

# Laplace Approximation

The following section sets up the equations required to compute the exact and the Laplace approximations of the conditional probabilities

```

SetOptions[$FrontEndSession, EvaluationCompletionAction -> "ShowTiming"]

(* Exact probability *)
p[y_, Inp_, k_, V_] := 
$$e^{\frac{2 V y (Inp^2 k^2 - \alpha^2)}{(-Inp k + \alpha)^2}} e^{-\text{Log}\left[1 + \frac{(\alpha - Inp k) y}{Inp k 0_{tot}}\right]} e^{\frac{4 Inp k 0_{tot} \alpha V}{(\alpha - Inp k)^2} \text{Log}\left[1 + \frac{(\alpha - Inp k) y}{Inp k 0_{tot}}\right]}$$
;
(* I checked that this probability is the correct solution to FP equation *)
(* define f(x) and find max 0_max. f(x) is the
   exponent in the conditional probability as V -> infinity *)
f[y_, Inp_, k_] := 
$$\frac{4 Inp k 0_{tot} \alpha}{(\alpha - Inp k)^2} \text{Log}\left[1 + \frac{(\alpha - Inp k) y}{Inp k 0_{tot}}\right] - \frac{2 y (Inp k + \alpha)}{\alpha - Inp k}$$

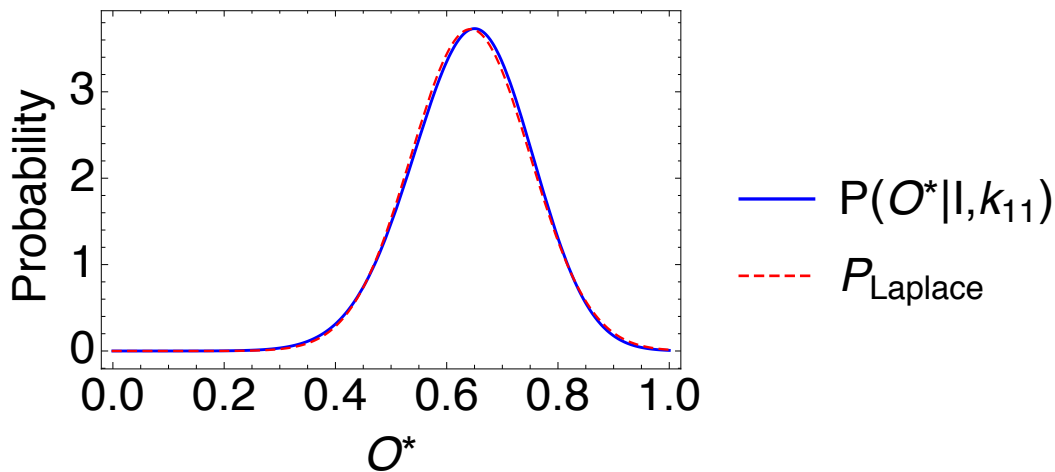
(*Solve[D[f[y,Inp,k],y]==0,y]; *)
(* the following with the linear correction works really well *)
(*fDoublePrimeyMax[Inp_,k_] := -
$$\frac{2 (\alpha + Inp k)}{Inp k k 0_{tot}} - Inp k$$
; *)
fDoublePrimeyMax[Inp_, k_] := -
$$\frac{2}{0_{tot}} - \frac{Inp k}{\alpha 0_{tot}} - \frac{\alpha}{Inp k 0_{tot}}$$
;
(*fDoublePrimeyMax[Inp_,k_] := -
$$\frac{2 (\alpha + Inp k)}{Inp k k 0_{tot}}$$
; *)
fMax[Inp_, k_] := 
$$\frac{-2 Inp k}{\alpha - Inp k} + \frac{(4 Inp k \alpha 0_{tot})}{(\alpha - Inp k)^2} \text{Log}\left[\frac{2 \alpha}{\alpha - Inp k}\right]$$
;
Omax[Inp_, k_] := 
$$\frac{Inp k 0_{tot}}{Inp k + \alpha}$$
;
(* location of the mean of the conditional probability *)
(* Define the laplace approximation of the exact probability *)
(* works okay for boundary near 1 *)
(*pLaplace[y_,Inp_,k_,V_] :=
   Exp[V(
$$\frac{1}{2/(3+Inp)}$$
fDoublePrimeyMax[Inp,k] (y-Omax[Inp,k])^2)] *)
pLaplace[y_, Inp_, k_, V_] := Exp[V(
$$\frac{1}{2}$$
fDoublePrimeyMax[Inp, k] (y - Omax[Inp, k])^2)]
pLaplace[y, x, k, V]
(* Some sample parameter choices *)
0_tot = 1;  $\alpha$  = 1;
k = 2.001;
Inp = 0.9;
V = 20;

(* the following are the normalizations for one input *)
NormPLap = NIntegrate[pLaplace[y, Inp, k, V], {y, 0, 1}];
NormPCond = NIntegrate[p[y, Inp, k, V], {y, 0, 1}];
(* Recreate probability comparison *)

