = "l", lwd = 2, xlim = c(0,20), lty = 1,
         col = c("black", hcl.colors(n=Npoints, palette = "Zissou 1")),
         main = paste0( paste0( "Distribution of K for Lambda=4 and gamma varying" ) ) )
```

## Distribution of K for Lambda=4 and gamma varying



Come mi aspettavo, fissato  $\Lambda$  ci sono delle differenze iniziali, che diventano poi minime quando la media di avvicina a 4.

Rifaccio con  $\Lambda = 20$

```
Npoints = 100
Lambda = 20
gamma_grid = seq(0.01,2,length.out = Npoints)
prob_k2 = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){
  #threshold = 0.995
  for(kk in 1:n_j){
    prob_k2[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = gamma_grid[i],
                                     prior = "Poisson",
                                     lambda = Lambda)
```

```

    #if(sum(prob_k)>=threshold)
    #break
  }
}

```

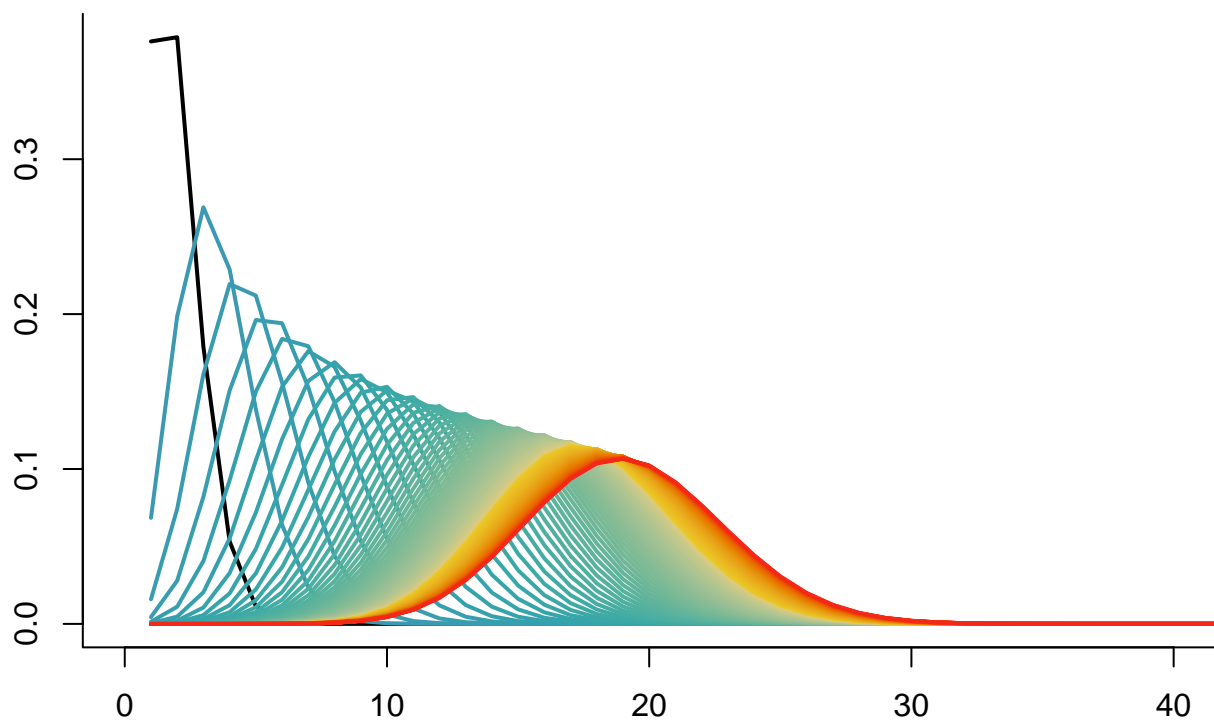
La curva nera è la prima, quella con  $\gamma$  minore

```

par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k2), type = "l", lwd = 2,xlim = c(0,40), lty = 1,
  col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
  main = paste0( paste0( "Distribution of K for Lambda=20 and gamma varying" ) ) )

```

## Distribution of K for Lambda=20 and gamma varying



La variabilità sembra essere sempre simile

## Distribution for $\gamma$ fixed

Ora faccio l'esperimento inverso, tengo fisso  $\gamma$  e faccio variare  $\Lambda$ .

**gamma = 2**

```

Npoints = 20
gamma = 2
Lambda_grid = seq(1,20,length.out = Npoints)
prob_k3 = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){

```

```

#threshold = 0.995
for(kk in 1:n_j){
  prob_k3[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                   gamma = gamma,
                                   prior = "Poisson",
                                   lambda = Lambda_grid[i])

  #if(sum(prob_k)>=threshold)
  #break
}
}

```

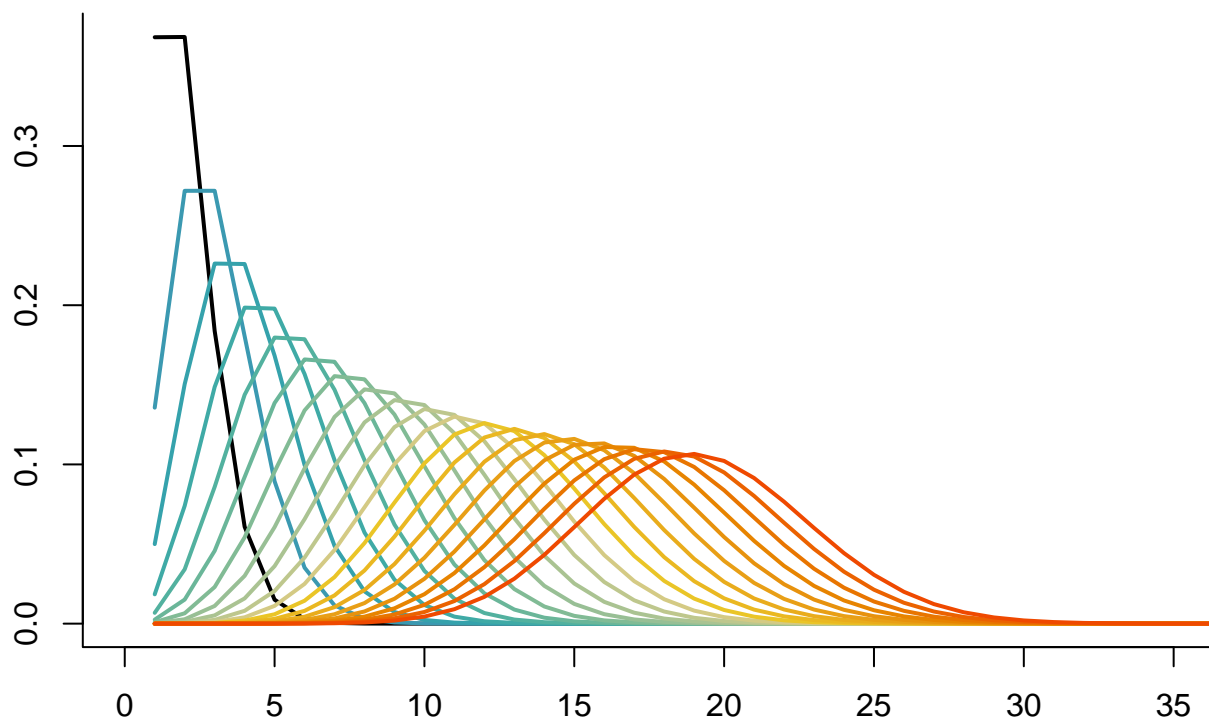
La curva nera è la prima, quella con  $\Lambda$  minore. Ogni curva, corrisponde ad aumentare  $\Lambda$  di 1.

