1 / 1 point
True
False
Correct! The noise must be from the Gaussian family.  2. Question 2 The product of several Gaussian PDFs with identical variances is also Gaussian 1/1 point
True
False
Correct! We used this fact to derive the connection between maximum likelihood and least squares.  3. Question 3 The least squares criterion is robust to outliers.  1/1 point
True
False
Correct! Least squares is particularly sensitive to outliers due to the use of squared errors!  4.  Question 4

For a scalar Gaussian random variable, what is the form of the full log likelihood

function?

The method of maximum likelihood gives the same parameter estimates as the

method of least squares for any measurement noise distribution.

<u> </u>			
2			
1			
log(2π)-			
log(σ	-		
)-	1		
2			
$(x-\mu)$			
		log \left( \sigma \right) -	\frac { 1 } { 2
		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ -		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ -		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+ log(σ)-		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+ log(σ)-		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+ log(σ)-		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma ^ { 2 } }\left( x - \ - log(2π)+ log(σ)- 2 (x-μ)		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma $^{2}$ \\left( x - \ - \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \		log \left( \sigma \right) -	\frac { 1 } { 2
- \frac $\{1\}\{2\} \setminus \{0\} (2 \neq i)$ \sigma $\{2\} \setminus \{1\} \in \{1\} (x - i)$ - $(2\pi) + (2\pi) = (2\pi) + (2\pi) = (2\pi)$		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma $^{2}$ \left( x - \ - \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \		log \left( \sigma \right) -	\frac { 1 } { 2
\sigma $^{2}$ \\left( x - \ - \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	mu \right) ^ { 2 }		\frac { 1 } { 2

argmin	
f(x)=argmax	
f(-x).	
1 / 1 point	
True	
False	
_	
Correct	
Correct!	aramavl ( v ) f(v)
\mathrm{argmin}_{~x~} f(x) = a argmin	arginax <sub>}_{~x~} = i(x)</sub>
31911111	
f(x)=argmax	
-f(x).	