```

```
Plot[{ $\frac{p[y, \text{Inp}, k, V]}{\text{NormPCond}}$ ,  $\frac{p\text{Laplace}[y, \text{Inp}, k, V]}{\text{NormPLap}}$ }, {y, 0.0, 1},
PlotRange → All, PlotLegends → {"P(O*|I, k11)", "PLaplace" },
PlotStyle → {{Thickness[0.005], Blue}, {Red, Dashed, Thickness[0.004]}},
{Thickness[0.005], Green}}, TicksStyle → Directive["Label", 30],
Ticks → {Automatic}, LabelStyle → {FontFamily → "Helvetica", FontSize → 24},
FrameLabel → {"O*", "Probability "}, Frame → True, FrameStyle → Directive[Black]]
```

$$e^{\frac{1}{2} V \left( -\frac{2}{O_{\text{tot}}} - \frac{k x}{\alpha O_{\text{tot}}} - \frac{\alpha}{k x O_{\text{tot}}} \right) \left( y - \frac{k x O_{\text{tot}}}{k x + \alpha} \right)^2}$$



(\* Reproduce corrected conditional probability in java. \*)

```
NIntegrate[pLaplace[y, 0.505, k, V], {y, 0, 1}]
```

0.274273

```
pLaplace[0.505, 0.505, 3, 20]
```

```
pLaplace[0.055, 0.055, 3, 20]
```

```
NIntegrate[pLaplace[y, 0.055, k, V], {y, 0, 1}]
```

2.16923

## Plot Probability Maxima

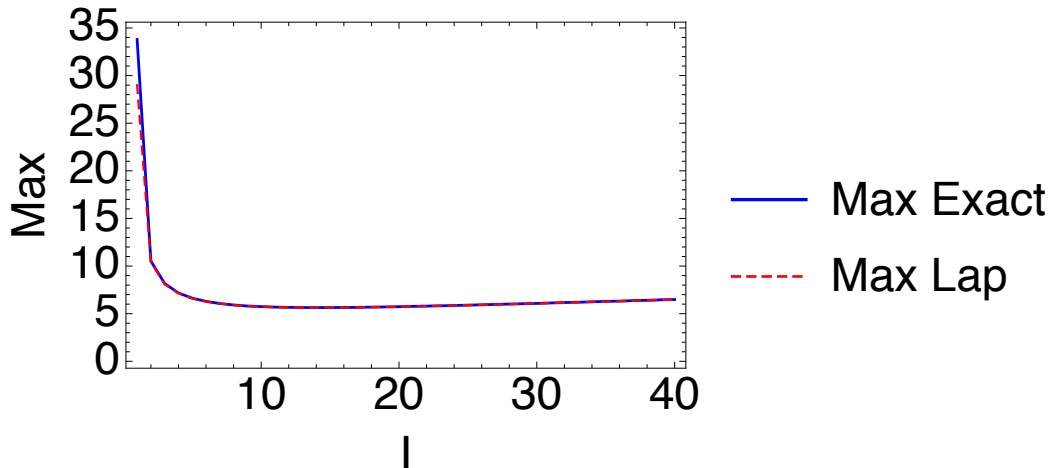
(\* This shows that the maxima disagrees a lot near the boundary y = 0 unless include the linear correction in the variance or fDoublePrimeyMax = -k\*input \*)

```

maxPy[k_, Inp_] := Max[Function[y,  $\frac{p[y, Inp, k, V]}{NIntegrate[p[yp, Inp, k, V], \{yp, 0, 1\}]}$ ] /@
  Range[0.005, 0.995, 0.01]];
maxPyLap[k_, Inp_] := Max[Function[y,
   $\frac{pLaplace[y, Inp, k, V]}{NIntegrate[pLaplace[yp, Inp, k, V], \{yp, 0, 1\}]}$ ] /@ Range[0.005, 0.995, 0.01]];
maxPyList[k_] := Function[Inp, maxPy[k, Inp]] /@ Range[0.005, 0.995, 0.025];
maxPyLapList[k_] := Function[Inp, maxPyLap[k, Inp]] /@ Range[0.005, 0.995, 0.025];

ListLinePlot[{maxPyList[3], maxPyLapList[3]},
  PlotLegends → {"Max Exact", "Max Lap"},
  PlotStyle → {{Thickness[0.005], Blue}, {Red, Dashed, Thickness[0.004]}},
  TicksStyle → Directive["Label", 30], Ticks → {Automatic},
  LabelStyle → {FontFamily → "Helvetica", FontSize → 24}, FrameLabel → {"I", "Max"},
  Frame → True, FrameStyle → Directive[Black], PlotRange → All]

