# EE 508 Lecture 20

### **Sensitivity Functions**

- Comparison of Filter Structures
- Performance Prediction

### **GB effects in KRC and -KRC Lowpass Filter**

$$T(s) = \frac{K_{0}\omega_{0}^{2}}{s^{2} + s\left[\frac{\omega_{0}}{Q}\right] + \omega_{0}^{2} + K_{0}\tau s\left(s^{2} + s\left[\frac{\omega_{0}}{Q}\left(1 + K_{0}Q\sqrt{\frac{R_{1}C_{1}}{R_{2}C_{2}}}\right)\right] + \omega_{0}^{2}\right)}$$

$$T(s) = -K_{0}\frac{\frac{1}{R_{1}R_{2}C_{1}C_{2}}}{\left(s^{2} + s\left[\frac{1}{R_{1}C_{1}}\left(1 + \frac{R_{1}}{R_{3}}\right) + \frac{1}{R_{4}C_{2}} + \frac{1}{R_{2}C_{2}}\left(1 + \frac{C_{2}}{C_{1}}\right)\right] + \left[\frac{1 + (R_{1}/R_{3})(1 + K_{0}) + (R_{1}/R_{4})(1 + (R_{2}/R_{3}) + (R_{2}/R_{1}))}{R_{1}R_{2}C_{1}C_{2}}\right]\right)}$$

$$+ \tau s(1 + K_{0})\left(s^{2} + s\left[\frac{1}{R_{1}C_{1}}\left(1 + \frac{R_{1}}{R_{3}}\right) + \frac{1}{R_{4}C_{2}} + \frac{1}{R_{2}C_{2}}\left(1 + \frac{C_{2}}{C_{1}}\right)\right] + \left[\frac{1 + (R_{1}/R_{3}) + (R_{1}/R_{4})(1 + (R_{2}/R_{3}) + (R_{2}/R_{1}))}{R_{1}R_{2}C_{1}C_{2}}\right]\right)\right]$$

- Analytical expressions for  $\omega_{0,}$  Q, poles, zeros, and other key parameters are unwieldly in these circuits and as bad or worse in many other circuits (require solution of 3<sup>rd</sup> order polynomial!!)
- Sensitivity metrics give considerable insight into how filters perform and are widely used to assess relative performance
- Need sensitivity characterization of real numbers as well as complex quantities such as poles and zeros
- If sensitivity expressions are obtained for a given structure, it can be catalogued rather than recalculated
- Since analytical expressions for key parameters are unwieldly in even simple circuits, obtaining expressions for the purpose of calculating sensitivity appears to be a formidable task !

Define the standard sensitivity function as

$$\mathbf{S}_x^f = \frac{\partial f}{\partial x} \bullet \frac{x}{f}$$



### Define the derivative sensitivity function as

$$\boldsymbol{s}_{x}^{f} = \frac{\partial f}{\partial x}$$



Is more useful when x or f ideally assume extreme values of 0 or  $\infty$ 



**Dependent on circuit structure** (for some circuits, also not dependent on components)

The sensitivity functions are thus useful for comparing different circuit structures

The variability which is the product of the sensitivity function and the normalized component differential is more important for predicting circuit performance

### **Variability Formulation**

$$V_{\mathbf{x}_{i}}^{f} = \mathbf{S}_{\mathbf{x}_{i}}^{f} \Big|_{\bar{X}_{N}} \cdot \frac{\mathbf{d}\mathbf{x}_{i}}{\mathbf{x}_{iN}}$$
$$\frac{\mathbf{d}\mathbf{F}}{\mathbf{F}} = \sum_{i=1}^{k} V_{\mathbf{x}_{i}}^{f} \Big|_{\bar{X}_{N}}$$

Variability includes effects of both circuit structure and components on performance

If component variations are small, high sensitivities are acceptable

If component variations are large, low sensitivities are usually critical



At this stage, this is just an observation about summed sensitivities but later will establish some fundamental properties of summed sensitivities

$$\frac{\mathsf{dF}}{\mathsf{F}} = \sum_{i=1}^{k} \left( \mathsf{S}_{\mathsf{X}_{i}}^{\mathsf{f}} \Big|_{\bar{X}_{N}} \bullet \frac{\mathsf{d}\mathsf{X}_{i}}{\mathsf{X}_{i}} \right) + \frac{1}{\mathsf{F}_{\mathsf{N}}} \sum_{i=1}^{ky} \left( \mathsf{s}_{\mathsf{Y}_{i}}^{\mathsf{f}} \Big|_{\bar{\mathsf{Y}}_{\mathsf{N}}=0} \bullet \mathsf{Y}_{i} \right)$$

Low sensitivities in a circuit are often preferred but in some applications, low sensitivities would be totally unacceptable

Examples where low sensitivities are unacceptable are circuits where a charactristics F must be tunable or adjustable!

