

NMST 434, Exercise session VI: M-estimators

April 1, 2019

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SetOptions[Plot, BaseStyle → FontSize → 16];
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Example: S_n as an M-estimator

```
Clear[M];
$Assumptions = τ > 0 && σ2 > 0;
ψ[μ_, σ2_, τ_] = {x - μ, (x - μ)^2 - σ2, √σ2 - τ};
D[ψ[μ, σ2, τ], {{μ, σ2, τ}}] /. x → μ
Ai = Inverse[%] /. σ → √σ2;
B =
  ((Outer[Times, ψ[μ, σ2, τ], ψ[μ, σ2, τ]] // Expand) /. Table[x^4 - j → M[4 - j], {j, 0, 4}])
Ai.B.Transpose[Ai] /. {M[1] → μ, M[2] → σ2 + μ^2, σ2 → τ^2} // Simplify // MatrixForm
```

$$\left\{ \{-1, 0, 0\}, \{0, -1, 0\}, \left\{0, \frac{1}{2\sqrt{\sigma^2}}, -1\right\} \right\}$$

$$\left\{ \left\{ \mu^2 - 2\mu M[1] + M[2], -\mu^3 + \mu\sigma^2 + 3\mu^2 M[1] - \sigma^2 M[1] - 3\mu M[2] + M[3], \right. \right.$$

$$\left. -\mu\sqrt{\sigma^2} + \mu\tau + \sqrt{\sigma^2} M[1] - \tau M[1] \right\}, \left\{ -\mu^3 + \mu\sigma^2 + 3\mu^2 M[1] - \sigma^2 M[1] - 3\mu M[2] + M[3], \right.$$

$$\left. \mu^4 - 2\mu^2\sigma^2 + \sigma^2^2 - 4\mu^3 M[1] + 4\mu\sigma^2 M[1] + 6\mu^2 M[2] - 2\sigma^2 M[2] - 4\mu M[3] + M[4], \right.$$

$$\left. \mu^2\sqrt{\sigma^2} - \sigma^2^{3/2} - \mu^2\tau + \sigma^2\tau - 2\mu\sqrt{\sigma^2} M[1] + 2\mu\tau M[1] + \sqrt{\sigma^2} M[2] - \tau M[2] \right\},$$

$$\left\{ -\mu\sqrt{\sigma^2} + \mu\tau + \sqrt{\sigma^2} M[1] - \tau M[1], \mu^2\sqrt{\sigma^2} - \sigma^2^{3/2} - \mu^2\tau + \sigma^2\tau - \right.$$

$$\left. 2\mu\sqrt{\sigma^2} M[1] + 2\mu\tau M[1] + \sqrt{\sigma^2} M[2] - \tau M[2], \sigma^2 - 2\sqrt{\sigma^2}\tau + \tau^2 \right\}$$

$$\begin{pmatrix} \sigma^2 & -\mu^3 - 3\mu\sigma^2 + M[3] & -\frac{\mu^3 + 3\mu\sigma^2 - M[3]}{2\tau} \\ -\mu^3 - 3\mu\sigma^2 + M[3] & 3\mu^4 + 6\mu^2\sigma^2 - 2\sigma^2\tau^2 + \tau^4 - 4\mu M[3] + M[4] & \frac{3\mu^4 + 6\mu^2\sigma^2 - 2\sigma^2\tau^2 + \tau^4 - 4\mu M[3] + M[4]}{2\tau} \\ -\frac{\mu^3 + 3\mu\sigma^2 - M[3]}{2\tau} & \frac{3\mu^4 + 6\mu^2\sigma^2 - 2\sigma^2\tau^2 + \tau^4 - 4\mu M[3] + M[4]}{2\tau} & \frac{3\mu^4 + 6\mu^2\sigma^2 - 2\sigma^2\tau^2 + \tau^4 - 4\mu M[3] + M[4]}{4\tau^2} \end{pmatrix}$$

Direct use of a Δ -theorem

```

Clear[M]
g[a_, b_] = {a, b - a^2, Sqrt[b - a^2]};
(J = (D[g[a, b], {{a, b}}] /. {a -> M[1], b -> M[2]})) // MatrixForm
Sigma = Table[M[i + j] - M[i] M[j], {i, 1, 2}, {j, 1, 2}];
J.S.Transpose[J] // Simplify // MatrixForm
Ai.B.Transpose[Ai] /. {mu -> M[1], sigma2 -> M[2] - M[1]^2, tau -> Sqrt[M[2] - M[1]^2]} // Simplify //
MatrixForm

