# EE 505

# Lecture 11

- Formalization of Statistical Models
- Offset Voltages

#### Review from Last Lecture String DAC Statistical Performance

INL<sub>k</sub> assumes a maximum variance at mid-code



$$\sigma_{INLk\max} = \sigma_{\frac{R_R}{R_{NOM}}} \frac{\sqrt{N}}{2}$$

#### **Review from Last Lecture**

#### Current Steering DAC Statistical Characterization Binary Weighted

$$\sigma_{INL_{b=<1000.0>}} = \sqrt{\frac{N}{2} \left[1 - \frac{N/2}{N-1}\right]^2} + \left(\frac{N}{2} - 1\right) \left[\frac{N/2}{N-1}\right]^2} \bullet \sigma_{\frac{I_{RGK}}{I_{LSB}}}$$



$$\sigma_{\text{INL}_{\text{MAX}}} \cong \sigma_{\text{INL}_{\text{b=<1,0,...0>}}} \cong \frac{\sqrt{N}}{2} \sigma_{\frac{I_{\text{RG}}}{I_{\text{LSBX}}}}^2$$

Note this is the same result as obtained for the unary DAC

But closed form expressions do not exist for the INL of this DAC since the INL is an order statistic

### **Review from Last Lecture Statistical Modeling of Current Sources**

$$\begin{split} \sigma_{\frac{l_{DR}}{l_{DN}}} &= \sqrt{\sigma_{\frac{l_{DR}}{l_{N}}}^{2} + \sigma_{\frac{C_{OXR}}{C_{OXN}}}^{2} + 4\left(\frac{V_{THN}}{V_{GS} - V_{THN}}\right)^{2} \sigma_{V_{THR}}^{2}} \quad \text{or} \quad \sigma_{\frac{l_{DR}}{l_{DN}}} &= \sqrt{\sigma_{\frac{l_{DR}}{l_{N}}}^{2} + \sigma_{\frac{C_{OXR}}{C_{OXN}}}^{2} + \left(\frac{2}{V_{GS} - V_{THN}}\right)^{2} \sigma_{V_{THR}}^{2}} \\ It will be assumed that (will discuss assumption later) \quad \sigma_{\frac{l_{DR}}{l_{N}}}^{2} = \frac{A_{\mu}^{2}}{WL} \\ &\sigma_{\frac{C_{OXR}}{C_{OXR}}}^{2} = \frac{A_{C_{OX}}^{2}}{WL} \quad \text{where } A_{\mu}, A_{Cox}, A_{VT0} \text{ are Pelgrom process parameters} \\ &\sigma_{V_{THR}}^{2} = \frac{A_{VT0}^{2}}{WL} \\ &\sigma_{\frac{l_{DR}}{l_{DN}}}^{2} = \frac{1}{\sqrt{WL}} \sqrt{A_{\mu}^{2} + A_{Cox}^{2} + \frac{4}{V_{EB}^{2}}} A_{VT0}^{2} \\ Define \quad A_{\beta} = \sqrt{A_{\mu}^{2} + A_{Cox}^{2}} \\ Thus \quad \sigma_{\frac{l_{DR}}{l_{DN}}}^{2} = \frac{1}{\sqrt{WL}} \sqrt{A_{\beta}^{2} + \frac{4}{V_{EB}^{2}}} A_{VT0}^{2} \\ Often only A_{\beta} \text{ is available} \end{split}$$

I<sub>DN</sub>

#### Review from Last Lecture Statistical Modeling of Current Sources

$$\sigma_{\frac{I_{DR}}{I_{DN}}} = \frac{1}{\sqrt{WL}} \sqrt{A_{\beta}^2 + \frac{4}{V_{EB}^2} A_{VT0}^2}$$

Gate area: A=WL

- Standard deviation decreases with  $\sqrt{A}$
- Large V<sub>EB</sub> reduces standard deviation
- Operating near cutoff results in large mismatch
- Often threshold voltage variations dominate mismatch

$$\sigma_{\underline{I_{DR}}\atop{I_{DN}}} \cong \frac{2}{V_{EB}\sqrt{WL}} A_{VT0}$$

Theorem: If the random part of two uncorrelated current sources  $I_1$  and  $I_2$  are identically distributed with normalized variance,  $\sigma_{I_R/I_N}^2$  then the random

variable  $\Delta I = I_2 - I_1$  has a variance given by the equation  $\sigma_{\Delta I_{I_N}}^2 = 2\sigma_{I_R_{I_N}}^2$ 

Proof:  

$$\Delta I = I_{1} - I_{2}$$

$$\frac{\Delta I}{I_{N}} = \frac{I_{1}}{I_{N}} - \frac{I_{2}}{I_{N}}$$

$$\frac{\Delta I}{I_{N}} = \frac{I_{N} + I_{R1}}{I_{N}} - \frac{I_{N} + I_{R2}}{I_{N}} = \frac{I_{R1}}{I_{N}} - \frac{I_{R2}}{I_{N}}$$

$$\sigma_{\frac{\Delta I}{I_{N}}}^{2} = \sigma_{\frac{I_{R}}{I_{N}}}^{2} + \sigma_{\frac{I_{R}}{I_{N}}}^{2} = 2\sigma_{\frac{I_{R}}{I_{N}}}^{2}$$

The previous statistical analysis was somewhat tedious

# Will try to formalize the process for obtaining two important statistics, the <u>mean</u> and <u>variance</u>, of a function of interest

Assume Y is a function of n uncorrelated random variables  $x_{R1},...x_{Rn}$  where the mean and variance of  $x_{Ri}$  are "small"



pdf of the random part of Y is invariably highly nonlinear joint function of a large number of random variables