```

par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k3), type = "l", lwd = 2,xlim = c(0,35),lty = 1,
        col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
        main = paste0( paste0( "Distribution of K for gamma=2 and Lambda varying" ) ) )

```

## Distribution of K for gamma=2 and Lambda varying



gamma = 0.1

Ripeto lo stesso esperimento abbassando il valore di  $\gamma$ . Ogni curva, corrisponde ad aumentare  $\Lambda$  di 1.

```

Npoints = 20
gamma = 0.1
Lambda_grid = seq(1,20,length.out = Npoints)
prob_k3 = matrix(nrow = Npoints, ncol = n_j)

```

```

for(i in 1:Npoints){

  #threshold = 0.995
  for(kk in 1:n_j ){
    prob_k3[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = gamma,
                                     prior = "Poisson",
                                     lambda = Lambda_grid[i])

    #if(sum(prob_k)>=threshold)
    #break
  }
}

```

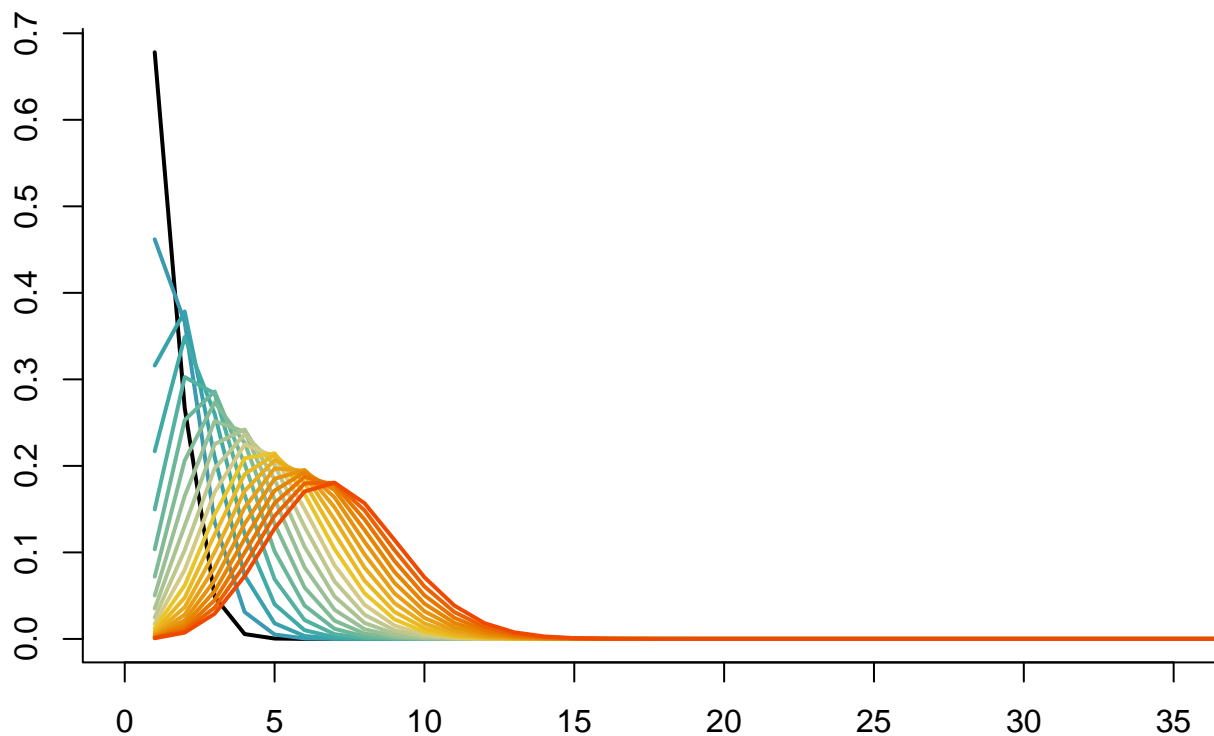
La curva nera è la prima, quella con  $\Lambda$  minore

```

par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k3), type = "l", lwd = 2,xlim = c(0,35), lty = 1,
        col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
        main = paste0( paste0( "Distribution of K for gamma=0.1 and Lambda varying") ))

```

## Distribution of K for gamma=0.1 and Lambda varying



Al crescere di  $\Lambda$  le curve si spostano ma molto più lentamente.

## CONCLUSIONI

- $\Lambda$  è una sorta di upper bound per il numero di cluster,  $E[K] < \Lambda$ .
- $\gamma$  aiuta a ridurre  $E[K]$  al di sotto della soglia  $\Lambda$ . Partendo da valori di  $\gamma$  piccoli:
  - Se  $\gamma$  cresce, cresce anche  $E[K]$  ma rispettando il vincolo  $E[K] < \Lambda$ . Quindi da un certo valore di  $\gamma$  in poi, anche aumentando  $\gamma$  non cambia niente.



- **Non possiamo controllare sia media che varianza.** Possiamo fissare la media al valore che vogliamo ma non stringere significativamente la varianza attorno a quel valore, quindi non possiamo essere troppo informativi sul numero di cluster che vogliamo.

## Two levels

I repeat the experiment when  $d = 2$ . Now, the process parameters are  $\gamma_1, \gamma_2, \Lambda$ .

I consider an unbalanced number of points in the two levels,  $n = (100, 50)$ .

### Grid for fixed Lambda

Fisso  $\Lambda = 6$

```
n_j = c(100,50)
Lambda = 6

Npoints = 20
gamma1_grid = seq(0.01,1,length.out = Npoints)
gamma2_grid = seq(0.01,1,length.out = Npoints)

#Lambda_grid = seq(2,10,length.out = Npoints)

