## EE 505

## Lecture 11

- Formalization of Statistical Models
- Offset Voltages

String DAC Statistical Performance
$I L_{k}$ assumes a maximum variance at mid-code


Current Steering DAC Statistical Characterization
Binary Weighted

$$
\begin{aligned}
& \sigma_{I L_{\text {max }}} \cong \sigma_{I N L_{b=1,1, \ldots)}} \cong \frac{\sqrt{N}}{2} \sigma_{\frac{I_{R G}}{I_{\text {LSBX }}}}^{2}
\end{aligned}
$$

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

## Statistical Modeling of Current Sources

$\sigma_{\frac{\mathrm{DR}^{\prime}}{\mathrm{CO}_{\mathrm{ON}}}}=\sqrt{\sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}+\sigma_{\frac{C_{\text {OXX }}}{\mathrm{C}_{\text {OXN }}}}^{2}+4\left(\frac{V_{T H N}}{V_{G S}-V_{T H N}}\right)^{2} \sigma_{\frac{V_{T H R}}{V_{T H N}}}^{2}}$
or $\quad \sigma_{\frac{\mathrm{C}_{\mathrm{OR}}}{\mathrm{CON}^{\prime}}}=\sqrt{\sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}+\sigma_{\frac{C_{O X R}}{C_{\mathrm{OXN}}}}^{2}+\left(\frac{2}{V_{G S}-V_{T H N}}\right)^{2} \sigma_{V_{T H R}}^{2}}$
It will be assumed that (will discuss assumption later)

$$
\begin{aligned}
& \sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}=\frac{A_{\mu}^{2}}{W L} \\
& \sigma_{\frac{C_{\text {OXXR }}}{2}}^{C_{\mathrm{OXNO}}}=\frac{A_{\mathrm{CoX}}^{2}}{W L} \\
& \sigma_{V_{T H R}}^{2}=\frac{A_{V T 0}^{2}}{W L}
\end{aligned}
$$

$$
\text { where } \mathrm{A}_{\mu}, \mathrm{A}_{\text {Cox }}, \mathrm{A}_{\text {VT0 }} \text { are Pelgrom }
$$

process parameters

Define

$$
\sigma_{\frac{\mathrm{DOR}}{\mathrm{DON}}}=\frac{1}{\sqrt{W L}} \sqrt{A_{\mu}^{2}+A_{C O X}^{2}+\frac{4}{V_{E B}^{2}} A_{V T 0}^{2}}
$$

$$
A_{\beta}=\sqrt{A_{\mu}^{2}+A_{C o x}^{2}}
$$

Thus

$$
\sigma_{\frac{\mathrm{bg}}{\mathrm{bN}}}=\frac{1}{\sqrt{W L}} \sqrt{A_{\beta}^{2}+\frac{4}{V_{E B}^{2}} A_{V T 0}^{2}} \quad \text { Often only } \mathrm{A}_{\beta} \text { is available }
$$

## Review from Last Lecture

## Statistical Modeling of Current Sources

$$
\sigma_{\frac{\mathrm{CoR}}{\mathrm{OD}}}=\frac{1}{\sqrt{W L}} \sqrt{A_{\beta}^{2}+\frac{4}{V_{E B}^{2}} A_{V T 0}^{2}}
$$

Gate area: $\quad \mathrm{A}=\mathrm{WL}$

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

$$
\sigma_{\frac{\mathrm{I}_{\mathrm{DR}}}{\mathrm{DN}}} \cong \frac{2}{V_{\mathrm{EB}} \sqrt{W L}} A_{V T 0}
$$

Theorem: If the random part of two uncorrelated current sources $I_{1}$ and $I_{2}$ are identically distributed with normalized variance, $\sigma_{\text {th//s }}^{2}$ then the random variable $\Delta I=I_{2}-I_{1}$ has a variance given by the equation $\sigma_{\Delta / / \omega}^{2}=2 \sigma_{t / / \omega}^{2}$

Proof:

$$
\begin{aligned}
& \Delta I=I_{1}-I_{2} \\
& \frac{\Delta I}{I_{N}}=\frac{I_{1}}{I_{N}}-\frac{I_{2}}{I_{N}} \\
& \frac{\Delta I}{I_{N}}=\frac{I_{N}+I_{R 1}}{I_{N}}-\frac{I_{N}+I_{R 2}}{I_{N}}=\frac{I_{R 1}}{I_{N}}-\frac{I_{R 2}}{I_{N}} \\
& \sigma_{\frac{\Delta}{I_{N}}}^{2}=\sigma_{\frac{I_{R}}{2}}^{I_{N}}+\sigma_{\frac{I_{R}}{2}}^{2}=2 \sigma_{\frac{I_{R}}{2}}^{2}
\end{aligned}
$$

## Statistical Modeling of Circuits

The previous statistical analysis was somewhat tedious
Will try to formalize the process for obtaining two important statistics, the mean and variance, of a function of interest

Assume $Y$ is a function of $n$ uncorrelated random variables $X_{R 1}, \ldots x_{R n}$ where the mean and variance of $\mathrm{X}_{\mathrm{R} i}$ are "small"

$$
\mathrm{Y}=\mathrm{f}\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)
$$

$$
\mathrm{X}_{R}=\left[\begin{array}{l}
x_{R 1} \\
x_{R 2} \\
\ldots \\
x_{R n}
\end{array}\right]
$$

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

Recall if $\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)$ uncorrelated and $f=\sum_{i=1}^{m} a_{i} x_{R i}$ then $\sigma_{f}^{2}=\sum_{i=1}^{m} a_{i}^{2} \sigma_{X_{R i}}^{2}$
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 $\sigma$

## Statistical Modeling of Circuits

$$
\mathrm{Y}=\mathrm{f}\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right) \quad \mathrm{X}_{R}=\left[\begin{array}{l}
x_{R 1} \\
x_{R 2} \\
\ldots \\
x_{R n}
\end{array}\right]
$$

Assuming means are all $0, \quad \mathrm{Y}$ can be expressed in a Taylor's series expanded around mean as

$$
\mathrm{Y}=\left.f(X)\right|_{X_{R}=0}+\left.\sum_{j=1}^{n} \frac{\partial f}{\partial x_{R j}}\right|_{X_{R}=0} x_{R j}+\varepsilon\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)
$$

where $\varepsilon\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)$ is due to higher-order terms and is small