Recall if 
$$(x_{R1}, x_{R2}, ..., x_{Rn})$$
 uncorrelated and  $f = \sum_{i=1}^{m} a_i x_{Ri}$  then  $\sigma_f^2 = \sum_{i=1}^{m} a_i^2 \sigma_{X_{Ri}}^2$ 

Since random variables are invariably small, will try to linearize the dependence of the random variables on Y and use previous theorem to obtain  $\mu$  and  $~\sigma$ 

$$\mathbf{Y} = \mathbf{f}\left(x_{R1}, x_{R2}, \dots x_{Rn}\right)$$



Assuming means are all 0, Y can be expressed in a Taylor's series expanded around mean as

$$\mathbf{Y} = f(X)\Big|_{X_R=0} + \sum_{j=1}^n \frac{\partial f}{\partial x_{Rj}}\Big|_{X_R=0} x_{Rj} + \varepsilon(x_{R1}, x_{R2}, \dots, x_{Rn})$$

where  $\varepsilon(x_{R1}, x_{R2}, ..., x_{Rn})$  is due to higher-order terms and is small

$$\mathbf{Y} \simeq f(X)\Big|_{X_R=0} + \sum_{j=1}^n \frac{\partial f}{\partial x_{Rj}}\Big|_{X_R=0} x_{Rj}$$

Note power series expansion linearized Y in the variables  $(x_{R1}, x_{R2}, ..., x_{Rn})$ 

$$\frac{\mathbf{Y}}{\mathbf{Y}_{\mathrm{N}}} = \frac{f(X)\big|_{X_{R}=0}}{\mathbf{Y}_{\mathrm{N}}} + \sum_{j=1}^{n} \frac{1}{\mathbf{Y}_{\mathrm{N}}} \frac{\partial f}{\partial x_{Rj}}\Big|_{X_{R}=0} x_{Rj}$$

From Theorem:

$$\sigma_{\frac{Y}{Y_{N}}}^{2} = \sum_{j=1}^{n} \left( \left( \frac{1}{Y_{N}} \frac{\partial f}{\partial x_{Rj}} \right|_{X_{R}=0} \right)^{2} \sigma_{x_{Rj}}^{2} \right)$$

Define:

$$\hat{S}_{xRj}^{f} = \frac{1}{Y_{N}} \frac{\partial f}{\partial x_{Rj}} \bigg|_{X_{R}=0}$$
$$\sigma_{\frac{Y}{Y_{N}}}^{2} = \sum_{j=1}^{n} \left( \left[ \hat{S}_{xRj}^{f} \right]^{2} \sigma_{x_{Rj}}^{2} \right)$$

But  

$$\sigma_{\frac{Y}{Y_{N}}}^{2} = \sum_{j=1}^{n} \left( \left[ \hat{S}_{xRj}^{f} \right]^{2} \sigma_{x_{Rj}}^{2} \right) = \sum_{j=1}^{n} \left( \left[ \hat{S}_{xRj}^{f} \right]^{2} \left( \frac{x_{jN}}{x_{jN}} \right)^{2} \sigma_{x_{Rj}}^{2} \right) = \sum_{j=1}^{n} \left( \left[ x_{jN} \hat{S}_{xRj}^{f} \right]^{2} \sigma_{\frac{x_{Rj}}{x_{jN}}}^{2} \right)$$
Alternatively, from the more standard definition of sensitivity  $S_{xR_{j}}^{f}$ :

we thus obtain

$$\sigma_{\frac{\mathbf{Y}}{\mathbf{Y}_{N}}}^{2} = \sum_{j=1}^{n} \left( \left[ S_{xRj}^{f} \right]^{2} \sigma_{\frac{x_{Rj}}{x_{Ni}}}^{2} \right)$$



- Determine sensitivity function by analyzing circuit
- Determine variances by characterizing process

This approach is a formalized approach to statistical analysis that is more systematic than the ad hoc approach used in last lecture

Will now focus on characterizing the process parameters

#### Theorem:

For any arbitrarily shaped transistor there exists a rectangular transistor that has the same static I-V characteristics



Only channel is shown which is the intersection of Poly and Active

### Arbitrarily Shaped Transistor



# Claim: $\sigma_{\frac{\mu_R}{\mu_N}}^2 = \frac{A_{\mu}^2}{A}$

where  $A_{\mu}$  is the Pelgrom matching parameter and A is the gate area

### Argument: Ak Assume (define) $\mu_{eq} = \frac{\int \mu(x, y) dx dy}{\Lambda}$ Channel Let $S_k$ be a square of area $A_k$ in the channel Х $=\frac{\sum_{i=1}^{N}\int_{A_{ki}}\mu(x,y)dxdy}{\mathbf{A}}$

where the channel has been partitioned into N disjoint regions each of area Aki

For convenience, assume  $A_{ki}=A_{kj}=A_k$  for all i,j



#### Concept can be extended so now have:

$$\sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2} = \frac{A_{\mu}^{2}}{WL}$$

$$\sigma_{\frac{C_{OXR}}{C_{OXN}}}^{2} = \frac{A_{Cox}^{2}}{WL}$$

$$\sigma_{\frac{C_{OXR}}{C_{OXN}}}^{2} = \frac{A_{Cox}^{2}}{WL}$$

$$\sigma_{\frac{C_{OXR}}{V_{THR}}}^{2} = \frac{A_{VT0}^{2}}{WL}$$

where  $A_{\mu}, A_{Cox}, A_{VT0}$  ,  $A_{R\,\Box}\,$  are Pelgrom process parameters

### **Statistical Simulations**

Often simulations are used to predict statistical performance of a circuit

Variable of interest are often Gaussian (e.,g. R<sub>R</sub>, C<sub>R</sub>, V<sub>OSR</sub>, I<sub>R</sub>,...)

