Digital Data Acquisition

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Introduction

- Instruments and data can be either *analog* (continuous) or *digital* (discrete).
- In this learning module, we discuss digital data, and how to convert between analog and digital signals. •

Review - Decimal to Binary Conversion

- While we are used to working with *decimal numbers* (base 10), computers and many modern instruments and electronic devices use *binary numbers* (base 2).
- Decimal numbers contain the digits 0 through 9, while binary numbers contain only digits 0 and 1. •
- As a review, here is a listing of the first eight (including zero) decimal and binary numbers:

Decimal number	Binary number
0	0
1	1
2	10
3	11
4	100
5	101
6	110
7	111

- There are simple techniques to convert from decimal to binary and vice-versa: •
 - Decimal to binary: 0
 - Successively divide by 2, as in the following example: convert the decimal number 29 into binary:

Value/2	Integer value	Remainder	
29/2 = 14 + 1/2	14	1	
14/2 = 7 + 0/2	7	0	
7/2 = 3 + 1/2	3	1	
3/2 = 1 + 1/2	1	1	
1/2 = 0 + 1/2	0	1	

- The binary number is obtained from the "remainder" column, reading *from bottom to top* (backwards); the answer is 11101.
- Binary to decimal: 0
 - Add up each non-zero bit, from left to right, converting the binary number 11101 into decimal:

,	0,	5	
Bit considered (underlined)	Value of the place	Value of the bit	
1110 <u>1</u>	1	$1 \times 1 = 1$)
111 <u>0</u> 1	2	$0 \times 2 = 0$	
11 <u>1</u> 01	4	$1 \times 4 = 4$	
1 <u>1</u> 101	8	$1 \times 8 = 8$	
<u>1</u> 1101	16	1 × 16=16	J



ead this way, ttom to top.

The decimal number is obtained by summing the "Value of the bit" column; the answer is 29.

Example:

Given: The decimal number x = 4

To do: Convert *x* to a binary 4-bit number.

Solution: We use the technique of successive division by 2, as per the following table:

Value/2	Integer value	Remainder	
4/2 = 2 + 0/2	2	0	Deed this way
2/2 = 1 + 0/2	1	0	Read this way,
1/2 = 0 + 1/2	0	1	bottom to top.

The resulting binary number is read from the bottom to the top, x = 0100 in binary form. We have added a leading zero so that the result is a 4-bit number. If we were asked for 8 bits, we would write x = 0000 0100.

Some calculators have built-in binary-to-decimal and decimal-to-binary converters.

Analog to Digital Conversion (A/D conversion)

- Digital data acquisition is used in digital multimeters, digital oscilloscopes, computer-controlled data acquisition systems, and many other modern instruments and electronic devices such as cell phones.
- In all these examples, the conversion of an analog signal into a digital signal is accomplished with an electronic device called an *analog-to-digital converter*, which we abbreviate *A/D converter*.
- The goal of an A/D converter is to change an *analog voltage signal* into a *digital number* (in binary form).
- An A/D converter is labeled as *N*-bit, where the *number of bits N* represents how many bits (ones and zeroes) are used in the digital output of the A/D converter. E.g., an 8-bit converter creates an 8-bit output.
- For simplicity, consider a 2-bit A/D converter (N = 2) with a *range* of -5 to 5 volts. The voltage range is divided into *bins* as follows: [*Note*: This is just one of several optional ways to assign the bins.]

Analog voltage (volts)	Bin number	Digital output (binary)	Assigned voltage, V (volts)
$-5 \le V < -2.5$	0	00	$V = V_{\min} + 0.5\Delta V = -3.75$
$-2.5 \le V \le 0$	1	01	$V = V_{\min} + 1.5\Delta V = -1.25$
$0 \le V < 2.5$	2	10	$V = V_{\min} + 2.5\Delta V = 1.25$
$2.5 \le V < 5$	3	11	$V = V_{\min} + 3.5\Delta V = 3.75$

- The *assigned voltage* for each bin is here defined as half-way between the limits of the bin.
- The *number of bins* = 2^{N} for an *N*-bit A/D converter. For our example 2-bit converter, there are $2^{2} = 4$ bins.