```

$$\begin{pmatrix} 1 & 0 \\ -2M[1] & 1 \\ -\frac{M[1]}{\sqrt{-M[1]^2+M[2]}} & \frac{1}{2\sqrt{-M[1]^2+M[2]}} \end{pmatrix}$$

$$\begin{pmatrix} -M[1]^2 + M[2] & 2M[1]^3 - 3M[1]M[2] + M[3] & \frac{2M[1]^3-3}{2\sqrt{-I}} \\ 2M[1]^3 - 3M[1]M[2] + M[3] & -4M[1]^4 + 8M[1]^2M[2] - M[2]^2 - 4M[1]M[3] + M[4] & \frac{-4M[1]^4+8M[1]^2M[2]}{2\sqrt{-I}} \\ \frac{2M[1]^3-3M[1]M[2]+M[3]}{2\sqrt{-M[1]^2+M[2]}} & \frac{-4M[1]^4+8M[1]^2M[2]-M[2]^2-4M[1]M[3]+M[4]}{2\sqrt{-M[1]^2+M[2]}} & \frac{4M[1]^4-8M[1]^2M[2]}{4M[1]} \end{pmatrix}$$

$$\begin{pmatrix} -M[1]^2 + M[2] & 2M[1]^3 - 3M[1]M[2] + M[3] & \frac{2M[1]^3-3}{2\sqrt{-I}} \\ 2M[1]^3 - 3M[1]M[2] + M[3] & -4M[1]^4 + 8M[1]^2M[2] - M[2]^2 - 4M[1]M[3] + M[4] & \frac{-4M[1]^4+8M[1]^2M[2]}{2\sqrt{-I}} \\ \frac{2M[1]^3-3M[1]M[2]+M[3]}{2\sqrt{-M[1]^2+M[2]}} & \frac{-4M[1]^4+8M[1]^2M[2]-M[2]^2-4M[1]M[3]+M[4]}{2\sqrt{-M[1]^2+M[2]}} & \frac{4M[1]^4-8M[1]^2M[2]}{4M[1]} \end{pmatrix}$$

Special case of normal distribution

```

M[k_] = Moment[NormalDistribution[mu, sigma], k];
J.S.Transpose[J] // Simplify // MatrixForm
Ai.B.Transpose[Ai] /. {mu -> M[1], sigma2 -> M[2] - M[1]^2, tau -> Sqrt[M[2] - M[1]^2]} // Simplify //
MatrixForm

```

$$\begin{pmatrix} \sigma^2 & 0 & 0 \\ 0 & 2\sigma^4 & (\sigma^2)^{3/2} \\ 0 & (\sigma^2)^{3/2} & \frac{\sigma^2}{2} \end{pmatrix}$$

$$\begin{pmatrix} \sigma^2 & 0 & 0 \\ 0 & 2\sigma^4 & (\sigma^2)^{3/2} \\ 0 & (\sigma^2)^{3/2} & \frac{\sigma^2}{2} \end{pmatrix}$$

Example: Sample correlation coefficient as an M-estimator

For a general, properly integrable distribution. We find the **joint distribution** of the sample means of X and Y, sample variances of X and Y, sample average of XY, and the sample correlation coefficient.

For the joint distribution, WLOG centered distributions with unit variances does not apply (compare with the example Session1.nb). We must compute in full generality.

```

Clear[M];
$Assumptions = Element[μx, Reals] && Element[μy, Reals] &&
  Element[μxy, Reals] && σx2 > 0 && σy2 > 0 && -1 < ρ < 1;
ψ[μx_, μy_, μxy_, σx2_, σy2_, ρ_] = {x - μx, y - μy, xy - μxy,
  (x - μx)² - σx2, (y - μy)² - σy2,  $\frac{\mu xy - \mu x \mu y}{\sqrt{(\sigma x2 \sigma y2)}} - \rho$ };
(A0 = D[ψ[μx, μy, μxy, σx2, σy2, ρ], {{μx, μy, μxy, σx2, σy2, ρ}}]);
(A = A0 /. {x → M[1, 0], y → M[0, 1]});
(A = A /. μxy → M[1, 1]) // MatrixForm
(B0 = Outer[Times, ψ[μx, μy, μxy, σx2, σy2, ρ], ψ[μx, μy, μxy, σx2, σy2, ρ]]);
(B = (B0 // Expand) /. Flatten[Table[x4-j y4-k → M[4-j, 4-k], {j, 0, 4}, {k, 0, 4}]]);
(B = B /. μxy → M[1, 1] // Simplify) // MatrixForm