Most CAD tools do not have a rich set of random variable distributions (maybe not even the Gaussian distribution)

Many tools only have a single random variable generator that is U [0,1]

Theorem: f(y) and F(y) are any pdf/cdf pair and if X~ U[0,1], then  $y = F^{-1}(x)$  has a pdf of f(y).

Corollary: If h is a rv with 
$$F(h) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{h} e^{-\frac{x^2}{2}} dx$$

then  $y=F^{-1}(h)$  is N[0,1]



#### Theorem: If $y \sim N[0,1]$ , then $z = \sigma y + \mu$ is $N[\mu,\sigma]$

$$=(h)=\frac{1}{2}\left[1+\mathrm{erf}\left(\frac{h-\mu}{\sigma\sqrt{2}}\right)\right]$$

### In Excel:

NORM.S.INV(h)= $F^{-1}(h)$  where

$$F(h) = \int_{-\infty}^{h} \frac{1}{\sqrt{2}\pi} e^{-\frac{x^2}{2}} dx$$

NORM.DIST(h, $\mu$ , $\sigma$ ,TRUE) = f(h) where

$$f(h) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{h-\mu}{\sigma}\right)^2}$$

NORM.DIST(h,
$$\mu$$
, $\sigma$ ,FALSE) = F(h) where

$$F(h) = \int_{-\infty}^{h} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^{2}} dx$$

Some useful relationships:

$$\mathrm{ERF}(\mathbf{x}) = \frac{2}{\pi} \int_{0}^{\mathbf{x}} \mathrm{e}^{-t^{2}} \mathrm{d}t$$

The CDF of the N(0,1) random variable x is given by

$$F_N(x) = \frac{1}{2} \left( 1 + ERF\left(\frac{x}{\sqrt{2}}\right) \right)$$

Excel	Older Excel	
@NORM.DIST		f(x)
@NORM.S.DIST		$f_N(x)$
@NORM.INV	@NORMINV	$F^{-1}(x)$
@NORM.S.INV	@NORMSINV	$F_N^{-1}(x)$
	@NORMDIST	$F_{N}(x)$
	@NORMINV	F(x)

where f: PDF F:CDF

Example: Determine the area required for the resistors for an n-bit R-string DAC to achieve a yield of P if the device is marketable provided  $|INL_{KMAX}| < \frac{1}{2}LSB$ 

Solution:

Want:  $P = \int_{x=-\frac{1}{2}}^{x=+\frac{1}{2}} f(INL_{kMAX}) dx$ Let  $X_N = \frac{X}{\sigma_{INL_{kMAX}}}$  recall  $\mu_{INL_{kMAX}} = 0$  $\sigma_{INL_{kMAX}} = \frac{\sqrt{N}}{2} \sigma_{\frac{R_R}{R_N}} \longrightarrow X_N = \frac{\frac{1}{2}}{\frac{\sqrt{N}}{2} \sigma_{\frac{R_R}{R_N}}} = \frac{1}{\sqrt{N} \sigma_{\frac{R_R}{R_N}}}$ thus  $P = \int_{V_N}^{X_N} f_N(x) dx$   $f_N \sim N(0,1)$ Since P is fixed, can solve for  $X_N$ 

 $P = 2F_N(X_N) - 1 \longrightarrow X_N = F_N^{-1}\left(\frac{P+1}{2}\right)$ 

where  $F_N(X_N)$  is the CDF of a N(0,1) rv

$$X_{N} = \frac{1}{\sigma_{\frac{R_{R}}{R_{N}}}\sqrt{N}} \quad \text{recall} \quad \sigma_{\frac{R_{R}}{R_{N}}} = \frac{A_{R}}{\sqrt{WL}}$$
  
thus  $X_{N} = \frac{\sqrt{WL}}{A_{R}\sqrt{N}} \quad \longrightarrow \quad \sqrt{WL} = A_{R}\sqrt{N}X_{N}$   
thus, we obtain  $\sqrt{WL} = A_{R}\sqrt{N} \bullet F_{N}^{-1}\left(\frac{P+1}{2}\right)$ 

Since there are N=2<sup>n</sup> resistors, total area becomes

$$A_{TOT} = 2^{n} \sqrt{WL} = 2^{n} A_{R} \sqrt{N} \bullet F_{N}^{-1} \left(\frac{P+1}{2}\right) = 2^{n} A_{R} 2^{\frac{n}{2}} \bullet F_{N}^{-1} \left(\frac{P+1}{2}\right)$$

or equivalently

$$A_{TOT} = \sqrt{2^{3n}} A_R \bullet F_N^{-1} \left(\frac{P+1}{2}\right)$$

# All ADCs have comparators and many ADCs and DACs have operational amplifiers

The offset voltages of both amplifiers and comparators are random variables and invariably are key factors affecting the performance of a data converter

Operational Amplifiers:

Generally differential amplifiers whose offset is dominantly determined by randomness in the first stage

Comparators:

High Gain Operational Amplifiers Latching Structures (often clocked) Combination of High Gain Amplifiers and Latching Structures

- Offset voltages of high-gain amplifiers well understood
- Offset voltage of Latching Structures often difficult to determine and can be very large

**Consider First Offset in Operational Amplifiers** 



Input-referred Offset Voltage: Differential Voltage that must be applied to the input to make the output assume its <u>desired value</u>

With a good design, a designer will have  $V_{OUT}$  at the desired value if the components assume the values used in the design

Any difference in the output from what is desired when components assume the nominal values used in a design is attributable to a systematic offset voltage

Two types of offset voltage:

- Systematic Offset Voltage
- Random Offset Voltage



Definition: The output offset voltage is the difference between the desired output and the actual output when  $V_{id}$ =0 and  $V_{ic}$  is the quiescent common-mode input voltage.