## Statistical Modeling of Circuits

$$
\left.\mathrm{Y} \simeq f(X)\right|_{X_{R}=0}+\left.\sum_{j=1}^{n} \frac{\partial f}{\partial x_{R j}}\right|_{X_{R}=0} x_{R j}
$$

Note power series expansion linearized Y in the variables $\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)$

$$
\frac{\mathrm{Y}}{\mathrm{Y}_{\mathrm{N}}}=\frac{\left.f(X)\right|_{X_{R}=0}}{\mathrm{Y}_{\mathrm{N}}}+\left.\sum_{j=1}^{n} \frac{1}{\mathrm{Y}_{\mathrm{N}}} \frac{\partial f}{\partial x_{R j}}\right|_{X_{R}=0} x_{R j}
$$

From Theorem:

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

Define:

$$
\begin{gathered}
\hat{S}_{x k j}^{f}=\left.\frac{1}{\mathrm{Y}_{\mathrm{N}}} \frac{\partial f}{\partial x_{R j}}\right|_{X_{R}=0} \\
\sigma_{\frac{Y}{Y_{\mathrm{N}}}}^{2}=\sum_{j=1}^{n}\left(\left[\hat{S}_{x f j}^{f}\right]^{2} \sigma_{x_{k j}}^{2}\right)
\end{gathered}
$$

## Statistical Modeling of Circuits

But

$$
\sigma_{\frac{\mathrm{Y}}{Y_{\mathrm{N}}}}^{2}=\sum_{j=1}^{n}\left(\left[\hat{S}_{x R j}^{f}\right]^{2} \sigma_{x_{R j}}^{2}\right)=\sum_{j=1}^{n}\left(\left[\hat{S}_{x R j}^{f}\right]^{2}\left(\frac{x_{j N}}{x_{j N}}\right)^{2} \sigma_{x_{R j}}^{2}\right)=\sum_{j=1}^{n}\left(\left[x_{j N} \hat{S}_{x R j}^{f}\right]^{2} \sigma_{\frac{x_{R j}}{x_{j N}}}^{2}\right)
$$

Alternatively, from the more standard definition of sensitivity $S_{x R_{j}}^{f}$ :

$$
S_{x R_{j}}^{f}=\left.\frac{x_{j N}}{\mathrm{Y}_{\mathrm{N}}} \frac{\partial f}{\partial x_{R j}}\right|_{X_{R}=0} \quad \square \quad S_{x R_{j}}^{f}=x_{j N} \hat{S}_{x R_{j}}^{f}
$$

we thus obtain

$$
\sigma_{\frac{\mathrm{Y}}{\mathrm{Y}_{\mathrm{N}}}}^{2}=\sum_{j=1}^{n}\left(\left[S_{x R j}^{f}\right]^{2} \sigma_{\frac{x_{R j}}{2}}^{x_{N i}}\right)
$$

## Statistical Modeling of Circuits

$$
\mathrm{Y}=\mathrm{f}\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right)
$$

Y is any function of interest

$$
\left(x_{R 1}, x_{R 2}, \ldots x_{R n}\right) \text { Random part of process parameters }
$$



Determined by Process

- 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



## Consider first the mobility

Claim:

$$
\sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}=\frac{\mathrm{A}_{\mu}^{2}}{\mathrm{~A}}
$$

where $A_{\mu}$ is the Pelgrom matching parameter and $A$ is the gate area
Argument:
Assume (define)

$$
\mu_{\mathrm{eq}}=\frac{\int_{A} \mu(x, y) d x d y}{\mathrm{~A}}
$$

Let $S_{k}$ be a square of area $A_{k}$ in the channel


$$
\mu_{\mathrm{eq}}=\frac{\sum_{i=1}^{N} \int_{A_{k i}} \mu(x, y) d x d y}{\mathrm{~A}}
$$

where the channel has been partitioned into N disjoint regions each of area $\mathrm{A}_{\mathrm{ki}}$
For convenience, assume $A_{k i}=A_{k j}=A_{k}$ for all $i, j$

Claim: $\quad \sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}=\frac{\mathrm{A}_{\mu}^{2}}{\mathrm{~A}}$


Argument continued:

$$
\mu_{\mathrm{eq}} \simeq \frac{\sum_{i=1}^{N} \int_{A_{k i}} \mu(x, y) d x d y}{\mathrm{~A}}
$$

Assume the random variables $\int_{A_{k i}} \mu(x, y) d x d y$ are uncorrelated and identically distributed with variance $\sigma_{\mu_{-} A_{k}}^{2}$
It thus follows that $\sigma_{\mu_{e q}}^{2} \simeq \frac{1}{\mathrm{~A}^{2}} \mathrm{~N} \sigma_{\mu_{-} A_{k}}^{2}$
But nominal $A_{k}$ is $A_{k N}=A / N \Longrightarrow N=A / A_{k N} \Longrightarrow \sigma_{\mu_{e q}}^{2} \simeq \frac{1}{\mathrm{~A}^{2}} \frac{\mathrm{~A}}{\mathrm{~A}_{\mathrm{kN}}} \sigma_{\mu_{-} \mathrm{A}_{k}}^{2}=\frac{1}{\mathrm{~A}} \frac{\sigma_{\mu_{-} \mathrm{A}_{k}}^{2}}{\mathrm{~A}_{\mathrm{kN}}}$

Define $\quad A_{\mu}=\frac{\sigma_{\mu_{-} A_{k}}}{\mu_{N} \sqrt{A_{k N}}}$

$$
\sigma_{\frac{\mu_{e q}}{\mu_{N}}}^{2}=\frac{1}{\mathrm{~A}} \frac{\sigma_{\mu_{-} \mathrm{A}_{\mathrm{k}}}^{2}}{\mu_{N}^{2} \mathrm{~A}_{\mathrm{kN}}}=\frac{\mathrm{A}_{\mu}^{2}}{\mathrm{~A}}
$$

## Concept can be extended so now have:

$$
\begin{aligned}
& \sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}=\frac{A_{\mu}^{2}}{W L} \\
& \sigma_{\frac{C_{\text {OXX }}}{C_{\text {OXN }}}}^{2}=\frac{A_{\text {Cox }}^{2}}{W L} \\
& \sigma_{V_{\text {THR }}}^{2}=\frac{A_{V T 0}^{2}}{W L} \\
& \sigma_{\frac{R}{R_{N}}}^{2}=\frac{A_{R \square}^{2}}{W L}
\end{aligned}
$$

where $A_{\mu}, A_{\text {Cox }}, A_{V T 0}, A_{R \square}$ 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. $R_{R}, C_{R}, V_{O S R}, I_{R}, \ldots$. )

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 $X \sim U[0,1]$, then $y=F^{-1}(x)$ has a pdf of $f(y)$.

