

## Problem 6: Maximum Likelihood Estimation of Phase Noise

### Phase Noise Estimation under Gaussian Noise: Setup

This problem is motivated by estimation in communication systems (Wi-Fi, cellphones, etc). The solution obtained in this problem is implemented real-time in many communication systems. For example, your laptop Wi-Fi adapter, which is downloading and uploading all the content that you are consuming in this course, is performing this estimation (albeit in a more complicated statistical model) tens of hundreds of times every second.

Let

$$\mathbf{x} = \begin{bmatrix} \cos(\theta) \\ \sin(\theta) \end{bmatrix}$$

be a known vector, i.e. **we assume that we know  $\theta$** . Let  $\theta \in [0, \pi/2]$ .

Let  $\mathbf{Y}_1, \dots, \mathbf{Y}_n$  be defined as follows:

$$\mathbf{Y}_i = \begin{bmatrix} Y_i^{(1)} \\ Y_i^{(2)} \end{bmatrix} = \begin{bmatrix} \cos(\theta + \phi) \\ \sin(\theta + \phi) \end{bmatrix} + \mathbf{Z}_i, \quad i = 1, \dots, n,$$

where  $\mathbf{Z}_i \sim \mathcal{N}(0, \sigma^2 \mathbf{I}_2)$  for a known  $\sigma^2$  and  $\phi$  is an unknown constant. Assume that  $\mathbf{Z}_i, i = 1, \dots, n$  are independent.

**Objective:** Upon observing  $\mathbf{Y}_i, i = 1, \dots, n$  we wish to produce an estimate  $\hat{\phi}$  of  $\phi \in [-\pi, \pi]$ .

#### (a) True or False

1/1 point (graded)

Select whether the following statement is **true or false**: " $\mathbf{Y}_i$  are iid."

☒ True ✓

☐ False

#### Solution:

The statement is true. The multivariate Gaussian vectors  $\mathbf{Z}_i$  are iid. Therefore,  $\mathbf{Y}_i$ , which are deterministic functions of the  $\mathbf{Z}_i$ 's, respectively for  $i = 1, \dots, n$ , are iid.

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You have used 1 of 3 attempts

**i** Answers are displayed within the problem

#### (b) The Underlying Problem

1/1 point (graded)

Referring to the **objective** in the problem setup given above, select from the following the statements that are correct. (Choose all that apply.)

☐ We are trying to estimate the **magnitude** by which  $\mathbf{x}$  is scaled (in the presence of vector Gaussian noise).

☒ We are trying to estimate the **phase rotation** undergone by  $\mathbf{x}$  (in the presence of vector Gaussian noise).

☐ We are trying to estimate the **magnitude and phase changes** undergone by  $\mathbf{x}$  (in the presence of vector Gaussian noise).



**Grading Note:** Partial credit is given.

**Solution:**

The objective of the problem states that we wish to produce an estimate of  $\phi$ , which is the phase rotation undergone by  $\mathbf{x}$  under an additive Gaussian noise.

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You have used 1 of 3 attempts

Answers are displayed within the problem

(c) Observation Under Zero Noise

1.0/1 point (graded)

For a moment, assume that there is **no Gaussian noise in the problem**. That is, let  $\mathbf{Y}_i = [\cos(\theta + \phi) \quad \sin(\theta + \phi)]^T \triangleq \begin{pmatrix} Y^{(1)} \\ Y^{(2)} \end{pmatrix}$ , for all  $i$ , in this sub-problem.

For simplicity, assume that  $\theta \in [0, \pi/2]$ ,  $\theta + \phi \in [0, \pi/2]$ .

What is  $\phi$ ?

(Express your answer in terms of  $Y^{(1)}$ ,  $Y^{(2)}$ ,  $\theta$ , and the  $\arctan(x)$  function. Use **Y\_1** for  $Y^{(1)}$  and **Y\_2** for  $Y^{(2)}$ . Type **arctan(x)** for  $\arctan(x)$  (where  $x$  can be any expression). **Do not use** any trigonometric function other than  $\arctan$ .)