#### VOUTOFF = VOUT - VOUTDES

Note:  $V_{OUTOFF}$  is dependent upon  $V_{ICQ}$  although this dependence is usually quite weak and often not specified



Definition: The input-referred offset voltage is the differential dc input voltage that must be applied to obtain the desired output when  $V_{ic}$  is the quiescent common-mode input voltage.

Note: V<sub>OFF</sub> is usually related to the output offset voltage by the expression

$$V_{OFF} = \frac{V_{OUTOFF}}{A_D}$$

Note:  $V_{OFF}$  is dependent upon  $V_{ICQ}$  although this dependence is usually quite weak and often not specified

Two types of offset voltage:

- Systematic Offset Voltage
- Random Offset Voltage



After fabrication it is impossible (difficult) to distinguish between the systematic offset and the random offset in any individual op amp

Measurements of offset voltages for a large number of devices will provide mechanism for identifying systematic offset and statistical Characteristics of the random offset voltage

#### Systematic Offset Voltage

Offset voltage that is present if all device and model parameters assume their nominal value

Easy to simulate the systematic offset voltage

Almost always the designer's responsibility to make systematic offset voltage very small

Generally easy to make the systematic offset voltage small

Can tweak out systematic offset after design is almost done

#### Random Offset Voltage

Due to random variations in process parameters and device dimensions

Random offset is actually a random variable at the design level but deterministic after fabrication in any specific device

Distribution of native offset nearly Gaussian (If offset compensation is not employed)

Has zero mean

Characterized by its standard deviation or variance

Often strongly layout and area dependent



Can be modeled as a dc voltage source in series with the input (on either terminal)

Effects of Offset Voltage - an example



Desired I/O relationship







Effects can be reduced or eliminated by adding equal amplitude opposite phase DC signal (many ways to do this)

One such technique is "dynamic offset compensation"

Widely used in offset-critical applications

Comes at considerable effort and expense

Prefer to have designer make V<sub>OS</sub> small in the first place though penalty (e.g. cost) for making it sufficiently small without correction is often unacceptable

# **Dynamic Offset Compensation**





Fig. 2.2 Auto rawood amplifiar with input offect storage

Most basic dynamic offset compensation at input

# Effects of Offset Voltage

- Deviations in performance will change from one instantiation to another due to the random component of the offset
- Particularly problematic in high-gain circuits
- A major problem in many other applications
- Not of concern in many applications as well

# **Offset Voltage Distribution**



Typical histogram of native offset voltage (binned) after fabrication

# **Offset Voltage Distribution**



Typical histogram of offset voltage (binned) after fabrication Mean is nearly 0 (actually the systematic offset voltage)

# **Offset Voltage Distribution**



Typical histogram of offset voltage (binned) in shipped parts when entire population used for a single produce

Extreme offset parts have been sifted at test



Typical histogram of offset voltage (binned) in shipped parts

Low-offset parts sold at a premium

Extreme offset parts have been sifted at test

Consider as an example:



Ideally  $R_1 = R_2 = R_N$ ,  $M_1$  and  $M_2$  are matched

$$V_{OUT} = V_{DD} - \left(\frac{I_T}{2}\right) R_N$$

Assume this is the desired output voltage (note not assuming  $V_{OUT-DES}=0V$ )

Consider as an example:



If everything ideal except R<sub>1</sub> and R<sub>2</sub>

 $R_1 = R_N + R_{R1}$   $R_2 = R_N + R_{R2}$ 

Thus at the design stage,  $V_{\mbox{\scriptsize OUT}}$  is also a random variable

$$V_{OUT} = V_{DD} - \left(\frac{I_T}{2}\right) \left[R_N + R_{R2}\right]$$
$$V_{OUT-R} = -\left(\frac{I_T}{2}\right) R_{R2}$$

Consider as an example:



Determine the offset voltage – i.e. value of  $V_X$  needed to obtain desired output



Setting  $V_{OUT}=V_{OUT-DES}$  and solving for  $V_X$ , we obtain

$$V_{X} = V_{OFF} = \frac{-1}{A_{V}} \left(\frac{I_{T}}{2}\right) R_{R2}$$

Determine the offset voltage -i.e. value of V<sub>x</sub> needed to obtain desired output



Determine the offset voltage -i.e. value of V<sub>X</sub> needed to obtain desired output



$$\sigma_{V_{OS}} = V_{EB} \frac{A_R}{\sqrt{A}}$$

Thus

The random offset voltage is almost entirely that of the input stage in most op amps





n-channel MOSFET

Impurities vary randomly with position as do edges of gate, oxide and diffusions

Model and design parameters vary throughout channel and thus the corresponding equivalent lumped model parameters will vary from device to device

The random offset is due to missmatches in the four transistors, dominantly missmatches in the parameters {V<sub>T</sub>,  $\mu$ ,C<sub>OX</sub>,W and L}

The relative missmatch effects become more pronounced as devices become smaller

 $V_{Ti} = V_{TN} + V_{TRi}$   $C_{OXi} = C_{OXN} + C_{OXRi}$   $\mu_i = \mu_N + \mu_{Ri}$   $W_i = W_N + W_{Ri}$   $L_i = L_N + L_{Ri}$ 

Each design and model parameter is comprised of a nominal part and a random component