### Some useful sensitivity theorems

 $S_x^{kf} = S_x^{f}$ where k is a constant  $S_x^{f^n} = n \bullet S_x^{f}$  $S_x^{1/f} = -S_x^f$  $S_x^{\sqrt{f}} = \frac{1}{2}S_x^f$  $S_{x}^{\hat{\prod} f_{i}} = \sum_{i=1}^{k} S_{x}^{f_{i}}$ 

Some useful sensitivity theorems (cont)

 $S_x^{f/g} = S_x^f - S_x^g$  $\mathbf{S}_{\mathbf{x}}^{\sum_{i=1}^{k} \mathbf{f}_{i}} = \frac{\sum_{i=1}^{k} \mathbf{f}_{i} \mathbf{S}_{\mathbf{x}}^{\mathbf{f}_{i}}}{\sum_{i=1}^{k} \mathbf{f}_{i}}$  $S_{1/x}^{f} = -S_{x}^{f}$ 



Assume ideally R=1K, C=3.18nF so that  $I_0$ =50KHz

Actually GB=600KHz, R=1.05K, and C=3.3nF

- a) Determine an approximation to the actual unity gain frequency using a sensitivity analysis
- b) Write an analytical expression for the actual unity gain frequency



Assume ideally R=1K, C=3.18nF so that  $I_0$ =50KHz

Actually GB=600KHz, R=1.05K, and C=3.3nF

Observe

$$\frac{\Delta R}{R} = \frac{.05K}{1K} = .05$$
$$\frac{\Delta C}{C} = \frac{.12nF}{3.18nF} = .038$$
$$\frac{I_0}{GB} = \tau I_0 = \frac{50KHz}{600KHz} = .083$$



Define  $I_{0A}$  to be the actual unity gain frequency



Example:



Solution:

Still need  $\delta_{\tau}^{\mathbf{I}_{\mathsf{OA}}}\Big|_{\bar{X}_N,\bar{\mathbf{Y}}_{\mathsf{N}}=0}$ 

$$I_{A}(s) = -\frac{1}{RCs + \tau s(1 + RCs)}$$

$$I_{A}(j\omega) = -\frac{1}{-\tau RC\omega^{2} + j(\omega RC + \tau \omega)}$$

$$|I_{A}(j\omega)|^{2} = \frac{1}{(RC)^{2}\tau^{2}\omega^{4} + \omega^{2}(RC + \tau)^{2}}$$
$$|I_{A}(j\omega)|^{2} = \frac{1}{(RC)^{2}\tau^{2}\omega^{4} + \omega^{2}(RC + \tau)^{2}} = 1$$

$$(\text{RC})^2 \tau^2 I_{0A}^4 + I_{0A}^2 (\text{RC} + \tau)^2 = 1$$

Define  $I_{0A}$  to be the actual unity gain frequency

$$\boldsymbol{\mathcal{S}}_{\tau}^{\mathsf{I}_{\mathsf{OA}}}\Big|_{\boldsymbol{\bar{X}}_{N},\boldsymbol{\bar{\mathsf{Y}}}_{\mathsf{N}}=\boldsymbol{0}} = \boldsymbol{?}$$

$$\frac{(J\omega)}{(RC)^{2} \tau^{2} \omega^{4} + \omega^{2} (RC+\tau)} = \frac{1}{(RC)^{2} \tau^{2} I_{0A}^{4} + I_{OA}^{2} (RC+\tau)^{2}} = 1$$





Solution:

Still nee

Define 
$$I_{0A}$$
 to be the actual unity gain frequency  

$$\left(\operatorname{RC}\right)^{2} \tau^{2} I_{0A}^{4} + I_{OA}^{2} \left(\operatorname{RC} + \tau\right)^{2} = 1$$

$$s_{\tau}^{I_{0A}} \Big|_{\bar{X}_{N}, \bar{Y}_{N}=0} = \left(\frac{\partial I_{0A}}{\partial \tau}\right)\Big|_{\bar{X}_{N}, \bar{Y}_{N}=0}$$

$$\left(\operatorname{RC}\right)^{2} \tau^{2} 4 I_{0A}^{3} \left(\frac{\partial I_{0A}}{\partial \tau}\right) + 2\tau \left(\operatorname{RC}\right)^{2} I_{0A}^{4} + 2 I_{0A}^{1} \left(\frac{\partial I_{0A}}{\partial \tau}\right) (\operatorname{RC} + \tau)^{2} + 2(\operatorname{RC} + \tau) I_{OA}^{2} = 0$$
Evaluating at  $\bar{X}_{N}, \bar{Y}_{N} = 0$ 

$$2 I_{0}^{4} \left(\frac{\partial I_{0A}}{\partial \tau}\Big|_{\bar{X}_{N}, \bar{Y}_{N}=0}\right) (\operatorname{RC})^{2} + 2(\operatorname{RC}) I_{0}^{2} = 0$$

$$\left(\frac{\partial I_{0A}}{\partial \tau}\Big|_{\bar{X}_{N}, \bar{Y}_{N}=0}\right) = \frac{-I_{0}}{\operatorname{RC}} = s_{\tau}^{I_{0A}} \Big|_{\bar{X}_{N}, \bar{Y}_{N}=0} = -I_{0}^{2}$$



Solution:

$$\frac{dI_{0A}}{I_{0A}} = \begin{bmatrix} S_{R}^{I_{0A}} \Big|_{R_{N},C_{N},\tau=0} \end{bmatrix} \frac{dR}{R_{N}} + \begin{bmatrix} S_{C}^{I_{0A}} \Big|_{R_{N},C_{N},\tau=0} \end{bmatrix} \frac{dC}{C_{N}} + \frac{1}{I_{0N}} \left( s_{\tau}^{I_{0A}} \Big|_{\bar{X}_{N},\bar{Y}_{N}=0} \bullet \tau \right)$$

$$S_{R}^{I_{0}} \Big|_{R_{N},C_{N}} = S_{C}^{I_{0}} \Big|_{R_{N},C_{N}} = -1 \qquad s_{\tau}^{I_{0A}} \Big|_{\bar{X}_{N},\bar{Y}_{N}=0} = -I_{ON}^{2}$$

$$\frac{\Delta R}{R} = .05 \qquad \frac{\Delta C}{C} = .038 \qquad \tau I_{0} = .083$$

$$\frac{dI_{0A}}{I_{0A}} = \begin{bmatrix} -1 \end{bmatrix} .05 + \begin{bmatrix} -1 \end{bmatrix} .038 + \frac{1}{I_{0N}} \left( -I_{0N}^{2} \bullet \tau \right)$$

$$\frac{dI_{0A}}{I_{0A}} = \begin{bmatrix} -1 \end{bmatrix} .05 + \begin{bmatrix} -1 \end{bmatrix} .038 + (-.083)$$
Due to passives
$$\frac{dI_{0A}}{I_{0A}} = -.088 - .083 \qquad Due \text{ to actives}$$



Note that with the sensitivity analysis, it was not necessary to ever determine I<sub>0A</sub> !!

- a) Determine an approximation to the actual unity gain frequency using a sensitivity analysis
- b) Write an analytical expression for the actual unity gain frequency  $(\text{RC})^2 \tau^2 I_{0A}^4 + I_{OA}^2 (\text{RC} + \tau)^2 = 1$

**Must solve this quadratic for I\_{0A}** Solving, obtain  $I_{OA}$ =42.6KHz

Note this is close to the value obtained with the sensitivity analysis

Although in this simple numerical example, it may have been easier to go directly to this expression, in more complicated circuits sensitivity analysis is much easier



- Note that with the sensitivity analysis, it was not necessary to ever determine  $I_{0A}$  !!
- The sensitivity analysis was analytical, and only at the end was a numerical result obtained
- A parametric solution is usually necessary to compare different structures
- Though a closed-form analytical expression for I<sub>OA</sub> could have been obtained for this simple circuit, closed-form solutions for parameters of interest often do not exist !
- Though the active sensitivity analysis was tedious, major simplifications for active sensitivity analysis will be discussed later.

# How can sensitivity analysis be used to compare the performance of different circuits?

Circuits have many sensitivity functions

If two circuits have exactly the same number of sensitivity functions and all sensitivity functions in one circuit are lower than those in the other circuit, then the one with the lower sensitivities is a less sensitive circuit

But usually this does not happen !