Exp_val = matrix(nrow = Npoints,ncol = Npoints)
Var      = matrix(nrow = Npoints,ncol = Npoints)

for(i in 1:Npoints){
  for(j in 1:Npoints){
    Moments = Expected_prior(n_j = n_j, gamma = c(gamma1_grid[i],gamma2_grid[j]),
                          type = "distinct",
                          prior = "Poisson", lambda = Lambda )

    Exp_val[i,j] = Moments$Mean
    Var[i,j]     = Moments$Variance
  }
}

gamma1_names = unlist( lapply( as.list(as.character(round(gamma1_grid, digits = 2))),
                             FUN = function(x){ paste0( "gamma1", "=",x) }
                           )
                     )
gamma2_names = unlist( lapply( as.list(as.character(round(gamma2_grid, digits = 2))),
                             FUN = function(x){ paste0( "gamma2", "=",x) }
                           )
                     )

rownames(Exp_val) <- rownames(Var) <- gamma1_names
colnames(Exp_val) <- colnames(Var) <- gamma2_names
```

Guardo la media

```
#Exp_val
round(Exp_val, digits = 3)

##          gamma2=0.01 gamma2=0.06 gamma2=0.11 gamma2=0.17 gamma2=0.22
```

## gamma1=0.01	2.298	3.147	3.744	4.194	4.546
## gamma1=0.06	3.284	3.955	4.426	4.781	5.060
## gamma1=0.11	3.970	4.516	4.901	5.190	5.417
## gamma1=0.17	4.478	4.933	5.253	5.493	5.682
## gamma1=0.22	4.871	5.254	5.524	5.727	5.886
## gamma1=0.27	5.181	5.509	5.738	5.912	6.047
## gamma1=0.32	5.431	5.713	5.912	6.061	6.178
## gamma1=0.37	5.636	5.881	6.053	6.183	6.285
## gamma1=0.43	5.806	6.021	6.171	6.284	6.373
## gamma1=0.48	5.948	6.137	6.270	6.369	6.447
## gamma1=0.53	6.069	6.236	6.353	6.441	6.510
## gamma1=0.58	6.171	6.320	6.424	6.502	6.564
## gamma1=0.64	6.259	6.392	6.485	6.555	6.610
## gamma1=0.69	6.335	6.454	6.538	6.600	6.650
## gamma1=0.74	6.401	6.508	6.583	6.640	6.684
## gamma1=0.79	6.458	6.555	6.623	6.674	6.714
## gamma1=0.84	6.509	6.597	6.658	6.704	6.741
## gamma1=0.9	6.553	6.633	6.689	6.731	6.764
## gamma1=0.95	6.592	6.665	6.716	6.754	6.784
## gamma1=1	6.627	6.694	6.740	6.775	6.803
##	gamma2=0.27	gamma2=0.32	gamma2=0.37	gamma2=0.43	gamma2=0.48
## gamma1=0.01	4.830	5.064	5.259	5.424	5.566
## gamma1=0.06	5.284	5.469	5.623	5.754	5.865
## gamma1=0.11	5.599	5.750	5.876	5.982	6.073
## gamma1=0.17	5.834	5.959	6.063	6.152	6.227
## gamma1=0.22	6.014	6.119	6.208	6.282	6.346
## gamma1=0.27	6.157	6.247	6.322	6.386	6.440
## gamma1=0.32	6.272	6.350	6.414	6.469	6.516
## gamma1=0.37	6.366	6.434	6.490	6.538	6.579
## gamma1=0.43	6.445	6.504	6.553	6.595	6.631
## gamma1=0.48	6.510	6.562	6.606	6.642	6.674
## gamma1=0.53	6.566	6.612	6.650	6.683	6.711
## gamma1=0.58	6.613	6.654	6.688	6.717	6.742
## gamma1=0.64	6.654	6.691	6.721	6.747	6.769
## gamma1=0.69	6.689	6.722	6.749	6.772	6.792
## gamma1=0.74	6.720	6.749	6.774	6.795	6.813
## gamma1=0.79	6.746	6.773	6.795	6.814	6.830
## gamma1=0.84	6.770	6.794	6.814	6.831	6.846
## gamma1=0.9	6.790	6.812	6.831	6.846	6.859
## gamma1=0.95	6.809	6.829	6.845	6.859	6.872
## gamma1=1	6.825	6.843	6.858	6.871	6.882
##	gamma2=0.53	gamma2=0.58	gamma2=0.64	gamma2=0.69	gamma2=0.74
## gamma1=0.01	5.688	5.794	5.888	5.970	6.043
## gamma1=0.06	5.962	6.046	6.120	6.185	6.242
## gamma1=0.11	6.152	6.220	6.281	6.334	6.381
## gamma1=0.17	6.293	6.350	6.400	6.444	6.483
## gamma1=0.22	6.402	6.450	6.492	6.529	6.562
## gamma1=0.27	6.488	6.529	6.565	6.597	6.625
## gamma1=0.32	6.557	6.593	6.624	6.651	6.675
## gamma1=0.37	6.614	6.645	6.672	6.696	6.717
## gamma1=0.43	6.661	6.688	6.712	6.733	6.752
## gamma1=0.48	6.701	6.725	6.746	6.764	6.780
## gamma1=0.53	6.735	6.756	6.774	6.791	6.805
## gamma1=0.58	6.764	6.782	6.799	6.813	6.826

## gamma1=0.64	6.788	6.805	6.820	6.833	6.844
## gamma1=0.69	6.810	6.824	6.838	6.849	6.860
## gamma1=0.74	6.828	6.842	6.853	6.864	6.873
## gamma1=0.79	6.844	6.856	6.867	6.877	6.885
## gamma1=0.84	6.858	6.870	6.879	6.888	6.895
## gamma1=0.9	6.871	6.881	6.890	6.898	6.905
## gamma1=0.95	6.882	6.891	6.899	6.906	6.913
## gamma1=1	6.892	6.900	6.908	6.914	6.920
##	gamma2=0.79	gamma2=0.84	gamma2=0.9	gamma2=0.95	gamma2=1
## gamma1=0.01	6.108	6.166	6.218	6.266	6.309
## gamma1=0.06	6.294	6.340	6.381	6.418	6.452
## gamma1=0.11	6.422	6.460	6.494	6.524	6.552
## gamma1=0.17	6.518	6.549	6.577	6.603	6.626
## gamma1=0.22	6.592	6.618	6.642	6.663	6.683
## gamma1=0.27	6.650	6.673	6.693	6.711	6.728
## gamma1=0.32	6.697	6.717	6.734	6.750	6.764
## gamma1=0.37	6.736	6.753	6.768	6.782	6.794
## gamma1=0.43	6.768	6.783	6.796	6.808	6.819
## gamma1=0.48	6.795	6.808	6.820	6.831	6.840
## gamma1=0.53	6.818	6.830	6.840	6.849	6.858
## gamma1=0.58	6.837	6.848	6.857	6.865	6.873
## gamma1=0.64	6.854	6.864	6.872	6.879	6.886
## gamma1=0.69	6.869	6.877	6.885	6.891	6.897
## gamma1=0.74	6.881	6.889	6.896	6.902	6.907
## gamma1=0.79	6.893	6.899	6.905	6.911	6.916
## gamma1=0.84	6.902	6.908	6.914	6.919	6.923
## gamma1=0.9	6.911	6.916	6.921	6.926	6.930
## gamma1=0.95	6.918	6.924	6.928	6.932	6.936
## gamma1=1	6.925	6.930	6.934	6.938	6.941

- L'impatto di  $\gamma_1$  e  $\gamma_2$  è un'estensione di quello che abbiamo visto nel caso  $d = 1$ .
- $E[K]$  cresce più velocemente con  $\gamma_1$  che con  $\gamma_2$ , ovvero con il livello più numeroso.
- Un'importante differenza è che ora l'upper bound per  $E[K]$  sembra essere  $\Lambda + 1$  e non solo  $\Lambda$

Guardo anche la varianza