Corollary: If $h$ is a rv with $F(h)=\frac{1}{\sqrt{2 \pi}} \int_{-\infty}^{h} e^{-\frac{x^{2}}{2}} d x$ then $\mathrm{y}=\mathrm{F}^{-1}(\mathrm{~h})$ is $\mathrm{N}[0,1]$

CDF showing random variable mapping of $x_{1}$ from $U(0,1)$


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

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

In Excel:
NORM.S.INV(h) $=\mathrm{F}^{-1}(\mathrm{~h})$ where

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

$\operatorname{NORM} . \operatorname{DIST}(h, \mu, \sigma, \operatorname{TRUE})=f(\mathrm{~h})$ where

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

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

## Some useful relationships:

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

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

$$
\mathrm{F}_{N}(\mathrm{x})=\frac{1}{2}\left(1+\operatorname{ERF}\left(\frac{\mathrm{x}}{\sqrt{2}}\right)\right)
$$

| Excel | Older Excel |  |
| :--- | :--- | :--- |
| @NORM.DIST |  | $\mathrm{f}(x)$ |
| @NORM.S.DIST |  | $\mathrm{f}_{N}(x)$ |
| @NORM.INV | @NORMINV | $F^{-1}(x)$ |
| @NORM.S.INV | @NORMSINV | $F_{N}^{-1}(x)$ |
|  | @NORMDIST | $F_{N}(x)$ |
|  | @NORMINV | $F(x)$ |

where f: PDF $\mathrm{F}: C D F$

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 $\left|\left|N L_{\text {KMAX }}\right|<\frac{1}{2} L S B\right.$
Solution:
Want:

$$
P=\int_{x=-\frac{1}{2}}^{x=+\frac{1}{2}} f\left(I N L_{k M A X}\right) d x
$$

Let $\quad X_{N}=\frac{x}{\sigma_{N L}} \quad \mu_{I L_{\text {LMAX }}}=0$

$$
P=2 F_{N}\left(X_{N}\right)-1 \Longrightarrow X_{N}=F_{N}^{-1}\left(\frac{P+1}{2}\right)
$$

where $F_{N}\left(X_{N}\right)$ is the CDF of a $N(0,1) r v$

$$
X_{N}=\frac{1}{\sigma_{\frac{R_{R}}{R_{N}}} \sqrt{N}} \quad \text { recall } \quad \sigma_{\frac{R_{R}}{R_{N}}}=\frac{\mathrm{A}_{\mathrm{R}}}{\sqrt{\mathrm{WL}}}
$$

thus $\quad X_{N}=\frac{\sqrt{W L}}{A_{R} \sqrt{N}} \quad \square \quad \sqrt{W L}=A_{R} \sqrt{N} X_{N}$
thus, we obtain

$$
\sqrt{W L}=A_{R} \sqrt{N} \cdot F_{N}^{-1}\left(\frac{P+1}{2}\right)
$$

Since there are $\mathrm{N}=2^{\mathrm{n}}$ resistors, total area becomes

$$
A_{T O T}=2^{n} \sqrt{W L}=2^{n} A_{R} \sqrt{N} \bullet F_{N}^{-1}\left(\frac{P+1}{2}\right)=2^{n} A_{R} 2^{\frac{n}{2}} \bullet F_{N}^{-1}\left(\frac{P+1}{2}\right)
$$

or equivalently

$$
A_{T O T}=\sqrt{2^{3 n}} A_{R} \bullet F_{N}^{-1}\left(\frac{P+1}{2}\right)
$$

## Offset Voltages

## 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 desired value

With a good design, a designer will have $\mathrm{V}_{\text {OUT }}$ 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

## 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 $\mathrm{V}_{\mathrm{id}}=0$ and $\mathrm{V}_{\mathrm{ic}}$ is the quiescent commonmode input voltage.

## $\mathrm{V}_{\text {OUTOFF }}=\mathrm{V}_{\text {OUT }}-\mathrm{V}_{\text {OUTDES }}$

Note: $\mathrm{V}_{\text {OUTOFF }}$ is dependent upon $\mathrm{V}_{\text {ICQ }}$ although this dependence is usually quite weak and often not specified

## Offset Voltage



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

Note: $\mathrm{V}_{\text {OFF }}$ is usually related to the output offset voltage by the expression

$$
V_{\text {OFF }}=\frac{V_{\text {OUTOFF }}}{A_{D}}
$$

Note: $\mathrm{V}_{\text {OFF }}$ is dependent upon $\mathrm{V}_{\text {ICQ }}$ although this dependence is usually quite weak and often not specified

## Offset Voltage

## 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

## Offset Voltage



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

## Offset Voltage

Effects of Offset Voltage - an example


Desired I/O relationship


## Offset Voltage

## Effects of Offset Voltage - an example

Desired I/O relationship



Actual I/O relationship due to offset



## Offset Voltage



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 $\mathrm{V}_{\mathrm{os}}$ small in the first place though penalty (e.g. cost) for making it sufficiently small without correction is often unacceptable

## Dynamic Offset Compensation



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

## Offset Voltage Distribution



Typical histogram of offset voltage (binned) in shipped parts
Low-offset parts sold at a premium
Extreme offset parts have been sifted at test

## Source of Random Offset Voltages

Consider as an example:


Ideally $R_{1}=R_{2}=R_{N}, M_{1}$ and $M_{2}$ are matched

$$
V_{\mathrm{OUT}}=\mathrm{V}_{\mathrm{DD}}-\left(\frac{\mathrm{I}_{\mathrm{T}}}{2}\right) \mathrm{R}_{\mathrm{N}}
$$