Designers would like a single metric for comparing two circuits !



### **Dependent on circuit structure**

(for some circuits, also not dependent on components) **Dependent only on components** (not circuit structure)

Consider:



$$\omega_0 = \frac{1}{RC}$$





 $S_{C}^{\omega_{0}}$ **Dependent only on components** (not circuit structure)  $d\omega_0$ dx,  $S_{x_i}^{\omega_0}$  $\vec{X}$  $\omega_0$  $d\omega_0$ R  $\omega_0$ 

Dependent only on circuit structure



### **Dependent on circuit structure**

(for some circuits, also not dependent on components) **Dependent only on components** (not circuit structure)

Consider now:





Note this is dependent upon the components as well ! Actually dependent upon component ratio! Theorem: If  $f(x_1, ..., x_m)$  can be expressed as

where { $\alpha_1, \alpha_2, ..., \alpha_m$ } are real numbers, then upon any of the variables in the set { $x_1, ..., x_m$ }  $S^f_{x_i}$ 

$$f = x_1^{\alpha_1} x_2^{\alpha_2} \dots x_m^{\alpha_m}$$

is not dependent

Proof:

 $S_{\mathbf{x}_{i}}^{f} = S_{\mathbf{x}_{i}}^{\mathbf{X}_{i}^{\alpha_{i}}}$  $S_{\mathbf{x}_{i}}^{\mathbf{X}_{i}^{\alpha_{i}}} = \frac{\partial \mathbf{X}_{i}^{\alpha_{i}}}{\partial \mathbf{X}_{i}} \bullet \frac{\mathbf{X}_{i}}{\mathbf{X}_{i}^{\alpha_{i}}}$  $S_{\mathbf{x}_{i}}^{\mathbf{X}_{i}^{\alpha_{i}}} = \alpha_{i} \mathbf{X}_{i}^{\alpha_{i}-1} \bullet \frac{\mathbf{X}_{i}}{\mathbf{X}_{i}^{\alpha_{i}}}$  $S_{\mathbf{x}_{i}}^{\mathbf{X}_{i}^{\alpha_{i}}} = \alpha_{i}$ 

$$S_{x_i}^f = \alpha_i$$

It is often the case that functions of interest are of the form expressed in the hypothesis of the theorem, and in these cases the previous claim is correct Theorem: If  $f(x_1, ..., x_m)$  can be expressed as  $f = x_1^{\alpha_1} x_2^{\alpha_2} ... x_m^{\alpha_m}$ 

where  $\{\alpha_1, \alpha_2, \dots, \alpha_m\}$  are real numbers, then the sensitivity terms in

$$\frac{\mathrm{d}\mathbf{f}}{\mathbf{f}} = \sum_{i=1}^{k} \left( \mathbf{S}_{\mathbf{x}_{i}}^{\mathsf{f}} \Big|_{\bar{X}_{N}} \bullet \frac{\mathrm{d}\mathbf{x}_{i}}{\mathbf{x}_{\mathsf{iN}}} \right)$$

are dependent only upon the circuit architecture and not dependent upon the components and the right terms are dependent only upon the components and not dependent upon the architecture

This observation is useful for comparing the performance of two or more circuits where the function f shares this property

### **Metrics for Comparing Circuits**

Summed Sensitivity

$$\rho_{S} = \sum_{i=1}^{m} \mathbf{S}_{\mathbf{x}_{i}}^{\mathsf{f}}$$

Not very useful because sum can be small even when individual sensitivities are large

Schoeffler Sensitivity

$$\rho = \sum_{i=1}^{m} \left| \mathbf{S}_{\mathbf{x}_{i}}^{\mathsf{f}} \right|$$

Strictly heuristic but does differentiate circuits with low sensitivities from those with high sensitivities

### **Metrics for Comparing Circuits**

$$\rho = \sum_{i=1}^{m} \left| \mathbf{S}_{\mathbf{x}_{i}}^{\mathsf{f}} \right|$$

Often will consider several distinct sensitivity functions to consider effects of different components

$$\rho_{R} = \sum_{All \ resistors} \left| \mathbf{S}_{\mathsf{R}_{\mathsf{i}}}^{\mathsf{f}} \right|$$
$$\rho_{C} = \sum_{All \ capacitors} \left| \mathbf{S}_{\mathsf{C}_{\mathsf{i}}}^{\mathsf{f}} \right|$$
$$\rho_{OA} = \sum_{All \ capacitors} \left| \mathbf{s}_{\tau_{\mathsf{i}}}^{\mathsf{f}} \right|$$