```
#Var
round(Var, digits = 3)
```

##	gamma2=0.01	gamma2=0.06	gamma2=0.11	gamma2=0.17	gamma2=0.22
## gamma1=0.01	0.604	1.405	1.930	2.314	2.614
## gamma1=0.06	1.552	2.213	2.645	2.962	3.211
## gamma1=0.11	2.184	2.752	3.123	3.397	3.612
## gamma1=0.17	2.651	3.151	3.478	3.721	3.912
## gamma1=0.22	3.018	3.464	3.759	3.977	4.149
## gamma1=0.27	3.316	3.721	3.988	4.186	4.342
## gamma1=0.32	3.567	3.935	4.179	4.361	4.505
## gamma1=0.37	3.780	4.119	4.343	4.511	4.643
## gamma1=0.43	3.965	4.277	4.485	4.640	4.762
## gamma1=0.48	4.127	4.416	4.609	4.753	4.866
## gamma1=0.53	4.270	4.539	4.718	4.852	4.958
## gamma1=0.58	4.398	4.648	4.815	4.940	5.039
## gamma1=0.64	4.512	4.746	4.902	5.019	5.111
## gamma1=0.69	4.615	4.833	4.980	5.089	5.175
## gamma1=0.74	4.708	4.913	5.050	5.152	5.233
## gamma1=0.79	4.792	4.984	5.113	5.210	5.286

##	gamma1=0.84	4.869	5.050	5.171	5.262	5.333
##	gamma1=0.9	4.939	5.109	5.223	5.309	5.376
##	gamma1=0.95	5.003	5.164	5.271	5.352	5.415
##	gamma1=1	5.062	5.213	5.315	5.391	5.451
##	gamma2=0.27	gamma2=0.32	gamma2=0.37	gamma2=0.43	gamma2=0.48	
##	gamma1=0.01	2.859	3.065	3.241	3.395	3.532
##	gamma1=0.06	3.414	3.585	3.732	3.861	3.974
##	gamma1=0.11	3.788	3.937	4.065	4.176	4.275
##	gamma1=0.17	4.068	4.201	4.314	4.413	4.501
##	gamma1=0.22	4.290	4.409	4.511	4.601	4.680
##	gamma1=0.27	4.471	4.579	4.672	4.754	4.825
##	gamma1=0.32	4.622	4.722	4.807	4.881	4.946
##	gamma1=0.37	4.751	4.843	4.921	4.989	5.049
##	gamma1=0.43	4.862	4.947	5.019	5.082	5.137
##	gamma1=0.48	4.959	5.037	5.104	5.162	5.213
##	gamma1=0.53	5.044	5.117	5.179	5.233	5.280
##	gamma1=0.58	5.119	5.187	5.245	5.295	5.339
##	gamma1=0.64	5.186	5.249	5.303	5.350	5.391
##	gamma1=0.69	5.246	5.305	5.355	5.399	5.437
##	gamma1=0.74	5.299	5.355	5.402	5.443	5.478
##	gamma1=0.79	5.348	5.400	5.444	5.482	5.516
##	gamma1=0.84	5.391	5.440	5.482	5.518	5.549
##	gamma1=0.9	5.431	5.477	5.516	5.550	5.579
##	gamma1=0.95	5.467	5.510	5.547	5.579	5.607
##	gamma1=1	5.500	5.541	5.576	5.606	5.632
##	gamma2=0.53	gamma2=0.58	gamma2=0.64	gamma2=0.69	gamma2=0.74	
##	gamma1=0.01	3.653	3.763	3.863	3.954	4.037
##	gamma1=0.06	4.076	4.168	4.251	4.326	4.396
##	gamma1=0.11	4.364	4.443	4.515	4.581	4.641
##	gamma1=0.17	4.579	4.650	4.713	4.771	4.824
##	gamma1=0.22	4.750	4.813	4.870	4.922	4.969
##	gamma1=0.27	4.889	4.946	4.997	5.044	5.087
##	gamma1=0.32	5.004	5.056	5.103	5.146	5.184
##	gamma1=0.37	5.102	5.150	5.193	5.231	5.267
##	gamma1=0.43	5.186	5.230	5.269	5.305	5.337
##	gamma1=0.48	5.259	5.299	5.335	5.368	5.398
##	gamma1=0.53	5.322	5.359	5.393	5.423	5.451
##	gamma1=0.58	5.378	5.412	5.443	5.471	5.497
##	gamma1=0.64	5.427	5.459	5.488	5.514	5.538
##	gamma1=0.69	5.471	5.501	5.528	5.552	5.574
##	gamma1=0.74	5.510	5.538	5.563	5.585	5.606
##	gamma1=0.79	5.545	5.571	5.595	5.616	5.635
##	gamma1=0.84	5.577	5.601	5.623	5.643	5.661
##	gamma1=0.9	5.605	5.628	5.649	5.667	5.684
##	gamma1=0.95	5.631	5.653	5.672	5.689	5.705
##	gamma1=1	5.655	5.675	5.693	5.710	5.724
##	gamma2=0.79	gamma2=0.84	gamma2=0.9	gamma2=0.95	gamma2=1	
##	gamma1=0.01	4.114	4.185	4.252	4.313	4.370
##	gamma1=0.06	4.460	4.519	4.574	4.625	4.673
##	gamma1=0.11	4.696	4.747	4.794	4.838	4.879
##	gamma1=0.17	4.873	4.918	4.960	4.998	5.034
##	gamma1=0.22	5.013	5.053	5.090	5.124	5.156
##	gamma1=0.27	5.126	5.162	5.195	5.226	5.254
##	gamma1=0.32	5.220	5.253	5.283	5.310	5.336

## gamma1=0.37	5.299	5.329	5.356	5.381	5.405
## gamma1=0.43	5.367	5.394	5.419	5.442	5.463
## gamma1=0.48	5.425	5.450	5.473	5.494	5.513
## gamma1=0.53	5.476	5.499	5.520	5.539	5.557
## gamma1=0.58	5.520	5.541	5.561	5.578	5.595
## gamma1=0.64	5.559	5.579	5.597	5.613	5.629
## gamma1=0.69	5.594	5.612	5.629	5.644	5.658
## gamma1=0.74	5.624	5.641	5.657	5.671	5.684
## gamma1=0.79	5.652	5.668	5.682	5.696	5.708
## gamma1=0.84	5.677	5.692	5.705	5.718	5.729
## gamma1=0.9	5.699	5.713	5.726	5.737	5.748
## gamma1=0.95	5.719	5.732	5.744	5.755	5.765
## gamma1=1	5.738	5.750	5.761	5.771	5.781

Grid for fixed  $\gamma_1 = \gamma_2$