```

$$\begin{pmatrix} -1 & 0 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 & 0 \\ -2(-\mu x + M[1, 0]) & 0 & 0 & -1 & 0 & 0 \\ 0 & -2(-\mu y + M[0, 1]) & 0 & 0 & -1 & 0 \\ -\frac{\mu y}{\sqrt{\sigma x2 \sigma y2}} & -\frac{\mu x}{\sqrt{\sigma x2 \sigma y2}} & \frac{1}{\sqrt{\sigma x2 \sigma y2}} & -\frac{\sigma y2(-\mu x \mu y + M[1, 1])}{2(\sigma x2 \sigma y2)^{3/2}} & -\frac{\sigma x2(-\mu x \mu y + M[1, 1])}{2(\sigma x2 \sigma y2)^{3/2}} & -1 \end{pmatrix}$$

$$\begin{pmatrix} \mu x^2 - 2 \mu x M[1, 0] + M[2, 0] \\ \mu x (\mu y - M[0, 1]) - \mu y M[1, 0] + M[1, 1] \\ -M[1, 0] M[1, 1] + M[2, 1] \\ -\mu x^3 + 3 \mu x^2 M[1, 0] - \sigma x2 M[1, 0] + \mu x (\sigma x2 - 3 M[2, 0]) + M[3, 0] & -\sigma x2 \\ -\mu x (\mu y^2 - \sigma y2 - 2 \mu y M[0, 1] + M[0, 2]) + \mu y^2 M[1, 0] - \sigma y2 M[1, 0] - 2 \mu y M[1, 1] + M[1, 2] \\ \frac{(\mu x - M[1, 0]) (\rho \sigma x2 \sigma y2 + \sqrt{\sigma x2 \sigma y2} (\mu x \mu y - M[1, 1]))}{\sigma x2 \sigma y2} \end{pmatrix}$$

Asymptotic variance matrix for general distributions.

```
(Var = Inverse[A].B.Transpose[Inverse[A]] // Simplify) // MatrixForm
Var /. {μx → M[1, 0], μy → M[0, 1], σx2 → M[2, 0] - M[1, 0]^2, σy2 → M[0, 2] - M[0, 1]^2} //
Simplify // MatrixForm
```

$$\frac{2 \sqrt{\sigma_x^2 \sigma_y^2} (\mu_x - M[1, 0]) \left(\rho \sigma_x^2 \sigma_y^2 + \sqrt{\sigma_x^2 \sigma_y^2} (\mu_x \mu_y - M[1, 1]) \right) - 2 \sigma_x^2 (\mu_x \sigma_y^2 - (\mu_y - M[0, 1]) (\mu_x \mu_y - M[1, 1])) (\mu_x (\mu_y - M[0, 1]) - \mu_y M[1, 0])}{M[1, 1] M[1, 2] (M[1, 0]^2 - M[2, 0]) - M[0, 1]^3 (M[1, 0]^2 M[2, 0] - 2 M[2, 0]^2 + M[1, 0] M[3, 0]) + M[0, 2] (M[1, 0]^3 M[1, 1] - 2 M[1, 0]^2 M[2, 1] + 2 M[1, 0] M[3, 1]) - M[0, 1] M[2, 1] (M[1, 0]^2 - M[2, 0]) - M[0, 1] M[3, 1] (M[1, 0] - M[2, 0]) + M[0, 2] (M[1, 0] M[3, 0] - M[2, 1] (M[1, 0] - M[2, 0]))}$$