Assume this is the desired output voltage (note not assuming $\mathrm{V}_{\text {OUt-DES }}=0 \mathrm{~V}$ )

## Source of Random Offset Voltages

Consider as an example:

If everything ideal except $R_{1}$ and $R_{2}$


$$
R_{1}=R_{N}+R_{R 1} \quad R_{2}=R_{N}+R_{R 2}
$$

Thus at the design stage, $\mathrm{V}_{\text {Out }}$ is also a random variable

$$
\begin{gathered}
V_{\text {OUT }}=V_{D D}-\left(\frac{I_{T}}{2}\right)\left[R_{N}+R_{R 2}\right] \\
V_{\text {OUT-R }}=-\left(\frac{I_{T}}{2}\right) R_{R 2}
\end{gathered}
$$

## Source of Random Offset Voltages

Consider as an example:


$$
A_{V N}=-\frac{g_{m}}{2} R_{N}
$$

## Source of Random Offset Voltages

Determine the offset voltage - i.e. value of $\mathrm{V}_{\mathrm{x}}$ needed to obtain desired output


Setting $\mathrm{V}_{\text {OUT }}=\mathrm{V}_{\text {OUt-DES }}$ and solving for $\mathrm{V}_{\mathrm{X}}$, we obtain

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

## Source of Random Offset Voltages

Determine the offset voltage - i.e. value of $\mathrm{V}_{\mathrm{x}}$ needed to obtain desired output

$$
\begin{aligned}
& A_{V}=-\frac{g_{m}}{2} R \\
& V_{X}=V_{\text {OFF }}=\frac{-1}{A_{V}}\left(\frac{I_{T}}{2}\right) R_{R 2} \\
& V_{X}=\frac{2}{g_{m} R_{N}}\left(\frac{I_{T}}{2}\right) R_{R 2}=\left(\frac{I_{T}}{g_{m}}\right) \frac{R_{R 2}}{R_{N}}=\left(\frac{I_{T}}{I_{T} / V_{E B}}\right) \frac{R_{R 2}}{R_{N}}=V_{E B} \frac{R_{R 2}}{R_{N}} \\
& V_{O S}=V_{E B} \frac{R_{R 2}}{R_{N}}
\end{aligned}
$$

## Source of Random Offset Voltages

Determine the offset voltage - i.e. value of $\mathrm{V}_{\mathrm{x}}$ needed to obtain desired output


$$
\begin{aligned}
V_{O S} & =V_{E B} \frac{R_{R 2}}{R_{N}} \\
\sigma_{V_{O S}} & =V_{E B} \frac{\sigma_{R_{R 2}}}{R_{N}}
\end{aligned}
$$

$\begin{aligned} & \text { If resistors are integrated } \\ & \text { and } A \text { is the resistor area }\end{aligned} \frac{\sigma_{R_{R 2}}}{R_{N}}=\frac{A_{R}}{\sqrt{A}}$ where $A_{R}$ is the Pelgrom parameter

Thus

$$
\sigma_{\mathrm{V}_{\mathrm{OS}}}=\mathrm{V}_{\mathrm{EB}} \frac{\mathrm{~A}_{\mathrm{R}}}{\sqrt{\mathrm{~A}}}
$$

## Source of Random Offset Voltages

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

(a)

(b)

## Random Offset Voltages



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

## Random Offset Voltages

The random offset is due to missmatches in the four transistors, dominantly missmatches in the parameters $\left\{\mathrm{V}_{\mathrm{T}}, \mu, \mathrm{C}_{\mathrm{OX}}, \mathrm{W}\right.$ and L$\}$

The relative missmatch effects become more pronounced as devices become smaller

$$
\begin{aligned}
& \mathrm{V}_{\mathrm{Ti}}=\mathrm{V}_{\mathrm{TN}}+\mathrm{V}_{\mathrm{TRi}} \\
& \mathrm{C}_{\mathrm{OXi}}=\mathrm{C}_{\mathrm{OXN}}+\mathrm{C}_{\mathrm{OXRi}} \\
& \mu_{\mathrm{i}}=\mu_{\mathrm{N}}+\mu_{\mathrm{Ri}} \\
& \mathrm{~W}_{\mathrm{i}}=\mathrm{W}_{\mathrm{N}}+\mathrm{W}_{\mathrm{Ri}} \\
& \mathrm{~L}_{\mathrm{i}}=\mathrm{L}_{\mathrm{N}}+\mathrm{L}_{\mathrm{Ri}}
\end{aligned}
$$



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

## Random Offset Voltages

$$
\begin{aligned}
& \mathrm{V}_{\mathrm{Ti}}=\mathrm{V}_{\mathrm{TN}}+\mathrm{V}_{\mathrm{TRi}} \\
& \mathrm{C}_{\mathrm{OXi}}=\mathrm{C}_{\mathrm{OXN}}+\mathrm{C}_{\mathrm{OXRi}} \\
& \mu_{\mathrm{i}}=\mu_{\mathrm{N}}+\mu_{\mathrm{Ri}} \\
& \mathrm{~W}_{\mathrm{i}}=\mathrm{W}_{\mathrm{N}}+\mathrm{W}_{\mathrm{Ri}} \\
& \mathrm{~L}_{\mathrm{i}}=\mathrm{L}_{\mathrm{N}}+\mathrm{L}_{\mathrm{Ri}}
\end{aligned}
$$



For each device, the device model is often expressed as

$$
\mathrm{I}_{\mathrm{Di}}=\frac{\left(\mu_{\mathrm{N}}+\mu_{\mathrm{Ri}}\right)\left(\mathrm{C}_{\mathrm{OXN}}+\mathrm{C}_{\mathrm{OXRi}}\right)\left(\mathrm{W}_{\mathrm{N}}+\mathrm{W}_{\mathrm{Ri}}\right)}{2\left(\mathrm{~L}_{\mathrm{N}}+\mathrm{L}_{\mathrm{Ri}}\right)}\left(\mathrm{V}_{\mathrm{GSi}}-\left(\mathrm{V}_{\mathrm{TN}}+\mathrm{V}_{\mathrm{TRi}}\right)\right)^{2}\left(1+\left(\lambda_{\mathrm{N}}+\lambda_{\mathrm{Ri}}\right)\left[\mathrm{V}_{\mathrm{DS}}\right]\right)
$$