```
n_j = c(100,50)

Npoints1 = 30
Npoints2 = 10
Lambda_grid = seq(1,10,length.out = Npoints2)
gamma_grid = seq(0.01,1,length.out = Npoints1)

#Lambda_grid = seq(2,10,length.out = Npoints)

Exp_val = matrix(nrow = Npoints1,ncol = Npoints2)
Var      = matrix(nrow = Npoints1,ncol = Npoints2)

for(i in 1:Npoints1){
  for(j in 1:Npoints2){
    Moments = Expected_prior(n_j = n_j, gamma = c(gamma_grid[i],gamma_grid[i]),
                          type = "distinct",
                          prior = "Poisson", lambda = Lambda_grid[j] )

    Exp_val[i,j] = Moments$Mean
    Var[i,j]     = Moments$Variance
  }
}

Lambda_names = unlist( lapply( as.list(as.character(round(Lambda_grid, digits = 1))),
                             FUN = function(x){ paste0( "Lambda", "=",x) }
                           )
                     )
gamma2_names = unlist( lapply( as.list(as.character(round(gamma_grid, digits = 2))),
                             FUN = function(x){ paste0( "gamma1=gamma2", "=",x) }
                           )
                     )

rownames(Exp_val) <- rownames(Var) <- gamma2_names
colnames(Exp_val) <- colnames(Var) <- Lambda_names
```

Guardo la media

```
#Exp_val
round(Exp_val, digits = 3)
```

	Lambda=1	Lambda=2	Lambda=3	Lambda=4	Lambda=5	Lambda=6
##						
## gamma1=gamma2=0.01	1.426	1.698	1.893	2.048	2.180	2.298
## gamma1=gamma2=0.04	1.581	2.045	2.448	2.816	3.162	3.494
## gamma1=gamma2=0.08	1.689	2.286	2.830	3.341	3.829	4.301
## gamma1=gamma2=0.11	1.767	2.457	3.101	3.714	4.304	4.875
## gamma1=gamma2=0.15	1.823	2.581	3.299	3.987	4.652	5.298
## gamma1=gamma2=0.18	1.864	2.674	3.446	4.191	4.914	5.617
## gamma1=gamma2=0.21	1.895	2.743	3.558	4.347	5.115	5.863
## gamma1=gamma2=0.25	1.918	2.796	3.644	4.468	5.271	6.056
## gamma1=gamma2=0.28	1.935	2.836	3.711	4.562	5.395	6.209
## gamma1=gamma2=0.32	1.949	2.868	3.763	4.637	5.493	6.332
## gamma1=gamma2=0.35	1.959	2.893	3.804	4.697	5.573	6.432
## gamma1=gamma2=0.39	1.967	2.912	3.838	4.746	5.638	6.514
## gamma1=gamma2=0.42	1.974	2.928	3.864	4.785	5.691	6.582
## gamma1=gamma2=0.45	1.979	2.940	3.886	4.817	5.735	6.638
## gamma1=gamma2=0.49	1.983	2.950	3.904	4.844	5.771	6.686
## gamma1=gamma2=0.52	1.986	2.958	3.919	4.866	5.802	6.725
## gamma1=gamma2=0.56	1.988	2.965	3.931	4.885	5.828	6.759
## gamma1=gamma2=0.59	1.990	2.971	3.941	4.900	5.849	6.788
## gamma1=gamma2=0.62	1.992	2.975	3.949	4.913	5.868	6.813
## gamma1=gamma2=0.66	1.993	2.979	3.956	4.924	5.884	6.834
## gamma1=gamma2=0.69	1.994	2.982	3.962	4.934	5.897	6.852
## gamma1=gamma2=0.73	1.995	2.985	3.967	4.942	5.909	6.868
## gamma1=gamma2=0.76	1.996	2.987	3.971	4.949	5.919	6.882
## gamma1=gamma2=0.8	1.997	2.989	3.975	4.955	5.928	6.894
## gamma1=gamma2=0.83	1.997	2.990	3.978	4.960	5.936	6.904
## gamma1=gamma2=0.86	1.998	2.991	3.981	4.964	5.942	6.914
## gamma1=gamma2=0.9	1.998	2.993	3.983	4.968	5.948	6.922
## gamma1=gamma2=0.93	1.998	2.994	3.985	4.972	5.953	6.929
## gamma1=gamma2=0.97	1.998	2.994	3.987	4.975	5.958	6.936
## gamma1=gamma2=1	1.999	2.995	3.988	4.977	5.962	6.941
##						
	Lambda=7	Lambda=8	Lambda=9	Lambda=10		
## gamma1=gamma2=0.01	2.408	2.512	2.612	2.709		
## gamma1=gamma2=0.04	3.814	4.126	4.431	4.730		
## gamma1=gamma2=0.08	4.759	5.206	5.642	6.070		
## gamma1=gamma2=0.11	5.431	5.974	6.505	7.025		
## gamma1=gamma2=0.15	5.927	6.542	7.144	7.734		
## gamma1=gamma2=0.18	6.304	6.975	7.632	8.277		
## gamma1=gamma2=0.21	6.595	7.311	8.014	8.703		
## gamma1=gamma2=0.25	6.824	7.577	8.316	9.042		
## gamma1=gamma2=0.28	7.008	7.791	8.561	9.317		
## gamma1=gamma2=0.32	7.156	7.965	8.760	9.542		
## gamma1=gamma2=0.35	7.277	8.107	8.924	9.728		
## gamma1=gamma2=0.39	7.377	8.225	9.061	9.884		
## gamma1=gamma2=0.42	7.460	8.324	9.176	10.015		
## gamma1=gamma2=0.45	7.529	8.407	9.273	10.127		
## gamma1=gamma2=0.49	7.588	8.477	9.356	10.222		
## gamma1=gamma2=0.52	7.637	8.537	9.426	10.304		
## gamma1=gamma2=0.56	7.680	8.589	9.487	10.375		
## gamma1=gamma2=0.59	7.716	8.633	9.540	10.436		
## gamma1=gamma2=0.62	7.747	8.672	9.586	10.490		