$\phi =$

arctan(Y\_2/Y\_1) - theta

**Answer:** -theta + arctan(Y\_2/Y\_1)

STANDARD NOTATION

**Solution:**

If there is no noise, the value of  $\phi$  is  $\arctan\left(\frac{Y^{(2)}}{Y^{(1)}}\right) - \theta$ , where we assume that  $\theta \in [0, \pi/2]$ ,  $\theta + \phi \in [0, \pi/2]$  for simplicity.

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You have used 1 of 3 attempts

Answers are displayed within the problem

(d) Maximum Likelihood Estimator of the Phase Noise - Log Likelihood

1.0/1 point (graded)

Now, let us return to the original setup. What is the log-likelihood  $\ell_n(\mathbf{Y}_1, \dots, \mathbf{Y}_n; \phi)$ ?

**For the answer box below, ignore the term  $\ln\left(\frac{1}{(\sqrt{2\pi\sigma^2})^{2n}}\right)$  in the log-likelihood and input the rest of the log-likelihood expression.**

(Use **Sigma\_i(X\_i)** for  $\sum_{i=1}^n (X_i)$  (where  $X_i$  can be any quantity in a series indexed by  $i$ ), **Y\_1** for  $Y_i^{(1)}$ , and **Y\_2** for  $Y_i^{(2)}$ . Enter **sin(x)** for  $\sin(x)$ , **cos(x)** for  $\cos(x)$ .)

- (Sigma\_i((Y\_1 - cos(theta+phi))^2 + (Y\_2 - sin(theta+phi))^2))/(2\*sigma^2)

**Answer:** -(1/(2\*sigma^2))\*Sigma\_i((Y\_1 - cos(theta + phi))^2 + (Y\_2 - sin(theta + phi))^2)

STANDARD NOTATION

Solution:

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You have used 1 of 3 attempts

$$\ell_n(\mathbf{Y}_1, \dots, \mathbf{Y}_n; \phi) = \log L_n(\mathbf{Y}_1, \dots, \mathbf{Y}_n; \phi) = \log \left( \prod_{i=1}^n \frac{\exp \left( -\frac{1}{2} \left[ \frac{(Y_i^{(1)} - \cos(\theta + \phi))^2}{\sigma^2} + \frac{(Y_i^{(2)} - \sin(\theta + \phi))^2}{\sigma^2} \right] \right)}{2\pi\sigma^2} \right)$$
$$= \log \left( \frac{1}{(2\pi\sigma^2)^n} \right) - \frac{1}{2\sigma^2} \sum_{i=1}^n \left[ (Y_i^{(1)} - \cos(\theta + \phi))^2 + (Y_i^{(2)} - \sin(\theta + \phi))^2 \right]$$

Note: My notes showed I also calculated the expression passed to the  $\exp()$  function in detail, but I am too tired to type that up in LaTeX now.

Answers are displayed within the problem

(e) Maximum Likelihood Estimator of the Phase Noise

1.0/1 point (graded)

Let  $\hat{\mu}_1 = \sum_{i=1}^n \frac{Y_i^{(1)}}{n}$  and  $\hat{\mu}_2 = \sum_{i=1}^n \frac{Y_i^{(2)}}{n}$ .

Compute the maximum likelihood estimator  $\hat{\phi}_{n,\text{MLE}}$  of  $\phi$  upon observing  $\mathbf{Y}_i, i = 1, \dots, n$

**Note:** Again for simplicity, assume while entering the expression in the following box that  $\theta \in [0, \pi/2]$ ,  $\hat{\mu}_1 > 0$ , and  $\hat{\mu}_2 > 0$ .

(Use **hatmu\_1** for  $\hat{\mu}_1$  and **hatmu\_2** for  $\hat{\mu}_2$ . Type **arctan(x)** for  $\arctan(x)$  (where  $x$  can be any expression). **Do not use** any trigonometric function other than **arctan**.)