```



## Get Output Probability

```

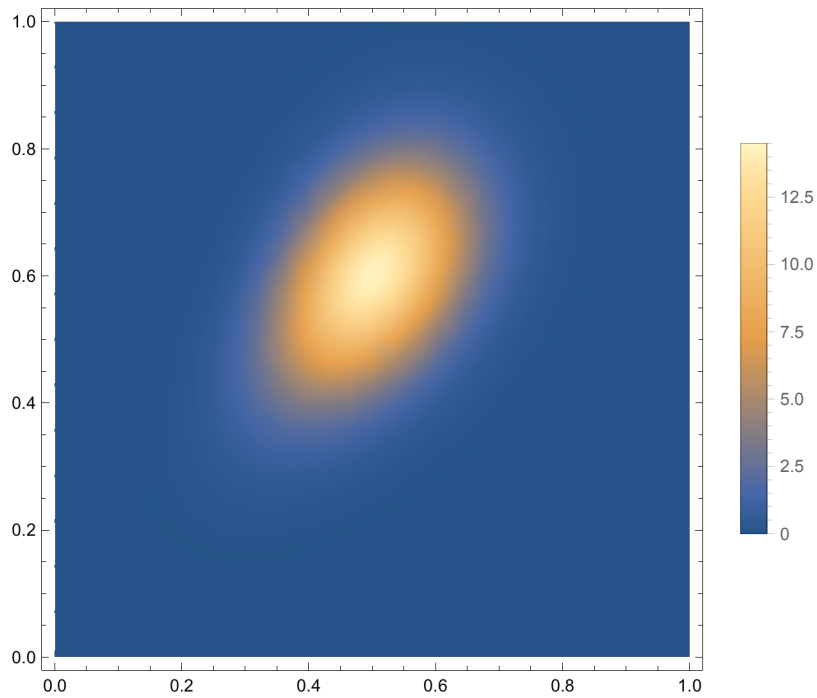
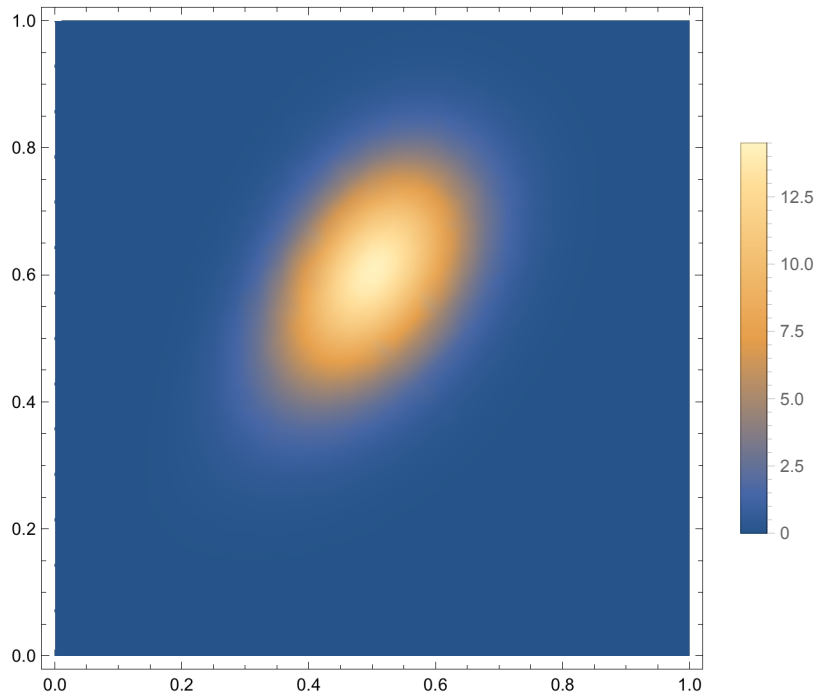
pInput[Inp_, mu_, sigma_] := Exp[ $\frac{-(\text{Inp} - \mu)^2}{2 \text{sigma}^2}$ ] *  $\frac{1}{(2 \text{Pi} \text{sigma}^2)^{\frac{1}{2}}}$ 

(* This probability is normalized *)
dx = 0.1;
pLaplaceNormalized[y_, Inp_, k_, V_] :=
  Exp[V ( $\frac{1}{2}$  fDoublePrimeyMax[Inp, k] (y - Omax[Inp, k])2)]
  (NIntegrate[pLaplace[yp, Inp, k, V], {yp, 0.0001, 1}])-1
pOutputLap[y_, k_, V_] := Sum[pLaplaceNormalized[y, Inp, k, V] *
  pInput[Inp, 0.5, 0.1] * dx, {Inp, 0.0001, 1, dx}]
pOutputLap1[y_, k_, V_] := Sum[pLaplace[y, Inp, k, V] * pInput[Inp, 0.5, 0.1],
  {Inp, 0.0001, 1, dx}]
(* Sum[pOutputLap[y,k,V]*0.1,{y,0,1,0.1}] *)
pNormalized[y_, Inp_, k_, V_] :=