```
## gamma1=gamma2=0.66    7.774    8.705    9.626    10.537
## gamma1=gamma2=0.69    7.798    8.734    9.661    10.579
## gamma1=gamma2=0.73    7.818    8.760    9.692    10.616
## gamma1=gamma2=0.76    7.836    8.782    9.719    10.648
## gamma1=gamma2=0.8     7.852    8.802    9.744    10.677
## gamma1=gamma2=0.83    7.866    8.820    9.765    10.703
## gamma1=gamma2=0.86    7.878    8.835    9.785    10.726
## gamma1=gamma2=0.9     7.889    8.849    9.802    10.747
## gamma1=gamma2=0.93    7.899    8.861    9.817    10.766
## gamma1=gamma2=0.97    7.907    8.873    9.831    10.783
## gamma1=gamma2=1       7.915    8.883    9.844    10.798
```

Il comportamento è simile a prima ma lo spazio di manovra sembra molto più limitato, i valori convergono molto più velocemente all'upper bound dato da  $E[K]$ .

Guardo la varianza

```
#Var
round(Var, digits = 3)
```

```
##          Lambda=1 Lambda=2 Lambda=3 Lambda=4 Lambda=5 Lambda=6
## gamma1=gamma2=0.01    0.296    0.386    0.437    0.487    0.543    0.604
## gamma1=gamma2=0.04    0.472    0.769    1.026    1.276    1.524    1.772
## gamma1=gamma2=0.08    0.597    1.041    1.438    1.817    2.186    2.544
## gamma1=gamma2=0.11    0.688    1.240    1.739    2.210    2.663    3.098
## gamma1=gamma2=0.15    0.756    1.389    1.965    2.508    3.024    3.519
## gamma1=gamma2=0.18    0.807    1.503    2.140    2.740    3.309    3.852
## gamma1=gamma2=0.21    0.846    1.591    2.279    2.925    3.539    4.123
## gamma1=gamma2=0.25    0.876    1.661    2.390    3.076    3.727    4.348
## gamma1=gamma2=0.28    0.899    1.718    2.481    3.201    3.885    4.537
## gamma1=gamma2=0.32    0.918    1.763    2.555    3.305    4.018    4.698
## gamma1=gamma2=0.35    0.932    1.800    2.617    3.393    4.132    4.837
## gamma1=gamma2=0.39    0.944    1.830    2.669    3.467    4.229    4.957
## gamma1=gamma2=0.42    0.954    1.855    2.712    3.531    4.313    5.062
## gamma1=gamma2=0.45    0.961    1.876    2.749    3.585    4.386    5.153
## gamma1=gamma2=0.49    0.968    1.893    2.781    3.632    4.449    5.234
## gamma1=gamma2=0.52    0.973    1.908    2.807    3.672    4.504    5.305
## gamma1=gamma2=0.56    0.977    1.920    2.830    3.708    4.553    5.368
## gamma1=gamma2=0.59    0.980    1.931    2.850    3.739    4.596    5.424
## gamma1=gamma2=0.62    0.983    1.940    2.867    3.766    4.634    5.473
## gamma1=gamma2=0.66    0.986    1.947    2.882    3.789    4.668    5.518
## gamma1=gamma2=0.69    0.988    1.954    2.895    3.810    4.698    5.558
## gamma1=gamma2=0.73    0.990    1.959    2.906    3.829    4.725    5.594
## gamma1=gamma2=0.76    0.991    1.964    2.916    3.845    4.748    5.626
## gamma1=gamma2=0.8     0.992    1.968    2.925    3.859    4.770    5.655
## gamma1=gamma2=0.83    0.993    1.972    2.933    3.872    4.789    5.681
## gamma1=gamma2=0.86    0.994    1.975    2.939    3.884    4.806    5.705
## gamma1=gamma2=0.9     0.995    1.978    2.945    3.894    4.822    5.727
## gamma1=gamma2=0.93    0.996    1.980    2.951    3.903    4.836    5.746
## gamma1=gamma2=0.97    0.996    1.983    2.955    3.911    4.848    5.764
## gamma1=gamma2=1       0.997    1.984    2.959    3.919    4.860    5.781
##          Lambda=7 Lambda=8 Lambda=9 Lambda=10
## gamma1=gamma2=0.01    0.671    0.740    0.811    0.883
## gamma1=gamma2=0.04    2.017    2.258    2.495    2.727
## gamma1=gamma2=0.08    2.892    3.230    3.559    3.878
## gamma1=gamma2=0.11    3.518    3.923    4.314    4.692
```

## gamma1=gamma2=0.15	3.994	4.450	4.890	5.315
## gamma1=gamma2=0.18	4.372	4.871	5.351	5.814
## gamma1=gamma2=0.21	4.682	5.218	5.732	6.228
## gamma1=gamma2=0.25	4.941	5.509	6.054	6.579
## gamma1=gamma2=0.28	5.160	5.757	6.330	6.881
## gamma1=gamma2=0.32	5.349	5.972	6.570	7.145
## gamma1=gamma2=0.35	5.512	6.159	6.780	7.377
## gamma1=gamma2=0.39	5.655	6.324	6.966	7.583
## gamma1=gamma2=0.42	5.780	6.469	7.130	7.766
## gamma1=gamma2=0.45	5.890	6.597	7.277	7.931
## gamma1=gamma2=0.49	5.988	6.712	7.409	8.079
## gamma1=gamma2=0.52	6.074	6.815	7.527	8.213
## gamma1=gamma2=0.56	6.152	6.907	7.634	8.334
## gamma1=gamma2=0.59	6.221	6.990	7.731	8.444
## gamma1=gamma2=0.62	6.283	7.065	7.818	8.544
## gamma1=gamma2=0.66	6.339	7.132	7.898	8.636
## gamma1=gamma2=0.69	6.390	7.194	7.970	8.720
## gamma1=gamma2=0.73	6.435	7.250	8.037	8.797
## gamma1=gamma2=0.76	6.477	7.300	8.097	8.867
## gamma1=gamma2=0.8	6.514	7.347	8.153	8.932
## gamma1=gamma2=0.83	6.548	7.389	8.204	8.992
## gamma1=gamma2=0.86	6.579	7.428	8.251	9.048
## gamma1=gamma2=0.9	6.608	7.464	8.294	9.099
## gamma1=gamma2=0.93	6.634	7.497	8.334	9.146
## gamma1=gamma2=0.97	6.657	7.527	8.371	9.191
## gamma1=gamma2=1	6.679	7.555	8.405	9.231

## Distribution for $\Lambda$ fixed and equal $\gamma$ s

Fisso  $\Lambda = 4$  e guardo la distribuzione del numero di cluster al variare di  $\gamma_1 = \gamma_2$