For each device, the device model is often expressed as

$$I_{Di} = \frac{(\mu_{N} + \mu_{Ri})(C_{OXN} + C_{OXRi})(W_{N} + W_{Ri})}{2(L_{N} + L_{Ri})} (V_{GSi} - (V_{TN} + V_{TRi}))^{2} (1 + (\lambda_{N} + \lambda_{Ri})[V_{DS}])$$

Because of the random components of the parameters in every device, matching from the left-half circuit to the right half-circuit is not perfect

This mismatch introduces an offset voltage which is a random variable



Assume currents at output node must satisfy relation  $I_2 = I_4$ 

Strategy:

- 1) Obtain expression for  $V_{OFF}$  (referred to input) that forces  $I_2=I_4$
- 2) Linearize expression in terms of design variables and decorrelate
- 3) Obtain  $\sigma_{VOS}$

Analysis of Offset Voltage (Neglect R<sub>L</sub>)

$$I_{D1} = \frac{\mu_{n1}C_{OX1}W_{1}}{2L_{1}} (V_{OFF} + V_{INC} - V_{S} - V_{TH1})^{2}$$

$$I_{D2} = \frac{\mu_{n2}C_{OX2}W_{2}}{2L_{2}} (V_{INC} - V_{S} - V_{TH2})^{2}$$

$$I_{D3} = \frac{\mu_{p3}C_{OX3}W_{3}}{2L_{3}} (V_{X} - V_{DD} - V_{TH3})^{2}$$

$$I_{D4} = \frac{\mu_{p4}C_{OX4}W_{4}}{2L_{4}} (V_{X} - V_{DD} - V_{TH4})^{2}$$



Since 
$$\sqrt{I_{D1}} = \sqrt{I_{D3}}$$
  
 $V_{OFF} + V_{INC} - V_{S} - V_{TH1} = \sqrt{\frac{\mu_{p3}C_{OX3}W_{3}L_{1}}{\mu_{n1}C_{OX1}W_{1}L_{3}}} (V_{X} - V_{DD} - V_{TH3})$ 

Since  $\sqrt{I_{D2}} = \sqrt{I_{D4}}$  $V_{INC} - V_S - V_{TH2} = \sqrt{\frac{\mu_{p4}C_{OX4}W_4L_2}{\mu_{n2}C_{OX2}W_22L_4}} (V_X - V_{DD} - V_{TH4})$ 

Define:  
$$\boldsymbol{a} = \sqrt{\frac{L_1 \mu_{p3} C_{OX3} W_3}{L_3 \mu_{n1} C_{OX1} W_1}} \qquad \boldsymbol{b} = \sqrt{\frac{L_2 \mu_{p4} C_{OX4} W_4}{L_4 \mu_{n2} C_{OX2} W_2}}$$

Substituting for a and b, it follows on eliminating  $V_s$  that

$$V_{OFF} = V_{TH1} - V_{TH2} + (a - b)(V_X - V_{DD}) + bV_{TH4} - aV_{TH3}$$
  

$$V_X = V_{XN} - V_{XR}$$
  

$$a = a_N + a_R$$
  

$$b = b_N + b_R$$



Assume

Observe 
$$a_N = b_N$$
 and  $V_{XN} - V_{DD} - V_{TpN} = V_{EB3}$ 

Since the random part of  $V_X$  multiplies only a-b which is small, it follows that

$$V_{OFF} = V_{TH1} - V_{TH2} + (a - b)(V_{EB3N}) + bV_{TH4} - aV_{TH3}$$
$$V_{OFF} = V_{THR1} - V_{THR2} + (a_R - b_R)V_{EB3N} + a_N(V_{THR4} - V_{THR3})$$
$$\sigma_{V_{OFF}}^2 = 2\sigma_{V_{TnR}}^2 + a_N^2 2\sigma_{V_{TpR}}^2 + V_{EB3N}^2 \sigma_{V_{a_R-b_R}}^2$$

Will now obtain a<sub>R</sub> and b<sub>R</sub>

$$V_{OFF} = V_{TnR2} - V_{TnR2} + (b_R - a_R)V_{EB3} + a_N(V_{TpR3} - V_{TpR4})$$

$$a = \sqrt{\frac{(L_{N1} + L_{R1})(\mu_{Np3} + \mu_{R3})(C_{OXN3} + C_{OXR3})(W_{N3} + W_{R3})}{(L_{N3} + L_{R3})(\mu_{Nn1} + \mu_{R1})(C_{OXN1} + C_{OXR1})(W_{N1} + W_{R1})}} \bigvee_{Vorf} + \frac{W_{A3}}{(V_{N3} + U_{A3})} + \frac{W_{A4}}{(V_{A3} + U_{A3})(\mu_{A3} + \mu_{A3})}$$
Recall for x small,  $\sqrt{1 + x} \equiv 1 + \frac{x}{2}$   $\frac{1}{1 + x} \equiv 1 - x$ 

$$a = \sqrt{\frac{(L_{N1} \mu_{Np3} W_{N3})}{(L_{N3} \mu_{Nn1} W_{N1})}} \left(1 + \frac{1}{2} \left[\frac{L_{R1}}{L_{N1}} - \frac{L_{R3}}{L_{N3}} + \frac{\mu_{R3}}{\mu_{N3}} - \frac{\mu_{R1}}{\mu_{N1}} + \frac{C_{OXR3}}{C_{OXN3}} - \frac{C_{OXR1}}{C_{OXN1}} + \frac{W_{R3}}{W_{N3}} - \frac{W_{R3}}{W_{N3}}\right]\right)$$
Thus
$$a_R = \sqrt{\frac{(L_{N1} \mu_{Np3} W_{N3})}{(L_{N3} \mu_{Nn1} W_{N1})}} \frac{1}{2} \left[\frac{L_{R1}}{L_{N1}} - \frac{L_{R3}}{L_{N3}} + \frac{\mu_{R3}}{\mu_{N3}} - \frac{\mu_{R1}}{\mu_{N1}} + \frac{C_{OXR3}}{C_{OXN3}} - \frac{C_{OXR1}}{C_{OXN1}} + \frac{W_{R3}}{W_{N3}} - \frac{W_{R3}}{W_{N3}}\right]$$