Asymptotic variance of the sample correlation coefficient.

```
Clear[M]
Short[(Var[[6, 6]] /. {μx → M[1, 0], μy → M[0, 1],
σx2 → M[2, 0] - M[1, 0]^2, σy2 → M[0, 2] - M[0, 1]^2}) // Simplify]
Var[[6, 6]] /. {μx → M[1, 0], μy → M[0, 1], σx2 → M[2, 0] - M[1, 0]^2,
σy2 → M[0, 2] - M[0, 1]^2, ρ → 0} // Simplify
```

$$\frac{(M[0, 4] M[1, 1]^2 (M[1, 0]^2 - M[2, 0])^2 - 4 \ll 1 \gg^3 \ll 1 \gg^2 ((-1 + \rho^2) M[1, 0]^2 - \rho^2 M[2, 0]) + \ll 7 \gg + \ll 1 \gg - 2 M[0, 1] (\ll 1 \gg))}{(4 (M[0, 1]^2 - M[0, 2])^3 (M[1, 0]^2 - M[2, 0])^3)}$$

1

$$\begin{aligned}
& 4 (M[0, 1]^2 - M[0, 2])^3 (M[1, 0]^2 - M[2, 0])^3 \\
& (M[0, 4] M[1, 1]^2 (M[1, 0]^2 - M[2, 0])^2 + 4M[0, 2]^3 (M[1, 0]^3 - M[1, 0] M[2, 0])^2 + \\
& 2M[0, 2] M[1, 1] (M[1, 0]^2 - M[2, 0]) (2M[1, 0] M[1, 1] M[1, 2] - 2M[1, 0]^2 M[1, 3] + \\
& 2M[1, 3] M[2, 0] + 2M[0, 3] (M[1, 0]^3 - M[1, 0] M[2, 0]) - M[1, 1] M[2, 2]) + \\
& M[0, 1]^6 (4M[1, 0]^6 - 8M[1, 0]^4 M[2, 0] + 4M[2, 0]^3 - 4M[1, 0] M[2, 0] M[3, 0] + \\
& M[1, 0]^2 (3M[2, 0]^2 + M[4, 0])) + \\
& M[0, 2]^2 (-8M[1, 0]^5 M[1, 2] + 4M[1, 0]^3 (4M[1, 2] M[2, 0] - 3M[1, 1] M[2, 1]) + \\
& 4M[2, 0]^2 M[2, 2] + M[1, 0]^4 (11M[1, 1]^2 + 4M[2, 2]) - 4M[1, 0] \\
& (2M[1, 2] M[2, 0]^2 + M[1, 1] (-3M[2, 0] M[2, 1] + M[1, 1] M[3, 0]))) - 4M[1, 1] \\
& M[2, 0] M[3, 1] - 4M[1, 0]^2 (3M[1, 1]^2 M[2, 0] + 2M[2, 0] M[2, 2] - M[1, 1] M[3, 1]) + \\
& M[1, 1]^2 (4M[2, 0]^2 + M[4, 0])) + M[0, 1]^4 (11M[1, 1]^2 M[2, 0]^2 - \\
& 4M[1, 0]^3 (M[1, 2] M[2, 0] + M[1, 1] M[2, 1]) + 6M[1, 0]^4 M[2, 2] + 4M[2, 0]^2 M[2, 2] + \\
& 4M[1, 0] (M[1, 2] M[2, 0]^2 + M[1, 1] M[2, 0] M[2, 1] - M[1, 1]^2 M[3, 0]) - 4M[1, 1] \\
& M[2, 0] M[3, 1] - 2M[1, 0]^2 (4M[1, 1]^2 M[2, 0] + 5M[2, 0] M[2, 2] - 2M[1, 1] M[3, 1])) + \\
& M[1, 1]^2 M[4, 0] - 2M[0, 2] (4M[1, 0]^6 - 7M[1, 0]^4 M[2, 0] + 4M[2, 0]^3 - \\
& 4M[1, 0] M[2, 0] M[3, 0] + M[1, 0]^2 (2M[2, 0]^2 + M[4, 0]))) + \\
& M[0, 1]^2 ((M[1, 0]^2 - M[2, 0]) (M[0, 4] (M[1, 0]^4 - M[1, 0]^2 M[2, 0]) + \\