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

## Dffort Woiteres



Assume currents at output node must satisfy relation $\quad I_{2}=I_{4}$
Strategy:

1) Obtain expression for $V_{\text {OFF }}$ (referred to input) that forces $I_{2}=I_{4}$
2) Linearize expression in terms of design variables and decorrelate
3) Obtain $\sigma_{V o s}$

Analysis of Offset Voltage (Neglect $R_{L}$ )

$$
\begin{aligned}
& \mathrm{I}_{\mathrm{D} 1}=\frac{\mu_{\mathrm{n} 1} \mathrm{C}_{\mathrm{OX} 1} \mathrm{~W}_{1}}{2 \mathrm{~L}_{1}}\left(\mathrm{~V}_{\text {OFF }}+\mathrm{V}_{\mathrm{INC}}-\mathrm{V}_{\mathrm{S}}-\mathrm{V}_{\mathrm{TH} 1}\right)^{2} \\
& \mathrm{I}_{\mathrm{D} 2}=\frac{\mu_{\mathrm{n} 2} \mathrm{C}_{\mathrm{OX} 2} \mathrm{~W}_{2}}{2 \mathrm{~L}_{2}}\left(\mathrm{~V}_{\mathrm{INC}}-\mathrm{V}_{\mathrm{S}}-\mathrm{V}_{\mathrm{TH} 2}\right)^{2} \\
& \mathrm{I}_{\mathrm{D} 3}=\frac{\mu_{\mathrm{p} 3} \mathrm{C}_{\mathrm{OX} 3} \mathrm{~W}_{3}}{2 \mathrm{~L}_{3}}\left(\mathrm{~V}_{\mathrm{X}}-\mathrm{V}_{\mathrm{DD}}-\mathrm{V}_{\mathrm{TH} 3}\right)^{2} \\
& \mathrm{I}_{\mathrm{D} 4}=\frac{\mu_{\mathrm{p} 4} \mathrm{C}_{\mathrm{OX} 4} \mathrm{~W}_{4}}{2 \mathrm{~L}_{4}}\left(\mathrm{~V}_{\mathrm{X}}-\mathrm{V}_{\mathrm{DD}}-\mathrm{V}_{\mathrm{TH} 4}\right)^{2}
\end{aligned}
$$



Since $\sqrt{I_{D 1}}=\sqrt{I_{D 3}}$

$$
\mathrm{V}_{O F F}+\mathrm{V}_{\mathrm{INC}}-\mathrm{V}_{\mathrm{S}}-\mathrm{V}_{\mathrm{TH} 1}=\sqrt{\frac{\mu_{\mathrm{p} 3} \mathrm{C}_{\mathrm{OX} 3} \mathrm{~W}_{3} \mathrm{~L}_{1}}{\mu_{\mathrm{n} 1} \mathrm{C}_{\mathrm{OX} 1} \mathrm{~W}_{1} \mathrm{~L}_{3}}}\left(\mathrm{~V}_{\mathrm{X}}-\mathrm{V}_{\mathrm{DD}}-\mathrm{V}_{\mathrm{TH} 3}\right)
$$

Since $\sqrt{I_{D 2}}=\sqrt{I_{D 4}}$

$$
\mathrm{V}_{\mathrm{INC}}-\mathrm{V}_{\mathrm{S}}-\mathrm{V}_{\mathrm{TH} 2}=\sqrt{\frac{\mu_{\mathrm{p} 4} \mathrm{C}_{\mathrm{OX} 4} \mathrm{~W}_{4} \mathrm{~L}_{2}}{\mu_{\mathrm{n} 2} \mathrm{C}_{\mathrm{OX} 2} \mathrm{~W}_{2} 2 \mathrm{~L}_{4}}}\left(\mathrm{~V}_{\mathrm{X}}-\mathrm{V}_{\mathrm{DD}}-\mathrm{V}_{\mathrm{TH} 4}\right)
$$

## Analysis of Offset Voltage

Define:

$$
a=\sqrt{\frac{L_{1} \mu_{p 3} C_{o X 3} W_{3}}{L_{3} \mu_{n 1} C_{o X 1} W_{1}}} \quad b=\sqrt{\frac{L_{2} \mu_{p 4} C_{o x 4} W_{4}}{L_{4} \mu_{n 2} C_{O X 2} W_{2}}}
$$

Substituting for a and b , it follows on eliminating $\mathrm{V}_{\mathrm{S}}$ that

$$
V_{\text {OFF }}=V_{T H 1}-V_{T H 2}+(a-b)\left(\mathrm{V}_{\mathrm{x}}-\mathrm{V}_{\mathrm{DD}}\right)+b \mathrm{~V}_{\mathrm{TH} 4}-a \mathrm{~V}_{\mathrm{TH} 3}
$$

Assume

$$
\begin{array}{ll}
V_{X}=V_{X N}-V_{X R} & \\
a=a_{N}+a_{R} & \\
b=b_{N}+b_{R} & \\
V_{T n i}=V_{T n N}+V_{T n R i} & i=1,2 \\
V_{T p i}=V_{T p N}+V_{T p R i} & i=3,4
\end{array}
$$



Observe $\mathrm{a}_{\mathrm{N}}=\mathrm{b}_{\mathrm{N}}$ and $\mathrm{V}_{\mathrm{XN}}-\mathrm{V}_{\mathrm{DD}}-\mathrm{V}_{\mathrm{TPN}}=\mathrm{V}_{\mathrm{EB} 3}$
Since the random part of $\mathrm{V}_{\mathrm{x}}$ multiplies only a-b which is small, it follows that

$$
\begin{aligned}
& V_{O F F}=V_{T H 1}-V_{T H 2}+(a-b)\left(\mathrm{V}_{\text {EB3N }}\right)+b \mathrm{~V}_{\text {TH4 }}-a \mathrm{~V}_{\text {TH }} \\
& V_{\text {OFF }}=V_{\text {THR } 1}-V_{\text {THR } 2}+\left(a_{R}-b_{R}\right) V_{\text {EB3N }}+a_{N}\left(V_{\text {THR } 4}-V_{\text {THR } 3}\right) \\
& \sigma_{V_{\text {ofF }}}^{2}=2 \sigma_{V_{T \text { NR }}}^{2}+a_{N}^{2} 2 \sigma_{V_{T \text { TRR }}}^{2}+V_{E B 3 N}^{2} \sigma_{V_{\text {aR }-b_{R}}}^{2}
\end{aligned}
$$