$$a_M = \sqrt{\frac{(L_{N1} \mu_{Np3} W_{N3})}{(L_{N3} \mu_{Nn1} W_{N1})}}$$

V<sub>DD</sub>

 $u_{N} = \sqrt{(L_{N3}\mu_{Nn1}W_{N1})}$ Likewise

$$b_{R} = \sqrt{\frac{\left(L_{N1}\mu_{Np3}W_{N3}\right)}{\left(L_{N3}\mu_{Nn1}W_{N1}\right)}}\frac{1}{2}\left[\frac{L_{R2}}{L_{N2}} - \frac{L_{R4}}{L_{N4}} + \frac{\mu_{R4}}{\mu_{Np4}} - \frac{\mu_{R2}}{\mu_{Nn2}} + \frac{C_{OXR4}}{C_{OXN4}} - \frac{C_{OXR2}}{C_{OXN2}} + \frac{W_{R4}}{W_{N4}} - \frac{W_{R2}}{W_{N2}}\right]$$



$$\boldsymbol{a}_{R} - \boldsymbol{b}_{R} = \sqrt{\frac{\left(L_{N1}\mu_{Np3}W_{N3}\right)}{\left(L_{N3}\mu_{Nn1}W_{N1}\right)}} \frac{1}{2} \begin{bmatrix} \frac{L_{R1}}{L_{N1}} - \frac{L_{R2}}{L_{N2}} + \frac{L_{R4}}{L_{N4}} - \frac{L_{R3}}{L_{N3}} + \frac{\mu_{R3}}{\mu_{Np3}} - \frac{\mu_{R4}}{\mu_{Np4}} + \frac{\mu_{R2}}{\mu_{Np2}} - \frac{\mu_{R1}}{\mu_{Nn1}} \\ + \frac{C_{OXR3}}{C_{OXN3}} - \frac{C_{OXR4}}{C_{OXN4}} + \frac{C_{OXR2}}{C_{OXN2}} - \frac{C_{OXR1}}{C_{OXN1}} + \frac{W_{R3}}{W_{N3}} - \frac{W_{R4}}{W_{N4}} + \frac{W_{R2}}{W_{N2}} - \frac{W_{R3}}{W_{N3}} \end{bmatrix} \\ \sigma_{\boldsymbol{a}_{R}-\boldsymbol{b}_{R}}^{2} = \frac{\left(L_{N1}\mu_{Np3}W_{N3}\right)}{\left(L_{N3}\mu_{Nn1}W_{N1}\right)} \frac{1}{2} \begin{bmatrix} \sigma_{\frac{L_{R1}}{L_{N1}}}^{2} + \sigma_{\frac{L_{R3}}{L_{N3}}}^{2} + \sigma_{\frac{\mu_{R3}}{\mu_{N3}}}^{2} + \sigma_{\frac{\mu_{R3}}{\mu_{N2}}}^{2} + \sigma_{\frac{C_{OXR4}}{C_{OXN4}}}^{2} + \sigma_$$

#### Thus

$$\sigma_{V_{OFF}}^{2} = 2\sigma_{V_{TnR2}}^{2} + 2\frac{L_{N1}\mu_{Np3}W_{N3}}{L_{N3}\mu_{Nn1}W_{N1}}\sigma_{V_{TpR3}}^{2} + V_{EB3}^{2}\frac{\left(L_{N1}\mu_{Np3}W_{N3}\right)}{\left(L_{N3}\mu_{Nn1}W_{N1}\right)}\frac{1}{2}\left[\sigma_{\frac{L_{R1}}{L_{N1}}}^{2} + \sigma_{\frac{L_{R3}}{L_{N3}}}^{2} + \sigma_{\frac{\mu_{R3}}{\mu_{Np3}}}^{2} + \sigma_{\frac{\mu_{R2}}{\mu_{Np3}}}^{2} + \sigma_{\frac{C_{OXR3}}{C_{OXN3}}}^{2} + \sigma_{\frac{W_{R3}}{W_{N3}}}^{2} + \sigma_{\frac{W_{R3}}{W_{N3$$

but

 $\sigma_{V_{T}}^{2} = \frac{A_{VT0}^{2}}{WL}$ 



$$\sigma_{V_{OFF}}^{2} = 2 \frac{A_{VTn0}^{2}}{W_{1}L_{1}} + 2 \frac{\mu_{p}L_{1}}{\mu_{n}W_{1}} \frac{A_{VTp0}^{2}}{L_{3}^{2}} + V_{EB3}^{2} \frac{\mu_{p}L_{1}W_{3}}{\mu_{n}L_{3}W_{1}} \frac{1}{2} \left[ \frac{A_{\mu_{n}}^{2}}{W_{3}L_{3}} + \frac{A_{\mu_{p}}^{2}}{W_{1}L_{1}} + A_{Cox}^{2} \left( \frac{1}{W_{3}L_{3}} + \frac{1}{W_{1}L_{1}} \right) + A_{W}^{2} \left( \frac{2}{W_{3}^{2}L_{3}} + \frac{2}{W_{1}^{2}L_{1}} \right) + A_{L}^{2} \left( \frac{2}{W_{1}L_{1}^{2}} + \frac{2}{W_{3}L_{3}^{2}} \right) \right]$$