& 2M[1, 1] (-2M[1, 0] M[1, 1] M[1, 2] + 2M[1, 0]^2 M[1, 3] - 2M[1, 3] M[2, 0] + \\
& 2M[0, 3] (M[1, 0]^3 - M[1, 0] M[2, 0]) + M[1, 1] M[2, 2])) + M[0, 2]^2 (3M[1, 0]^6 - \\
& 4M[1, 0]^4 M[2, 0] + 4M[2, 0]^3 - 4M[1, 0] M[2, 0] M[3, 0] + M[1, 0]^2 M[4, 0]) + \\
& 2M[0, 2] (4M[1, 0]^5 M[1, 2] + M[1, 0]^3 (-6M[1, 2] M[2, 0] + 8M[1, 1] M[2, 1]) - \\
& 4M[2, 0]^2 M[2, 2] - M[1, 0]^4 (4M[1, 1]^2 + 5M[2, 2]) + \\
& 2M[1, 0] (M[1, 2] M[2, 0]^2 + 2M[1, 1] (-2M[2, 0] M[2, 1] + M[1, 1] M[3, 0]))) + \\
& 4M[1, 1] M[2, 0] M[3, 1] + M[1, 0]^2 (7M[1, 1]^2 M[2, 0] + 9M[2, 0] M[2, 2] - \\
& 4M[1, 1] M[3, 1]) - M[1, 1]^2 (6M[2, 0]^2 + M[4, 0])) - \\
& 2M[0, 1]^5 (4M[1, 0]^5 M[1, 1] + 2M[2, 0] (2M[2, 0] M[2, 1] - M[1, 1] M[3, 0]) - \\
& 2M[1, 0]^2 (2M[2, 0] M[2, 1] + M[1, 1] M[3, 0]) + \\
& 2M[1, 0]^3 (-5M[1, 1] M[2, 0] + M[3, 1]) + \\
& M[1, 0] (-2M[2, 0] M[3, 1] + M[1, 1] (9M[2, 0]^2 + M[4, 0]))) + \\
& 4M[0, 1]^3 (- (M[1, 0]^2 - M[2, 0]) (M[1, 0]^2 M[1, 1] M[1, 2] + M[1, 0]^3 M[1, 3] + \\
& M[1, 1] (-3M[1, 2] M[2, 0] + M[1, 1] M[2, 1]) + \\
& M[1, 0] (-2M[1, 1]^3 - M[1, 3] M[2, 0] + M[1, 1] M[2, 2])) + M[0, 2] \\
& (5M[1, 0]^5 M[1, 1] - M[1, 0]^4 M[2, 1] + 2M[2, 0] (2M[2, 0] M[2, 1] - M[1, 1] M[3, 0]) - \\
& M[1, 0]^2 (3M[2, 0] M[2, 1] + 2M[1, 1] M[3, 0]) + M[1, 0]^3 (-9M[1, 1] M[2, 0] + \\
& 2M[3, 1]) + M[1, 0] (-2M[2, 0] M[3, 1] + M[1, 1] (7M[2, 0]^2 + M[4, 0]))) - \\
& 2M[0, 1] (M[1, 1] (M[0, 4] M[1, 0] + 2M[0, 3] M[1, 1]) (M[1, 0]^2 - M[2, 0])^2 + \\
& 2M[0, 2] (M[1, 0]^2 - M[2, 0]) (-M[1, 0]^2 M[1, 1] M[1, 2] - M[1, 0]^3 M[1, 3] + \\
& M[0, 3] (M[1, 0]^4 - M[1, 0]^2 M[2, 0]) + M[1, 1] (3M[1, 2] M[2, 0] - M[1, 1] M[2, 1]) + \\
& M[1, 0] (2M[1, 1]^3 + M[1, 3] M[2, 0] - M[1, 1] M[2, 2])) + \\
& M[0, 2]^2 (9M[1, 0]^5 M[1, 1] - 2M[1, 0]^4 M[2, 1] + \\
& 2M[2, 0] (2M[2, 0] M[2, 1] - M[1, 1] M[3, 0]) - 2M[1, 0]^2 \\
& (M[2, 0] M[2, 1] + M[1, 1] M[3, 0]) + 2M[1, 0]^3 (-7M[1, 1] M[2, 0] + M[3, 1]) + \\
& M[1, 0] (-2M[2, 0] M[3, 1] + M[1, 1] (8M[2, 0]^2 + M[4, 0])))
\end{aligned}$$