$\hat{\phi}_{n,\text{MLE}} =$  arctan(hatmu\_2/hatmu\_1) - theta ✓ Answer: -theta + arctan(hatmu\_2/hatmu\_1)

STANDARD NOTATION

Solution:

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You have used 1 of 3 attempts

$$\frac{\partial \ell_n}{\partial \phi} = \frac{\partial}{\partial \phi} \left[ -\frac{1}{2\sigma^2} \sum_{i=1}^n \left( (Y_i^{(1)} - \cos(\theta + \phi))^2 + (Y_i^{(2)} - \sin(\theta + \phi))^2 \right) \right]$$
$$= -\frac{1}{2\sigma^2} \sum_{i=1}^n \left( 2(Y_i^{(1)} - \cos(\theta + \phi)) \sin(\theta + \phi) - 2(Y_i^{(2)} - \sin(\theta + \phi)) \cos(\theta + \phi) \right) = -\frac{1}{\sigma^2} \sum_{i=1}^n (Y_i^{(1)} \sin(\theta + \phi) - Y_i^{(2)} \cos(\theta + \phi))$$

Setting  $\frac{\partial \ell_n}{\partial \phi} = 0$  and solving for  $\phi$ , we have

$$\sin(\theta + \phi) \sum_{i=1}^n Y_i^{(1)} = \cos(\theta + \phi) \sum_{i=1}^n Y_i^{(2)}$$
$$\implies \frac{\sin(\theta + \phi)}{\cos(\theta + \phi)} = \tan(\theta + \phi) = \frac{\sum_{i=1}^n Y_i^{(2)}}{\sum_{i=1}^n Y_i^{(1)}} = \frac{\hat{\mu}_2}{\hat{\mu}_1}$$
$$\implies \theta + \phi = \arctan\left(\frac{\hat{\mu}_2}{\hat{\mu}_1}\right) \text{ Note: This step assumes } \theta + \phi \in [0, \pi/2].$$
$$\implies \phi = \arctan\left(\frac{\hat{\mu}_2}{\hat{\mu}_1}\right) - \theta$$

So  $\hat{\phi}_{n,\text{MLE}} = \arctan\left(\frac{\hat{\mu}_2}{\hat{\mu}_1}\right) - \theta$ .

(f) Geometry of the MLE of Phase Noise

0.5/1 point (graded)

Select from the following all statements that are true. (Choose all that apply.)

- ☒ The MLE of  $\phi$  does not change if we scaled  $\mathbf{x}$  by  $r > 0$ . ✓
- ☐ The MLE of  $\phi$  does not change if the covariance matrix of the multivariate Gaussian is scaled by  $s > 0$ . ✓

\* [ptressel](#)  
11 days ago - endorsed 10 days ago by [sudarsanvsr\\_mit](#) (Staff)

**Grading Note:** Partial credit is given.

Solution:

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You have used 2 of 3 attempts

Here's a solution for part f (perhaps too wordy...):

"The MLE of  $\phi$  does not change if we scale  $x$  by  $r > 0$ ."

This is true: The factor of  $r$  will appear anywhere  $x$  does, and will not affect the location of the maximum of the log likelihood over  $\phi$  because it will drop out when we set the derivative to zero. The derivative of the log likelihood becomes:

$$\ell'(Y_1, \dots, Y_n; \phi) = -\frac{n}{\sigma^2} \hat{\mu}_1 r \sin(\theta + \phi) + \frac{n}{\sigma^2} \hat{\mu}_2 r \cos(\theta + \phi)$$

The constant factor  $\frac{nr}{\sigma^2}$  will cancel when we equate that to zero.

"The MLE of  $\phi$  does not change if the covariance matrix of the multivariate Gaussian is scaled by  $s > 0$ ."

This is true:  $\sigma^2$  does not appear in the MLE. Multiplying  $\sigma^2$  by  $s > 0$  is the same as changing  $\sigma^2$ . Since  $\sigma^2$  was arbitrary to begin with, we can just absorb the  $s$  into  $\sigma^2$ , and the derivation of the MLE remains unchanged.

Answers are displayed within the problem

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Just to add: The whole point of the problem is that it is **intuitive**. The case when we have no Gaussian noise says we just take the arctan of whatever point we see and we subtract off  $\theta$ . The same principle carries over to the case when we have the Gaussian noise. We estimate the mean of the two Gaussian-added- $\mathbf{x}$  coordinates. With enough samples, this is a consistent estimator. We then take the arctan of the estimated point on the 2-D space and once again subtract off  $\theta$ . All Rights Reserved