Often this can be approximated by

$$\sigma_{V_{OFF}}^{2} = 2\frac{A_{VTn0}^{2}}{W_{1}L_{1}} + 2\frac{\mu_{p}L_{1}}{\mu_{n}W_{1}}\frac{A_{VTp0}^{2}}{L_{3}^{2}} + V_{EB3}^{2}\frac{\mu_{p}L_{1}W_{3}}{\mu_{n}L_{3}W_{1}}\frac{1}{2}\left[\frac{A_{\mu_{n}}^{2}}{W_{3}L_{3}} + \frac{A_{\mu_{p}}^{2}}{W_{1}L_{1}} + A_{Cox}^{2}\left(\frac{1}{W_{3}L_{3}} + \frac{1}{W_{1}L_{1}}\right)\right]$$

Or even approximated by

$$\sigma_{V_{OFF}}^{2} = 2\frac{A_{VTn0}^{2}}{W_{1}L_{1}} + 2\frac{\mu_{p}L_{1}}{\mu_{n}W_{1}}\frac{A_{VTp0}^{2}}{L_{3}^{2}}$$

Since  $V_{EBn}$  and  $V_{EBp}$  are related, this is often expressed in simpler form as:

$$\sigma_{V_{OS}}^{2} = 2 \left[ \frac{A_{VTOn}^{2} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTOp}^{2} + \frac{V_{EBn}^{2}}{4} \left( \frac{\frac{1}{W_{n}L_{n}} A_{\mu_{n}}^{2} + \frac{1}{W_{p}L_{p}} A_{\mu_{p}}^{2} + A_{COX}^{2} \left[ \frac{1}{W_{n}L_{n}} + \frac{1}{W_{p}L_{p}} \right] \right) \right] + 2A_{L}^{2} \left[ \frac{1}{W_{n}L_{n}^{2}} + \frac{1}{W_{p}L_{p}^{2}} \right] + A_{W}^{2} \left[ \frac{1}{L_{n}W_{n}^{2}} + \frac{1}{L_{p}W_{p}^{2}} \right] \right]$$

where the terms  $A_{VT0}$ ,  $A_{\mu}$ ,  $A_{COX}$ ,  $A_{L}$ , and  $A_{W}$  are process parameters  $V_{DD}$   $A_{VT0} \approx \begin{cases} 21mV \cdot \mu & (n-ch) \\ 25mV \cdot \mu & (p-ch) \end{cases}$   $\sqrt{A_{\mu}^{2} + A_{C_{OX}}^{2}} \approx \begin{cases} .016\mu & (n-ch) \\ .023\mu & (p-ch) \end{cases}$   $A_{L} = A_{W} \approx 0.017\mu^{\frac{3}{2}}$ Usually the  $A_{VT0}$  terms are dominant, thus the variance simplifies to

↓ ) I<sub>T</sub>

$$\sigma_{V_{OS}}^{2} \cong 2 \left[ \frac{A_{VTO\,n}^{2}}{W_{n}L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTO\,p}^{2} \right]$$

Correspondingly:

$$\sigma_{V_{OS}}^{2} = 2 \left[ \frac{A_{VTOn}^{2}}{W_{n}L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTOp}^{2} + \frac{V_{EBn}^{2}}{4} \left( \frac{1}{W_{n}L_{n}} A_{\mu_{n}}^{2} + \frac{1}{W_{p}L_{p}} A_{\mu_{p}}^{2} + A_{COX}^{2} \left[ \frac{1}{W_{n}L_{n}} + \frac{1}{W_{p}L_{p}} \right] \right) + 2A_{L}^{2} \left[ \frac{1}{W_{n}L_{n}^{2}} + \frac{1}{W_{p}L_{p}^{2}} \right] + A_{w}^{2} \left[ \frac{1}{L_{n}W_{n}^{2}} + \frac{1}{L_{p}W_{p}^{2}} \right] \right) \right]$$

which again simplifies to

$$\sigma_{V_{OS}}^{2} \cong 2 \left[ \frac{A_{VTO n}^{2}}{W_{n} L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{VTO p}^{2} \right]$$

Note these offset voltage expressions are identical!



Example: Determine the  $3\sigma$  value of the input offset voltage for The MOS differential amplifier is a)  $M_1$  and  $M_3$  are minimum-sized and b) the area of  $M_1$  and  $M_3$  are 100 times minimum size

Assume  $L_{MIN}=W_{MIN}=0.5u$ ,  $A_{VTOn}=0.021V$  and  $A_{VTOP}=0.025V$ 

$$\sigma_{V_{OS}}^{2} \cong 2 \left[ \frac{A_{VTOn}^{2}}{W_{n}L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTOp}^{2} \right]$$
$$\sigma_{V_{OS}}^{2} \cong \frac{2}{W_{n}L_{n}} \left[ A_{VTOn}^{2} + \frac{\mu_{p}}{\mu_{n}} A_{VTOp}^{2} \right]$$
$$a) \qquad \sigma_{V_{OS}}^{2} \cong \frac{2}{(0.5\mu)^{2}} \left[ .021^{2} + \frac{1}{3} .025^{2} \right]$$
$$\sigma_{V_{OS}}^{2} \cong 72mV$$

 $3 \sigma_{V_{OS}} \cong 216 mV$ 

Note this is a very large offset voltage !