```
Npoints = 50
Lambda = 4
gamma_grid = seq(0.01,1,length.out = Npoints)
prob_k = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){
  #threshold = 0.995
  for(kk in 1:max(n_j) ){
    prob_k[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = c(gamma_grid[i],gamma_grid[i]),
                                     prior = "Poisson",
                                     lambda = Lambda)

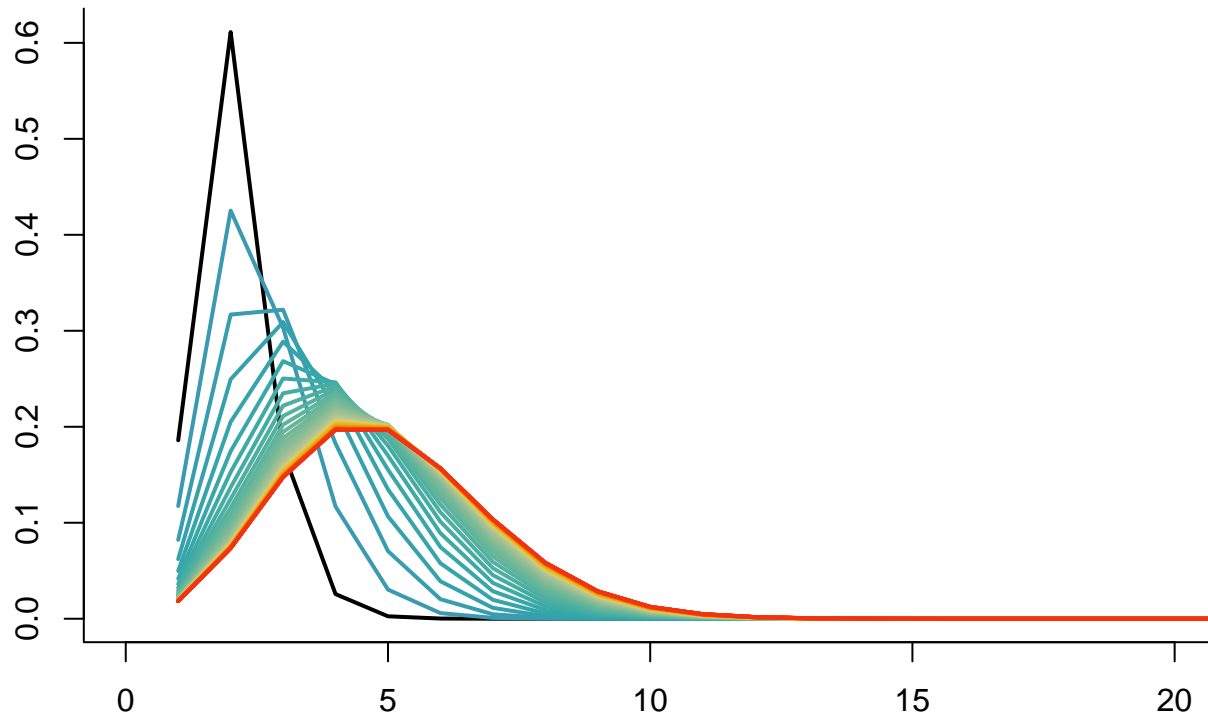
    #if(sum(prob_k)>=threshold)
    #break
  }
}
```

La curva nera è la prima, quella con  $\gamma$  minore

```
par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k), type = "l", lwd = 2,xlim = c(0,20),lty = 1,
         col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
         main = paste0( paste0( "Distribution of K for Lambda=4 and gamma varying") ))
```



## Distribution of K for Lambda=4 and gamma varying



Ripeto per  $\Lambda = 10$

```
Npoints = 50
Lambda = 10
gamma_grid = seq(0.01,1,length.out = Npoints)
prob_k = matrix(nrow = Npoints, ncol = n_j)

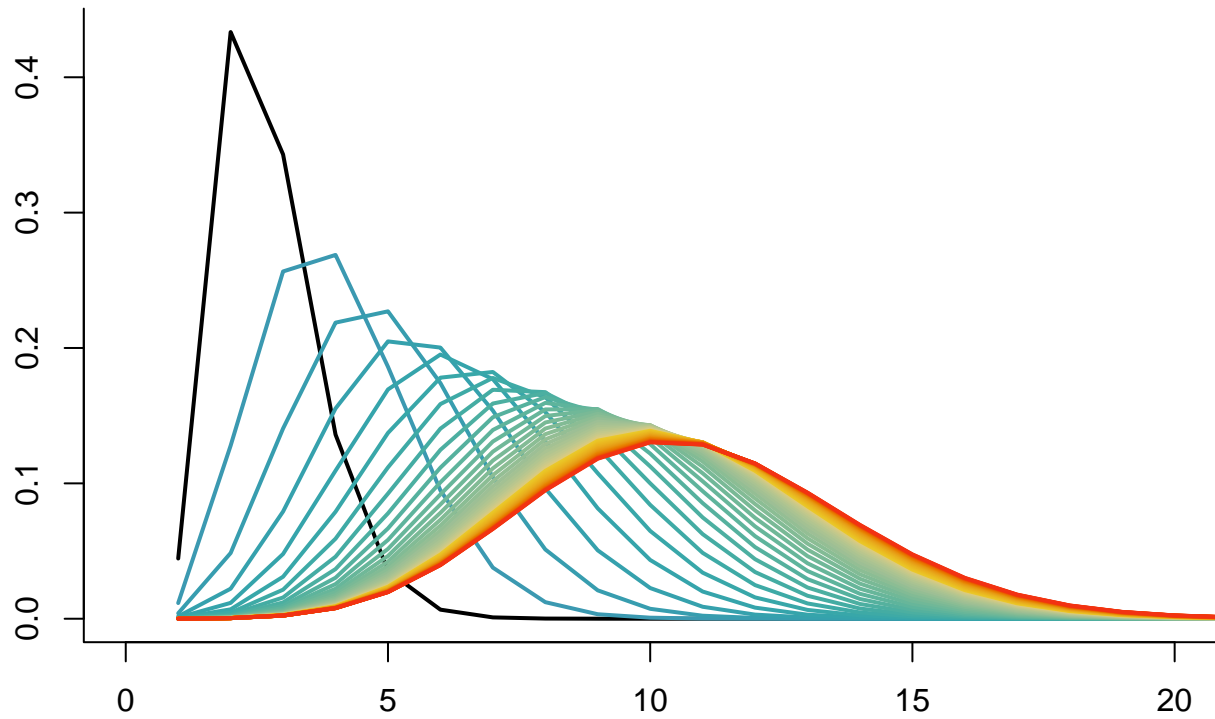
for(i in 1:Npoints){
  #threshold = 0.995
  for(kk in 1:max(n_j) ){
    prob_k[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                   gamma = c(gamma_grid[i],gamma_grid[i]),
                                   prior = "Poisson",
                                   lambda = Lambda)