  p[y, Inp, k, V] (NIntegrate[p[yp, Inp, k, V], {yp, 0.0001, 1}])-1
pOutput[y_, k_, V_] :=
  Sum[pNormalized[y, Inp, k, V] * pInput[Inp, 0.5, 0.1] * dx, {Inp, 0.0001, 1, dx}]
pOutput1[y_, k_, V_] :=
  Sum[p[y, Inp, k, V] * pInput[Inp, 0.5, 0.1] * dx, {Inp, 0.0001, 1, dx}]
Sum[pNormalized[y, 0.5, 1.5, 100] * 0.1, {y, 0, 1, 0.1}]

pJoint[y_, Inp_, k_, V_] := pNormalized[y, Inp, k, V] * pInput[Inp, 0.5, 0.1]
pJointLaplace[y_, Inp_, k_, V_] :=
  pLaplaceNormalized[y, Inp, k, V] * pInput[Inp, 0.5, 0.1]
(* check to see if joint probabilities look similar *)
DensityPlot[{pJoint[y, x, k, V]}, {x, 0, 1},
  {y, 0, 1}, PlotRange -> All, PlotLegends -> Automatic]
DensityPlot[{pJointLaplace[y, x, k, V]}, {x, 0, 1},
  {y, 0, 1}, PlotRange -> All, PlotLegends -> Automatic]