Example: Determine the  $3\sigma$  value of the input offset voltage for The MOS differential amplifier is

a)  $M_1$  and  $M_3$  are minimum-sized and

b) the area of  $M_1$  and  $M_3$  are 100 times minimum size

Assume  $L_{MIN}=W_{MIN}=0.5u$ ,  $A_{VTOn}=0.021V$  and  $A_{VTOP}=0.025V$ 

$$\sigma_{V_{OS}}^{2} \cong 2 \left[ \frac{A_{VTOn}^{2}}{W_{n}L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTOp}^{2} \right]$$
$$\sigma_{V_{OS}}^{2} \cong \frac{2}{W_{n}L_{n}} \left[ A_{VTOn}^{2} + \frac{\mu_{p}}{\mu_{n}} A_{VTOp}^{2} \right]$$
$$b)$$

$$\sigma_{V_{OS}}^{2} \cong \frac{2}{\left[100(0.5\mu)\right]^{2}} \left[ .021^{2} + \frac{1}{3}.025^{2} \right]$$

$$\sigma_{V_{OS}} \cong 7.2 mV$$

 $3 \; \sigma_{V_{OS}} \cong 21.6 mV$ 

Note this is much lower but still a large offset voltage !

The area of M<sub>1</sub> and M<sub>3</sub> needs to be very large to achieve a low offset voltage





It can be shown that

$$\sigma_{V_{OS}}^{2} \cong 2V_{t}^{2} \left[ \frac{A_{Jn}^{2}}{A_{En}} + \frac{A_{Jp}^{2}}{A_{Ep}} \right]$$

where very approximately

$$A_{Jn} = A_{Jp} = 0.1 \mu$$

 $V_{CC}$ 

 $V_X$ 

 $V_{F}$ 

 $Q_3$ 

Q₁

 $V_1$ 

Q4

 $Q_2$ 

I<sub>T</sub>

 $V_2$ 

Example: Determine the  $3\sigma$  value of the offset voltage of a the bipolar input stage if  $A_{E1}=A_{E3}=10\mu^2$ 

$$\sigma_{V_{OS}}^{2} \approx 2V_{t}^{2} \left[ \frac{A_{Jn}^{2}}{A_{En}} + \frac{A_{Jp}^{2}}{A_{Ep}} \right]$$

$$\sigma_{V_{OS}} \approx \sqrt{2}V_{t} A_{J} \frac{\sqrt{2}}{\sqrt{A_{E}}}$$

$$\sigma_{V_{OS}} \approx 2 \cdot 25 \text{mV} \cdot 0.1 \mu \cdot \frac{1}{\sqrt{10\mu^{2}}} = 1.6 \text{mV}$$

$$3\sigma_{V_{OS}} \approx 4.7 \text{mV}$$

Note this value is much smaller than that for the MOS input structure !

Typical offset voltages:

MOS - 5mV to 50MV BJT - 0.5mV to 5mV

These can be scaled with extreme device dimensions

Often more practical to include offset-compensation circuitry

Offset voltage difficult to determine in come classes of comparators



Dynamic clocked comparator

When  $\phi_1$  is low,  $V_1$  and  $V_2$  are precharged to  $V_{DD}$  and no static power is dissipated When  $\phi_1$  is high, enters evaluate state and no static power is dissipated

#### Offset voltage difficult to determine in come classes of comparators

Very small, very fast, low power

But offset voltage can be large (100mV or more)





Decision is being made shortly after clock transition when devices are deep in weak inversion and signal levels are very small



Dynamic Comparators widely used because of low power dissipation

Often include one or more pre-amp stages before regeneration applied

Previous-code dependence and kickback both of concern in dynamic comparators

Noise may significantly affect performance and difficult to analyze and simulate because transient noise models in deep weak inversion are questionable

Still major opportunities to make significant improvement in dynamic comparators

### Dynamic Comparators



Relatively small number of dynamic comparators have been introduced

Significant difference in performance among those available

Analysis and performance assessment either analytically or via simulation not trivial

Opportunity to make significant advances in dynamic comparator design likely available

Analyses of static and dynamic random offset voltages in dynamic comparators
J He, S Zhan, D Chen, <u>RL Geiger</u> - IEEE Transactions on …, 2009 - ieeexplore.ieee.org
… Geiger, "Yield enhancement with optimal area allocation for ratio critical analog circuits," …
He, S. Zhan, D. Chen, and RL Geiger, "A simple and accurate method to predict offset voltage in …
☆ Save 55 Cite Cited by 244 Related articles All 15 versions Web of Science: 105

Additional details about offset voltage, statistical circuit analysis, and matching can be found in the draft document

"Statistical Characterization of Circuit Functions" by R.L. Geiger

### Summary of Offset Voltage Issues

- Random offset voltage is generally dominant and due to mismatch in device and model parameters
- MOS Devices have large V<sub>OS</sub> if area is small
- $\sigma$  decreases approximately with  $1/\sqrt{A}$
- Multiple fingers for MOS devices offer benefits for common centroid layouts but too many fingers will ultimately degrade offset because perimeter/area ration will increase (A<sub>W</sub> and A<sub>L</sub> will become of concern)
- Offset voltage of dynamic comparators is often large and analysis not straightforward
- Offset compensation often used when low offsets important

MOS: 
$$\sigma_{V_{OS}}^{2} \cong 2 \left[ \frac{A_{VTO}n}{W_{n}L_{n}} + \frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n}L_{p}^{2}} A_{VTOp}^{2} \right]$$
Bipolar: 
$$\sigma_{V_{OS}}^{2} \cong 2V_{t}^{2} \left[ \frac{A_{Jn}^{2}}{A_{En}} + \frac{A_{Jp}^{2}}{A_{En}} \right]$$



# Stay Safe and Stay Healthy !

# End of Lecture 11