    #if(sum(prob_k)>=threshold)
    #break
  }
}
```

La curva nera è la prima, quella con  $\gamma$  minore

```
par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k), type = "l", lwd = 2,xlim = c(0,20),lty = 1,
         col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
         main = paste0( paste0( "Distribution of K for Lambda=10 and gamma varying") ))
```

## Distribution of K for Lambda=10 and gamma varying



## Distribution for $\gamma$ fixed

Ora faccio l'esperimento inverso, tengo fisso  $\gamma$  e faccio variare  $\Lambda$ .

`gamma1 = gamma2 = 2`

```
Npoints = 20
gamma1 <- gamma2 <- 2
Lambda_grid = seq(1,20,length.out = Npoints)
prob_k3 = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){

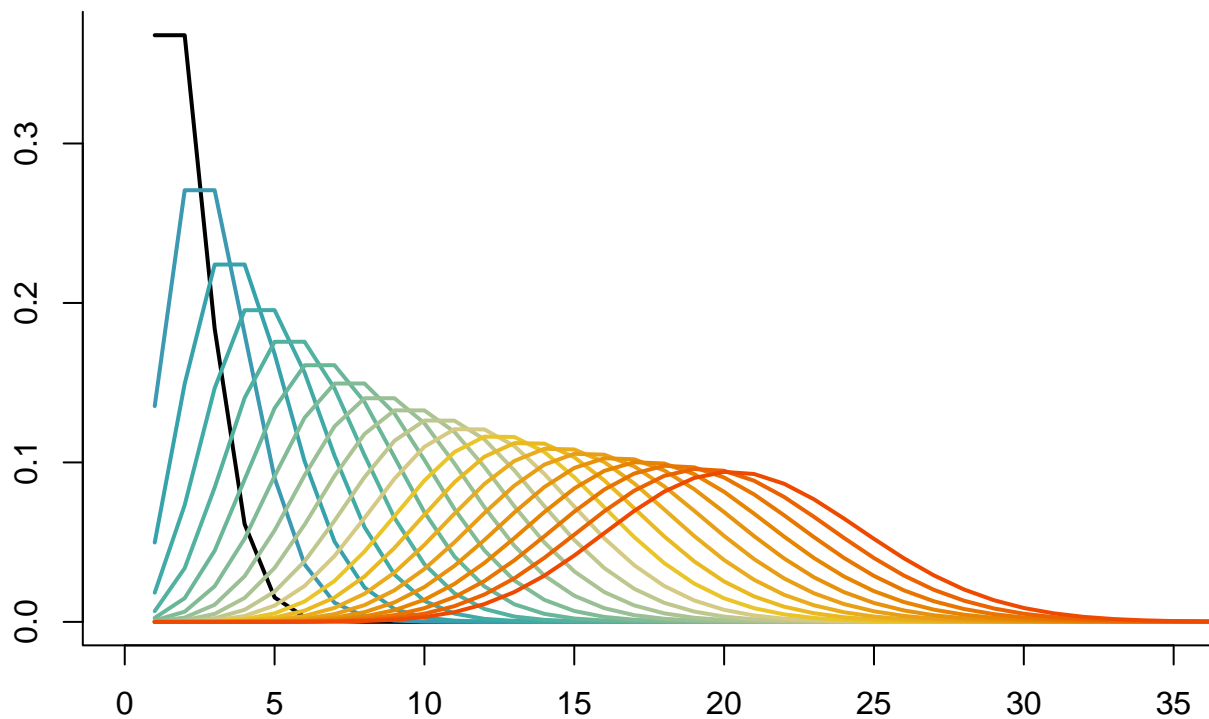
  #threshold = 0.995
  for(kk in 1:max(n_j) ){
    prob_k3[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = c(gamma1,gamma2),
                                     prior = "Poisson",
                                     lambda = Lambda_grid[i])

    #if(sum(prob_k)>=threshold)
    #break
  }
}
```

La curva nera è la prima, quella con  $\Lambda$  minore. Ogni curva, corrisponde ad aumentare  $\Lambda$  di 1.

```
par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k3), type = "l", lwd = 2,xlim = c(0,35),lty = 1,
        col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
        main = paste0( paste0( "Distribution of K for gamma1=gamma2=2 and Lambda varying") ))
```

## Distribution of K for gamma1=gamma2=2 and Lambda varying



$\gamma_1 = \gamma_2 = 0.1$

Ripeto lo stesso esperimento abbassando il valore di  $\gamma_1$  e  $\gamma_2$ . Ogni curva, corrisponde ad aumentare  $\Lambda$  di 1.

```
Npoints = 20
gamma1 <- gamma2 <- 0.1
Lambda_grid = seq(1,20,length.out = Npoints)
prob_k4 = matrix(nrow = Npoints, ncol = n_j)

for(i in 1:Npoints){
  #threshold = 0.995
  for(kk in 1:max(n_j)) {
    prob_k4[i,kk] = p_distinct_prior(k = kk, n_j = n_j,
                                     gamma = c(gamma1,gamma2),
                                     prior = "Poisson",
                                     lambda = Lambda_grid[i])

    #if(sum(prob_k)>=threshold)
    #break
  }
}
```

La curva nera è la prima, quella con  $\Lambda$  minore. Ogni curva, corrisponde ad aumentare  $\Lambda$  di 1.

```
par(mar = c(2,2,4,1), bty = "l")
matplot( t(prob_k4), type = "l", lwd = 2,xlim = c(0,35),lty = 1,
        col = c("black",hcl.colors(n=Npoints, palette = "Zissou 1")),
        main = paste0( paste0( "Distribution of K for gamma1=gamma1=0.1 and Lambda varying") ))
```