```

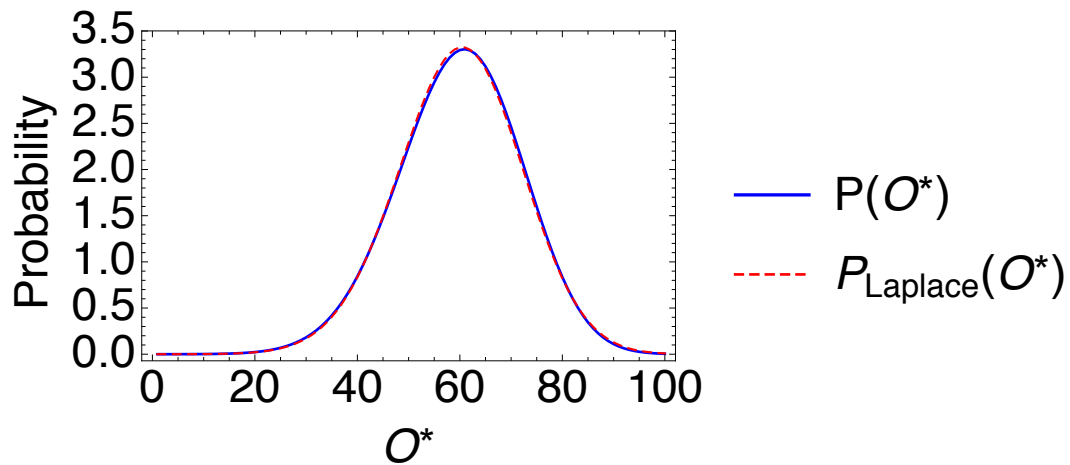
0.998724



Check that out probabilities match

```
lapList = Function[x, pOutputLap[x, k, V]] /@Range[0.005, 0.995, 0.01];
exactList = Function[x, pOutput[x, k, V]] /@Range[0.005, 0.995, 0.01];
```

```
ListLinePlot[{exactList, lapList}, PlotLegends → {"P(O*)", "PLaplace(O*)"},
  PlotStyle → {{Thickness[0.005], Blue}, {Red, Dashed, Thickness[0.004]}},
  TicksStyle → Directive["Label", 30], Ticks → {Automatic},
  LabelStyle → {FontFamily → "Helvetica", FontSize → 24},
  FrameLabel → {"O*", "Probability "}, Frame → True,
  FrameStyle → Directive[Black], PlotRange → All]
```




---

## Compute Mutual Information

(\* Use java program to compute these integrals \*)

```

MI[k_] := Sum[pJoint[y, x, k, V] * Log[ $\frac{pJoint[y, x, k, V]}{pOutput[y, k, V] pInput[x, 0.5, 0.1]}$ ] * 0.1 * 0.1,
  {x, 0.01, 1, 0.1}, {y, 0.01, 1, 0.1}]
MILap[k_] := Sum[pJointLaplace[y, x, k, V] *
  Log[ $\frac{pJointLaplace[y, x, k, V]}{pOutputLap[y, k, V] pInput[x, 0.5, 0.1]}$ ] * 0.1 * 0.1,
  {y, 0.01, 1, 0.1}, {x, 0.01, 1, 0.1}]

```

```

MI[3]
MILap[3]

```

```
0.0977908
```

```
0.0981962
```

```

V = 20;
k = 1.5;
MILoop = 0;
For[y = 10-6, y < 1, y += 0.1,
  For[
    x = 10-6, x < 1, x += 0.1,
    (*Print[y, " - ", x];*)
    If[pJoint[y, x, k, V] > 0,
      MILoop += pJoint[y, x, k, V] * Log[ $\frac{pJoint[y, x, k, V]}{pOutput[y, k, V] pInput[x, 0.5, 0.1]}$ ] * 0.1 * 0.1;
      (*Print[MILoop];*)
    ]
  ]
]
MILoop

```

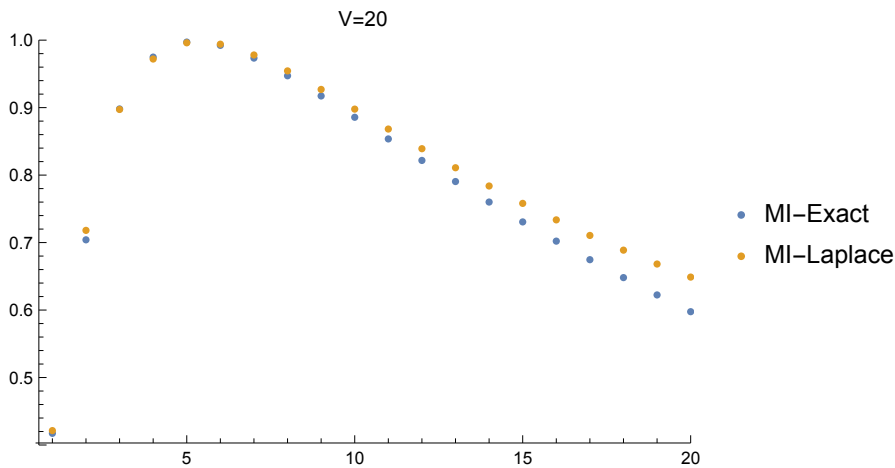
```
0.294445
```

```

MIList = Function[k, MI[k]]/@Range[0.1, 20, 0.5];
(*MILapList = Function[k, MILap[k]]/@Range[0.1, 50, 0.5];*)
MILapList = Function[k, MILap[k]]/@Range[0.1, 20, 0.5];
(* ListPlot[{MIList, MILapList}] *)
(*ListPlot[{MIList, MILapList}, PlotRange->All]*)