All op amps

Homogeniety (defn)

A function f is homogeneous of order m in the n variables  $\{x_1, x_2, ..., x_n\}$  if

$$f(\lambda x_1, \lambda x_2, \dots \lambda x_n) = \lambda^m f(x_1, x_2, \dots x_n)$$

Note: f may be comprised of more than n variables

Theorem: If a function f is homogeneous of order m in the n variables  $\{x_1, x_2, ..., x_n\}$  then

$$\sum_{i=1}^{n} S_{x_{i}}^{f} = m$$

Proof:

$$\mathbf{f}(\lambda \mathbf{x}_1, \lambda \mathbf{x}_2, \dots \lambda \mathbf{x}_n) = \lambda^m \mathbf{f}(\mathbf{x}_1, \mathbf{x}_2, \dots \mathbf{x}_n)$$

Differentiate WRT  $\lambda$ 

$$\frac{\partial \left(f\left(\lambda \mathbf{x}_{1},\lambda \mathbf{x}_{2},...\lambda \mathbf{x}_{n}\right)\right)}{\partial \lambda} = m\lambda^{m-1}f\left(\mathbf{x}_{1},\mathbf{x}_{2},...\mathbf{x}_{n}\right)$$
$$\frac{\partial f}{\partial \lambda \mathbf{x}_{1}}\mathbf{x}_{1} + \frac{\partial f}{\partial \lambda \mathbf{x}_{2}}\mathbf{x}_{2} + ... + \frac{\partial f}{\partial \lambda \mathbf{x}_{n}}\mathbf{x}_{n} = m\lambda^{m-1}f\left(\mathbf{x}_{1},\mathbf{x}_{2},...\mathbf{x}_{n}\right)$$

$$\frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_{1}} \mathbf{x}_{1} + \frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_{2}} \mathbf{x}_{2} + \dots + \frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_{n}} \mathbf{x}_{n} = m\lambda^{m-1} \mathbf{f} \left( \mathbf{x}_{1}, \mathbf{x}_{2}, \dots \mathbf{x}_{n} \right)$$

Simplify notation

$$\frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_1} \mathbf{x}_1 + \frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_2} \mathbf{x}_2 + \dots + \frac{\partial \mathbf{f}}{\partial \lambda \mathbf{x}_n} \mathbf{x}_n = m\lambda^m \mathbf{f}$$

Divide by f

$$\frac{\partial \mathbf{f}}{\partial \mathbf{x}_1} \frac{\mathbf{x}_1}{\mathbf{f}} + \frac{\partial \mathbf{f}}{\partial \mathbf{x}_2} \frac{\mathbf{x}_2}{\mathbf{f}} + \dots + \frac{\partial \mathbf{f}}{\partial \mathbf{x}_n} \frac{\mathbf{x}_n}{\mathbf{f}} = m\lambda^m$$

Since true for all  $\lambda$ , also true for  $\lambda$ =1, thus

$$\frac{\partial \mathbf{f}}{\mathbf{x}_1} \frac{\mathbf{x}_1}{\mathbf{f}} + \frac{\partial \mathbf{f}}{\mathbf{x}_2} \frac{\mathbf{x}_2}{\mathbf{f}} + \dots + \frac{\partial \mathbf{f}}{\mathbf{x}_n} \frac{\mathbf{x}_n}{\mathbf{f}} = m$$

This can be expressed as

$$\sum_{i=1}^{n} S_{x_{i}}^{f} = m$$

Theorem: If a function f is homogeneous of order m in the n variables  $\{x_1, x_2, ..., x_n\}$  then

$$\sum_{i=1}^{n} S_{x_{i}}^{f} = m$$

$$\mathbf{f}(\lambda \mathbf{x}_1, \lambda \mathbf{x}_2, \dots \lambda \mathbf{x}_n) = \lambda^m \mathbf{f}(\mathbf{x}_1, \mathbf{x}_2, \dots \mathbf{x}_n)$$

The concept of homogeneity and this theorem were somewhat late to appear

Are there really any useful applications of this rather odd observation?



## Stay Safe and Stay Healthy !

## End of Lecture 20