Special case of the bivariate normal distribution.

```

dist = MultinormalDistribution[{μx, μy}, {{σx2, ρ √(σx2 σy2)}, {ρ √(σx2 σy2), σy2}}];
M[i_, j_] = Moment[dist, {i, j}];
Var // Simplify // MatrixForm
Var /. {μx → 0, μy → 0, σx2 → 1, σy2 → 1} // Simplify // MatrixForm

```

$$\begin{pmatrix} \sigma x^2 & \rho \sqrt{\sigma x^2 \sigma y^2} & \mu y \sigma x^2 + \mu x \rho \sqrt{\sigma x^2 \sigma y^2} \\ \rho \sqrt{\sigma x^2 \sigma y^2} & \sigma y^2 & \mu x \sigma y^2 + \mu y \rho \sqrt{\sigma x^2 \sigma y^2} \\ \mu y \sigma x^2 + \mu x \rho \sqrt{\sigma x^2 \sigma y^2} & \mu x \sigma y^2 + \mu y \rho \sqrt{\sigma x^2 \sigma y^2} & \mu y^2 \sigma x^2 + (\mu x^2 + \sigma x^2 + \rho^2 \sigma x^2) \sigma y^2 + 2 \mu x \mu y \rho \sqrt{\sigma x^2 \sigma y^2} \\ 0 & 0 & 2 \rho \sigma x^2 \sqrt{\sigma x^2 \sigma y^2} \\ 0 & 0 & 2 \rho \sigma y^2 \sqrt{\sigma x^2 \sigma y^2} \\ 0 & 0 & -(-1 + \rho^2) \sqrt{\sigma x^2 \sigma y^2} \end{pmatrix}$$

$$\begin{pmatrix} 1 & \rho & 0 & 0 & 0 & 0 \\ \rho & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 + \rho^2 & 2\rho & 2\rho & 1 - \rho^2 \\ 0 & 0 & 2\rho & 2 & 2\rho^2 & \rho - \rho^3 \\ 0 & 0 & 2\rho & 2\rho^2 & 2 & \rho - \rho^3 \\ 0 & 0 & 1 - \rho^2 & \rho - \rho^3 & \rho - \rho^3 & (-1 + \rho^2)^2 \end{pmatrix}$$

Direct use of the Δ -theorem (only for the asymptotic distribution of $\hat{\rho}$, see Session1.nb)

```
Clear[μ];
```

$$g[a_, b_, c_, d_, e_] = \frac{(e - a b)}{((c - a^2)(d - b^2))^{1/2}};$$

```
Dg =
```

```
D[g[a, b, c, d, e], {{a, b, c, d, e}}] /. {a → 0, b → 0, c → 1, d → 1, e → ρ} // Simplify;
```

```
MatrixForm[Dg] (* gradient of g at E(a,b,c,d,e) *)
```

```
Σ = {{1, ρ, μ[3, 0], μ[1, 2], μ[2, 1]}, {ρ, 1, μ[2, 1], μ[0, 3], μ[1, 2]},
      {μ[3, 0], μ[2, 1], μ[4, 0] - 1, μ[2, 2] - 1, μ[3, 1] - μ[1, 1]},
      {μ[1, 2], μ[0, 3], μ[2, 2] - 1, μ[0, 4] - 1, μ[1, 3] - μ[1, 1]},
      {μ[2, 1], μ[1, 2], μ[3, 1] - μ[1, 1], μ[1, 3] - μ[1, 1], μ[2, 2] - μ[1, 1]^2}};
```

```
(* original variance matrix *)
```

```
MatrixForm[Σ]
```

```
Dg.Σ.Dg // Simplify (* asymptotic variance of  $\hat{\rho}$  *)
```

```
Dg.Σ.Dg /. ρ → 0 (* asymptotic variance under independence *)
```