Will now obtain $a_{R}$ and $b_{R}$

## Analysis of Offset Voltage

$$
V_{\text {OFF }}=V_{\text {TRR2 } 2}-V_{\text {TRR2 } 2}+\left(b_{R}-a_{R}\right) V_{\text {EB3 }}+a_{N}\left(V_{\text {TPR3 }}-V_{\text {TPR } 4}\right)
$$

$$
a=\sqrt{\frac{\left(L_{N 1}+L_{R 1}\right)\left(\mu_{N 03}+\mu_{R 3}\right)\left(C_{O X N 3}+C_{O X R 3}\right)\left(W_{N 3}+W_{R 3}\right)}{\left(L_{N 3}+L_{R 3}\right)\left(\mu_{N 1}+\mu_{R 1}\right)\left(C_{O X N 1}+C_{O X R 1}\right)\left(W_{N 1}+W_{R 1}\right)}} v_{\text {off }} \frac{\Gamma}{V_{\text {WMc }}}
$$



Recall for x small, $\sqrt{1+x} \cong 1+\frac{x}{2} \quad \frac{1}{1+x} \cong 1-\mathrm{x}$

$$
a=\sqrt{\frac{\left(L_{N 1} \mu_{N \rho 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N 1} W_{N 1}\right)}}\left(1+\frac{1}{2}\left[\frac{L_{R 1}}{L_{N 1}}-\frac{L_{R 3}}{L_{N 3}}+\frac{\mu_{R 3}}{\mu_{N \rho 3}}-\frac{\mu_{R 1}}{\mu_{N 11}}+\frac{C_{O X R 3}}{C_{O X N 3}}-\frac{C_{O X R 1}}{C_{O X N 1}}+\frac{W_{R 3}}{W_{N 3}}-\frac{W_{R 3}}{W_{N 3}}\right]\right)
$$

Thus

$$
\begin{aligned}
& a_{R}=\sqrt{\frac{\left(L_{11} \mu_{N 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N 1} W_{N 1}\right)}} \frac{1}{2}\left[\frac{L_{R 1}}{L_{N 1}}-\frac{L_{R 3}}{L_{N 3}}+\frac{\mu_{R 3}}{\mu_{N 3}}-\frac{\mu_{R 1}}{\mu_{N 1}}+\frac{C_{O X R 3}}{C_{O X N 3}}-\frac{C_{O X R 1}}{C_{O \times N 1}}+\frac{W_{R 3}}{W_{N 3}}-\frac{W_{R 3}}{W_{N 3}}\right] \\
& a_{N}=\sqrt{\frac{\left(L_{N 1} \mu_{N 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N 1} W_{N 1}\right)}}
\end{aligned}
$$

Likewise

$$
b_{R}=\sqrt{\frac{\left(L_{N 1} \mu_{N \rho} W_{N 3}\right)}{\left(L_{N 3} \mu_{N 11} W_{N 1}\right)}} \frac{1}{2}\left[\frac{L_{R 2}}{L_{N 2}}-\frac{L_{R 4}}{L_{N 4}}+\frac{\mu_{R 4}}{\mu_{N 4}}-\frac{\mu_{R 2}}{\mu_{N n 2}}+\frac{C_{O X R 4}}{C_{O X N 4}}-\frac{C_{O X R 2}}{C_{O X N 2}}+\frac{W_{R 4}}{W_{N 4}}-\frac{W_{R 2}}{W_{N 2}}\right]
$$

## Analysis of Offset Voltage

$$
\begin{aligned}
& a_{R}-b_{R}=\sqrt{\frac{\left(L_{N 1} \mu_{N \rho 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N 11} W_{N 1}\right)}} \frac{1}{2}\left[\begin{array}{l}
\frac{L_{R 1}}{L_{N 1}}-\frac{L_{R 2}}{L_{N 2}}+\frac{L_{R 4}}{L_{N 4}}-\frac{L_{R 3}}{L_{N 3}}+\frac{\mu_{R 3}}{\mu_{N p 3}}-\frac{\mu_{R 4}}{\mu_{N p 4}}+\frac{\mu_{R 2}}{\mu_{N n 2}}-\frac{\mu_{R 1}}{\mu_{N n 1}} \\
+\frac{C_{O X R 3}}{C_{O X N 3}}-\frac{C_{O X R 4}}{C_{O X N 4}}+\frac{C_{O X R 2}}{C_{O X N 2}}-\frac{C_{O X R 1}}{C_{O X N 1}}+\frac{W_{R 3}}{W_{N 3}}-\frac{W_{R 4}}{W_{N 4}}+\frac{W_{R 2}}{W_{N 2}}-\frac{W_{R 3}}{W_{N 3}}
\end{array}\right] \\
& \sigma_{a_{R}-b_{R}}^{2}=\frac{\left(L_{N 1} \mu_{N p 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N n 1} W_{N 1}\right)} \frac{1}{2}\left[\sigma_{\frac{L_{R 1}}{L_{N 1}}}^{2}+\sigma_{\frac{L_{R 3}}{L_{N 3}}}^{2}+\sigma_{\frac{\mu_{R 3}}{\mu_{N p 3}}}^{2}+\sigma_{\frac{\mu_{R 2}}{\mu_{N n 2}}}^{2}+\sigma_{\frac{C_{O X R 3}}{C_{O X N 3}}}^{2}+\sigma_{\frac{C_{O X R 1}}{C_{O X N 1}}}^{2}+\sigma_{\frac{W_{R 3}}{W_{N 3}}}^{2}+\sigma_{\frac{W_{R 1}}{W_{N 1}}}^{2}\right]
\end{aligned}
$$