```

```
ListPlot[{MIList * 10, MILapList * 10}, PlotRange -> {{0.5, 20}, {0.4, 1}},
PlotLegends -> {"MI-Exact", "MI-Laplace"}, PlotLabel -> "V=20"]
```



## Under Development

(\* Compute MI

```
pOutputLap[y_,k_,V_] :=
NIntegrate[pLaplace[y,Inp,k,V]*pInput[Inp,0.5,0.1],{Inp,0,1}]
pJointApprox[y_,Inp_,k_,V_] := pLaplace[y,Inp,k,V]*pInput[Inp,0.5,0.1]

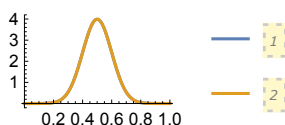
pOutput[y_,k_,V_] := NIntegrate[p[y,Inp,k,V]*pInput[Inp,0.5,0.1],{Inp,0,1}]
pJoint[y_,Inp_,k_,V_] := p[y,Inp,k,V]*pInput[Inp,0.5,0.1]

MI[k_] := Sum[Sum[pJoint[y,x,k]*Log[ $\frac{pJoint[y,x,k,V]}{pOutput[y,k,V]pInput[x,0.5,0.01]}$ ],
{x,0.01,1,0.05}],{y,0.01,1,0.05}]
MI[0.2]
$Aborted
```

```
MIList2 = Function[k,MI[k]]/@Range[0.5,5,0.5];
```

```
pIn = ProbabilityDistribution[pInput[x,0.5,0.1],{x,0,1},Method->"Normalize"]
```

```
Plot[{PDF[pIn,x],  $\frac{pInput[x,0.5,0.1]}{nIn}$ },{x,0,1},PlotRange->All,PlotLegends->Automatic]
ProbabilityDistribution[3.9894250911642475` e-49.99999999999999` (-0.5`+x)2,{x,0,1}]
```





```

nIn = NIntegrate[pInput[x,0.5,0.1],{x,0,1}]
0.25066268375731465`
Manipulate[
  nOutLap = Sum[pOutputLap[y,k,V]*0.005,{y,0.0,1,0.005}];
  nOut = Sum[pOutput[y,k,V]*0.005,{y,0,1,0.005}];
  Plot[{ $\frac{pOutput[y,k,V]}{nOut}$ ,  $\frac{pOutputLap[y,k,V]}{nOutLap}$ },{y,0.0,1},PlotRange→All],{V,1,1000}]

Sum[pOutputLap[y,k,V]* $\frac{0.01}{nOutLap}$ ,{y,0,1,0.01}]
1.00000000017787507`
V=5
5
  Normalize via lists
dx=dy=0.1;
pNormalizedList[y_,Inp_,k_,V_]:=
  p[y,Inp,k,V] (Total[Function[yp,p[yp,Inp,k,V]*dy]/@Range[0,1,dy]])-1
  (*Sum[pNormalizedList[y,Inp,k,V]*dy,{y,0,1,dy}]) (* Output → 1 *)
  pOutput[y_,k_,V_]:=
  Sum[pNormalizedList[y,Inp,k,V]*pInput[Inp,0.5,0.1]*dx,{Inp,10-8,1,dx}]
  (* Sum[pOutput[y,k,V]*dy,{y,0,1,dy}] Output → 0.9999991426275873 *)
  pJoint[y_,Inp_,k_,V_]:=pNormalizedList[y,Inp,k,V]*pInput[Inp,0.5,0.1]
(* Output probability as list *)
(*Column[Function[y,pOutput[y,0.2,5]]/@Range[0.005,0.995,0.1]]*)
(*Plot[pOutput[y,0.4,10],{y,0,1}])
*)

```