$$\begin{pmatrix} 0 \\ 0 \\ -\frac{\rho}{2} \\ -\frac{\rho}{2} \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & \rho & \mu[3, 0] & \mu[1, 2] & \mu[2, 1] \\ \rho & 1 & \mu[2, 1] & \mu[0, 3] & \mu[1, 2] \\ \mu[3, 0] & \mu[2, 1] & -1 + \mu[4, 0] & -1 + \mu[2, 2] & -\mu[1, 1] + \mu[3, 1] \\ \mu[1, 2] & \mu[0, 3] & -1 + \mu[2, 2] & -1 + \mu[0, 4] & -\mu[1, 1] + \mu[1, 3] \\ \mu[2, 1] & \mu[1, 2] & -\mu[1, 1] + \mu[3, 1] & -\mu[1, 1] + \mu[1, 3] & -\mu[1, 1]^2 + \mu[2, 2] \end{pmatrix}$$

$$2\rho\mu[1, 1] - \mu[1, 1]^2 + \mu[2, 2] -$$

$$\rho\left(\mu[1, 3] + \mu[3, 1]\right) + \frac{1}{4}\rho^2\left(-4 + \mu[0, 4] + 2\mu[2, 2] + \mu[4, 0]\right)$$

$$-\mu[1, 1]^2 + \mu[2, 2]$$

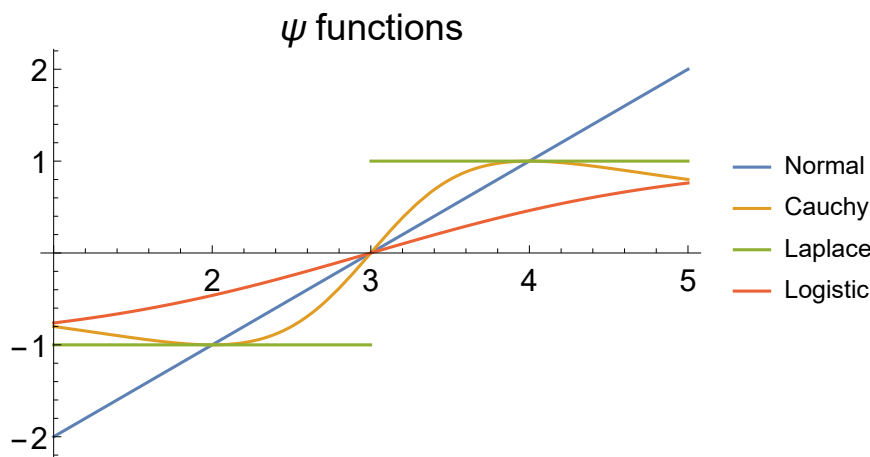
Example: Robust estimation of location

```

$Assumptions = Element[θ, Reals];
dist = {NormalDistribution[], StudentTDistribution[1],
        LaplaceDistribution[], LogisticDistribution[]};
ρ[x_, θ_] = Log[PDF[#, x - θ]] & /@ dist;
D[ρ[x, θ], θ]
ψ[x_, θ_] = %;
Plot[Evaluate[ψ[x, 3]], {x, 3 - 2, 3 + 2},
      PlotLegends -> {"Normal", "Cauchy", "Laplace", "Logistic"}, PlotLabel -> "ψ functions"]

```

$$\left\{ x - \theta, \frac{2(x - \theta)}{1 + (x - \theta)^2}, \begin{cases} \frac{e^{-x+\theta}}{2} & x - \theta \geq 0 \\ -\frac{e^{-x-\theta}}{2} & \text{True} \end{cases}, e^{x-\theta} (1 + e^{-x+\theta})^2 \left(-\frac{2e^{-2x+2\theta}}{(1 + e^{-x+\theta})^3} + \frac{e^{-x+\theta}}{(1 + e^{-x+\theta})^2} \right) \right\}$$



Inverse asymptotic variance (Fisher information)

```

Clear[x]
-D[ψ[x, θ], {θ}] // Simplify
Table[Integrate[%[[j]] PDF[dist[[j]], x - θ], {x, -∞, ∞}], {j, 1, Length[dist]}]

```

$$\left\{ 1, -\frac{2(-1 + x^2 - 2x\theta + \theta^2)}{(1 + x^2 - 2x\theta + \theta^2)^2}, \theta, \frac{2e^{x+\theta}}{(e^x + e^\theta)^2} \right\}$$

$$\left\{ 1, \frac{1}{2}, \theta, \frac{1}{3} \right\}$$