Thus

$$
\begin{aligned}
\sigma_{V_{\text {OFF }}}^{2} & =2 \sigma_{V_{T n R 2}}^{2}+2 \frac{L_{N 1} \mu_{N p 3} W_{N 3}}{L_{N 3} \mu_{N n 1} W_{N 1}} \sigma_{V_{T p R 3}}^{2} \\
& +V_{E B 3}^{2} \frac{\left(L_{N 1} \mu_{N p 3} W_{N 3}\right)}{\left(L_{N 3} \mu_{N n 1} W_{N 1}\right)} \frac{1}{2}\left[\sigma_{\frac{L_{R 1}}{L_{N 1}}}^{2}+\sigma_{\frac{L_{R 3}}{L_{N 3}}}^{2}+\sigma_{\frac{\mu_{R 3}}{\mu_{N p 3}}}^{2}+\sigma_{\frac{\mu_{R 2}}{\mu_{N n 2}}}^{2}+\sigma_{\frac{C_{O X R 3}}{C_{O X N 3}}}^{2}+\sigma_{\frac{C_{O X R 1}}{C_{O X N 1}}}^{2}+\sigma_{\frac{W_{R 3}}{W_{N 3}}}^{2}+\sigma_{\frac{W_{R 1}}{W_{N 1}}}^{2}\right]
\end{aligned}
$$

## Analysis of Offset Voltage

but

$$
\sigma_{V_{T}}^{2}=\frac{A_{V T 0}}{W L} \quad \sigma_{\frac{\mu_{R}}{\mu_{N}}}^{2}=\frac{A_{\mu}^{2}}{W L} \quad \sigma_{\frac{C_{\text {oxe }}}{C_{\text {oxx }}}}^{2}=\frac{A_{C \text { Cox }}^{2}}{W L} \quad \sigma_{\frac{L_{R}}{L_{N}}}^{2}=\frac{2 A_{L}^{2}}{W L^{2}} \quad \sigma_{\frac{W_{R}}{W_{N}}}^{2}=\frac{2 A_{W}^{2}}{W^{2} L}
$$

So the offset variance can be expressed as


$$
\begin{aligned}
\sigma_{V \text { ofF }}^{2} & =2 \frac{A_{V T n 0}^{2}}{W_{1} L_{1}}+2 \frac{\mu_{p} L_{1}}{\mu_{n} W_{1}} \frac{A_{V T p 0}^{2}}{L_{3}^{2}} \\
& +V_{E B 3}^{2} \frac{\mu_{p} L_{1} W_{3}}{\mu_{n} L_{3} W_{1}} \frac{1}{2}\left[\frac{A_{\mu_{n}}^{2}}{W_{3} L_{3}}+\frac{A_{\mu_{p}}^{2}}{W_{1} L_{1}}+A_{C o x}^{2}\left(\frac{1}{W_{3} L_{3}}+\frac{1}{W_{1} L_{1}}\right)+A_{w}^{2}\left(\frac{2}{W_{3}^{2} L_{3}}+\frac{2}{W_{1}^{2} L_{1}}\right)+A_{L}^{2}\left(\frac{2}{W_{1} L_{1}^{2}}+\frac{2}{W_{3} L_{3}^{2}}\right)\right]
\end{aligned}
$$

Often this can be approximated by

$$
\sigma_{V_{\text {OFF }}}^{2}=2 \frac{A_{V T T 0}^{2}}{W_{1} L_{1}}+2 \frac{\mu_{\rho} L_{1}}{\mu_{n} W_{1}} \frac{A_{V T \rho 0}^{2}}{L_{3}^{2}}+V_{E B 3}^{2} \frac{\mu_{p} L_{1} W_{3}}{\mu_{n} L_{3} W_{1}} \frac{1}{2}\left[\frac{A_{\mu_{n}}^{2}}{W_{3} L_{3}}+\frac{A_{\mu_{\rho}}^{2}}{W_{1} L_{1}}+A_{C o x}^{2}\left(\frac{1}{W_{3} L_{3}}+\frac{1}{W_{1} L_{1}}\right)\right]
$$

Or even approximated by

$$
\sigma_{V_{\text {ofF }}}^{2}=2 \frac{A_{V T n 0}^{2}}{W_{1} L_{1}}+2 \frac{\mu_{\rho} L_{1}}{\mu_{n} W_{1}} \frac{A_{V T p 0}^{2}}{L_{3}^{2}}
$$

## Random Offset Voltages

Since $\mathrm{V}_{\text {EBn }}$ and $\mathrm{V}_{\text {EBp }}$ are related, this is often expressed in simpler form as:

$$
\left.\sigma_{V_{0 S}}^{2}=2\left[\begin{array}{l}
\frac{A_{V T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{V T O p}^{2}+\frac{V_{E B n}^{2}}{4}\left(\begin{array}{l}
\frac{1}{W_{n} L_{n}} A_{\mu_{n}}^{2}+\frac{1}{W_{p} L_{p}} A_{\mu_{p}}^{2}+A_{\operatorname{COX}}^{2}\left[\frac{1}{W_{n} L_{n}}+\frac{1}{W_{p} L_{p}}\right.
\end{array}\right] \\
+2 A_{L}^{2}\left[\frac{1}{W_{n} L_{n}^{2}}+\frac{1}{W_{p} L_{p}^{2}}\right]+A_{w}^{2}\left[\frac{1}{L_{n} W_{n}^{2}}+\frac{1}{L_{p} W_{p}^{2}}\right]
\end{array}\right)\right]
$$

where the terms $A_{V T 0}, A_{\mu}, A_{C O X}, A_{L}$, and $A_{W}$ are process parameters $V_{D D}$

$$
\begin{aligned}
& A_{V T 0} \simeq \begin{cases}21 \mathrm{~m} V \cdot \mu & (n-c h) \\
25 \mathrm{~m} V \cdot \mu & (\mathrm{p}-\mathrm{ch})\end{cases} \\
& \sqrt{\mathrm{A}_{\mu}^{2}+\mathrm{A}_{\mathrm{C}_{\mathrm{OX}}}^{2}} \simeq \begin{cases}.016 \mu & (\mathrm{n}-\mathrm{ch}) \\
.023 \mu & (\mathrm{p}-\mathrm{ch})\end{cases} \\
& A_{L}=A_{W} \simeq 0.017 \mu^{3 / 2}
\end{aligned}
$$

Usually the $\mathrm{A}_{\text {VT0 }}$ terms are dominant, thus the variance simplifies to

$$
\sigma_{V_{O S}}^{2} \cong 2\left[\frac{A_{V T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{V T O p}^{2}\right]
$$



## Random Offset Voltages

Correspondingly:
which again simplifies to

$$
\sigma_{V_{o s}}^{2} \cong 2\left[\frac{A_{V T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{V T O p}^{2}\right]
$$

Note these offset voltage expressions are identical!


## Random Offset Voltages

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

$$
\text { Assume } \mathrm{L}_{\text {MIN }}=\mathrm{W}_{\text {MIN }}=0.5 \mathrm{u}, \mathrm{~A}_{\text {VTOn }}=0.021 \mathrm{~V} \text { and } \mathrm{A}_{\text {VTOP }}=0.025 \mathrm{~V}
$$

$$
\begin{gathered}
\sigma_{V_{O S}}^{2} \cong 2\left[\frac{A_{V T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{V T O p}^{2}\right] \\
\sigma_{V_{O S}}^{2} \cong \frac{2}{W_{n} L_{n}}\left[A_{V T O n}^{2}+\frac{\mu_{p}}{\mu_{n}} A_{V T O p}^{2}\right]
\end{gathered}
$$

a)

$$
\begin{aligned}
& \sigma_{V_{O S}}^{2} \cong \frac{2}{(0.5 \mu)^{2}}\left[.021^{2}+\frac{1}{3} .025^{2}\right] \\
& \sigma_{V_{O S}} \cong 72 \mathrm{mV} \\
& 3 \sigma_{V_{O S}} \cong 216 \mathrm{mV}
\end{aligned}
$$

Note this is a very large offset voltage!

## Random Offset Voltages

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

$$
\text { Assume } \mathrm{L}_{\mathrm{MIN}}=\mathrm{W}_{\mathrm{MIN}}=0.5 \mathrm{u}, \mathrm{~A}_{\text {VTOn }}=0.021 \mathrm{~V} \text { and } \mathrm{A}_{V T O P}=0.025 \mathrm{~V}
$$

$$
\begin{aligned}
& \sigma_{V_{0 S}}^{2} \cong 2\left[\frac{A_{V T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{V T O p}^{2}\right] \\
& \sigma_{V_{V S}}^{2} \cong \\
& \frac{2}{W_{n} L_{n}}\left[A_{V T O n}^{2}+\frac{\mu_{p}}{\mu_{n}} A_{V T O p}^{2}\right]
\end{aligned}=
$$

b)

$$
\begin{gathered}
\sigma_{V_{O S}}^{2} \cong \frac{2}{[100(0.5 \mu)]^{2}}\left[.021^{2}+\frac{1}{3} .025^{2}\right] \\
\sigma_{V_{O S}} \cong 7.2 \mathrm{mV} \\
3 \sigma_{V_{\mathrm{OS}}} \cong 21.6 \mathrm{mV}
\end{gathered}
$$

Note this is much lower but still a large offset voltage !
The area of $M_{1}$ and $M_{3}$ needs to be very large to achieve a low offset voltage

## Random Offset Voltages


(a)

(b)

It can be shown that

$$
\sigma_{V_{O S}}^{2} \cong 2 \mathrm{~V}_{\mathrm{t}}^{2}\left[\frac{\mathrm{~A}_{\mathrm{Jn}}^{2}}{\mathrm{~A}_{\mathrm{En}}}+\frac{\mathrm{A}_{\mathrm{Jp}}^{2}}{\mathrm{~A}_{\mathrm{Ep}}}\right]
$$

where very approximately

$$
A_{J n}=A_{J p}=0.1 \mu
$$

## Random Offset Voltages

Example: Determine the $3 \sigma$ value of the offset voltage of a the bipolar input stage if $A_{E 1}=A_{E 3}=10 \mu^{2}$

$$
\begin{aligned}
& \sigma_{V_{O S}}^{2} \cong 2 \mathrm{~V}_{\mathrm{t}}^{2}\left[\frac{\mathrm{~A}_{\mathrm{Jn}}^{2}}{\mathrm{~A}_{\mathrm{En}}}+\frac{\mathrm{A}_{\mathrm{Jp}}^{2}}{\mathrm{~A}_{\mathrm{Ep}}}\right] \\
& \sigma_{V_{O S}} \cong \sqrt{2} \mathrm{~V}_{\mathrm{t}} \mathrm{~A}_{\mathrm{J}} \frac{\sqrt{2}}{\sqrt{\mathrm{~A}_{\mathrm{E}}}} \\
& \sigma_{V_{O S}} \cong 2 \cdot 25 \mathrm{mV} \cdot 0.1 \mu \bullet \frac{1}{\sqrt{10 \mu^{2}}}=1.6 \mathrm{mV} \\
& 3 \sigma_{V_{O S}} \cong 4.7 \mathrm{mV}
\end{aligned}
$$



## Random Offset Voltages

## Typical offset voltages:

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

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 $\varphi_{1}$ is low, $\mathrm{V}_{1}$ and $\mathrm{V}_{2}$ are precharged to $\mathrm{V}_{\mathrm{DD}}$ and no static power is dissipated When $\varphi_{1}$ 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, RL Geiger - 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
s Save 20 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 $\mathrm{V}_{\text {os }}$ 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 ( $A_{W}$ and $A_{L}$ 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:

$$
\sigma_{V_{0 S}}^{2} \cong\left[\frac{A_{v T O n}^{2}}{W_{n} L_{n}}+\frac{\mu_{p}}{\mu_{n}} \frac{L_{n}}{W_{n} L_{p}^{2}} A_{\text {vTOp }}^{2}\right]
$$

Bipolar:

$$
\sigma_{V_{O S}}^{2} \cong 2 V_{t}^{2}\left[\frac{A_{J n}^{2}}{A_{E n}}+\frac{\mathrm{A}_{\mathrm{Jp}}^{2}}{\mathrm{~A}_{E \mathrm{p}}}\right]
$$



## Stay Safe and Stay Healthy !

## End of Lecture 11