• The *resolution* of the A/D converter is defined as resolution =
$$\Delta V = \frac{(V_{\text{max}} - V_{\text{min}})}{2^N}$$

- In the present example (2-bit A/D converter), the resolution is $\Delta V = (5 (-5))/2^2 = 2.5 \text{ V}$.
- Alternately, the resolution can be expressed as half of this value on either side, i.e., ± 1.25 V. Another name for the resolution expressed this way is *quantization error* (also sometimes called *quantizing error*), defined

as quantization error =
$$\pm \frac{1}{2}$$
 resolution = $\pm \frac{1}{2}\Delta V = \pm \frac{1}{2}\frac{(V_{\text{max}} - V_{\text{min}})}{2^N} = \pm \frac{(V_{\text{max}} - V_{\text{min}})}{2^{N+1}}$

- In the present example (2-bit A/D converter), the quantization error is ± 1.25 V.
- *This quantization error is too large for most practical applications*. Notice that we cannot tell the difference between an input of 2.6 V and 4.9 V both of these input voltages fall into bin number 3, and would be assigned the output voltage of 3.75 V.
- We might express this reading as V = 3.75 V +/- 1.25 V to reflect the resolution of the A/D converter.
- Obviously, the bigger N, the better the resolution. Consider, for example, a 12-bit A/D converter (N = 12).
 - The number of bins = 2^{12} = 4096, which is a digital output of 0 to 4095.
 - The resolution (for a converter with a range of -5 to 5 V) is $\Delta V = 0.00244141$ V.
 - The quantization error is $\pm \Delta V/2 = \pm 0.00122070$ V.
- Selected rows of the bin table for a 12-bit A/D converter with a range of -5 to 5 V are shown below: [*Note*: Again, this is just one of several optional ways to assign the bins.]

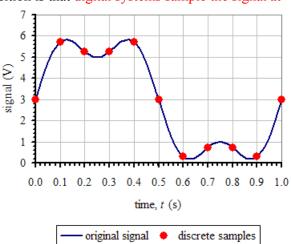
Analog voltage (volts)	Bin number	Digital output (binary)	Assigned voltage, V (volts)
$-5 \le V < -4.9976$	0	0000 0000 0000	$V = V_{\min} + 0.5\Delta V = -4.998779$
$-4.9976 \le V < -4.9951$	1	0000 0000 0001	$V = V_{\min} + 1.5\Delta V = -4.996338$
etc	etc	etc	etc
$0 \le V < 0.00244$	2048	1000 0000 0000	$V = V_{\min} + 2048.5 \Delta V = 0.0012207$
etc	etc	etc	etc
$4.9951 \le V \le 4.9976$	4094	1111 1111 1110	$V = V_{\min} + 4094.5 \Delta V = 4.996338$
$4.9976 \le V < 5$	4095	1111 1111 1111	$V = V_{\min} + 4095.5\Delta V = 4.998779$

- Comparing the two tables above, it should be obvious that the quantization error is much less for the 12-bit A/D converter than for the 2-bit (with both converters covering the same range of -5 to 5 V.)
- Most commercially available A/D converters are 8, 12, 14, or 16-bit. [Those in our lab are 16-bit.]
- The range of most modern A/D converters is adjustable, and can be either *monopolar* (e.g., 0 to 1 V, 0 to 5 V, 0 to 10 V, etc.) or *bipolar* (e.g., -1 to 1 V, -10 to 10 V, etc.).
- **Dynamic range**, **DR** = ratio of the largest voltage to the smallest *change* in voltage that can be measured, expressed in decibels (dB): monopolar: $DR = 20\log_{10}(2^N)$, bipolar: $DR = 20\log_{10}(2^{N-1})$.

Some authors call this the *signal-to-quantization noise ratio*, *SQNR* (same equations as for *DR*).

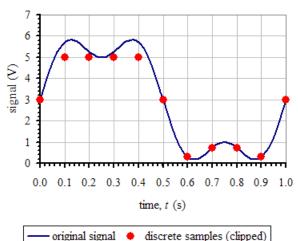
Discrete Sampling

- The main difference between analog and digital data acquisition is that digital systems sample the signal *at discrete times only*, not continuously.
- In a digital data acquisition system, no information is recorded at times *in between* the discrete sampling times.
- Consider a continuous signal that is sampled discretely at *sampling frequency fs*, as sketched to the right. Here, $f_s = 10$ Hz.
- For a given sampling frequency f_s , the *time period* Δt between samples is the inverse of f_s , $\Delta t = 1/f_s$. Here, $\Delta t = 1/(10 \text{ s}^{-1}) = 0.1 \text{ s}$.
- *Warning*: If we are not careful, discrete sampling can lead to incorrect conclusions about the original signal!
- Two of the potential problems (*clipping* and *aliasing*) are discussed in detail below:



Clipping

- If the original signal lies outside of the range of the A/D converter, it is *clipped* at the extreme value. In other words, any voltage greater than the upper limit of the A/D converter is assigned the maximum voltage of the A/D converter.
- Likewise, any voltage less than the lower limit of the A/D converter is assigned the minimum voltage of the A/D converter.
- For example, for a 12-bit A/D converter with a range of -5 to 5 V, any voltage above 5 V is assigned the maximum voltage of 4.9988 V, according to the above bin table.
- In most cases, the user does not know that the signal has been clipped, and this can lead to incorrect results, although some data acquisition systems give a warning when the signal gets clipped.
- A clipped signal (same signal as above, but clipped when using a -5 to 5 V A/D converter) is illustrated here.
- Notice that all voltages above 5 V in this example have been *clipped* to 5 V.
- Clipping would also occur if the voltage dips below -5 V.



Aliasing

- If the sampling frequency *f_s* is *too low*, the digital data acquisition system can actually measure an *incorrect frequency*! This is called *aliasing*.
- Consider a pure sine wave.
- The equation for a general sine wave signal is $g(t) = C + A\sin(2\pi ft \phi)$ or $g(t) = C + A\sin(\omega t \phi)$, where *f* is the *frequency* (hertz), ω is the *radian frequency* (radians/s) and $\omega = 2\pi f$, *A* is the *amplitude* (volts), *C* is the *DC offset* (volts), *t* is the *time* (seconds), and ϕ is the *phase shift* (radians).
- Aliasing is best illustrated by example.
- Suppose the original signal is a pure 10 Hz sine wave, with an amplitude of 3 volts, a DC offset of 3 volts, and a phase shift of 90 degrees ($\pi/2$ radians):
 - The frequency of the signal is f = 10.0 Hz.
 - The *period* of the signal is T = 1/f = 0.100 s.
 - The amplitude of the signal is A = 3.00 volts.
 - The DC offset of the signal is C = 3.00 volts.
 - The phase shift of the signal is $\phi = \pi/2$ radians.
 - For the sine wave used in this example, the equation of the signal is $|g(t) = 3 + 3\sin(2\pi(10)t \pi/2)|$
- This example sine wave is easily simulated in Excel or Matlab, as is the discrete sampling (and aliasing).

- Also, suppose we sample discretely (digitally) at a sampling frequency $f_s = 15$ Hz, 15 times per second, or • one sample every $\Delta t = 1/(15 \text{ s}^{-1}) = 0.06666... \text{ s}.$
- A half second of both the analog signal and the discretely ٠ sampled data points is shown on the plot to the right.
- The aliasing is obvious *the perceived signal looks* • *nothing like the original*! In fact, the *apparent frequency* of the *inferred* or *perceived signal* (formed by connecting the discrete data points with straight line segments) is 5 Hz, and it is an odd-looking trapezoidal waveform rather than a sine wave.
- It turns out that if the sampling frequency f_s is greater • than two-thirds of the actual frequency f, but less than twice the actual frequency, the *perceived frequency* (also called the *aliasing frequency*, f_a) is equal to the absolute value of the difference between the sampling frequency and the actual frequency.
- In equation form, the above statement is: 2

if
$$\frac{2}{3}f < f_s < 2f$$
, then $f_a = |f_s - f|$.

In the above example, f = 10 Hz and $f_s = 15$ Hz, so $\frac{2}{3}f = \frac{2}{3}10 = 6.66666...$, which is less than $f_s = 15$ Hz. • Therefore, the perceived frequency or aliasing

frequency is $f_a = |f_s - f| = |15 - 10| = 5$ Hz. This agrees with the observed perceived frequency on the above plot (red line).

- Here is another example. For the same sine wave (f =• 10 Hz), the signal is plotted to the right for 1 second. Data are sampled discretely at $f_s = 11$ Hz.
- Note that the perceived signal looks like a sine wave at 1 Hz! We check the predicted aliasing frequency by

using the above equation: Since $\frac{2}{3}f = \frac{2}{3}10 = 6.6666...$,

which is less than $f_s = 11$ Hz, the perceived frequency or aliasing frequency is $f_a = |f_s - f| = |11 - 10| = 1$ Hz. This agrees with the observed perceived frequency on the above plot (red line).

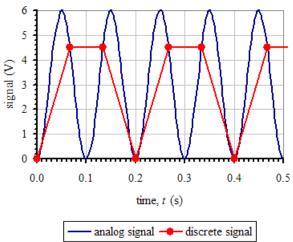
- One more example for the same sine wave (f = 10 Hz): Data are sampled discretely at $f_s = 9$ Hz, and the data • are plotted to the right.
- Note that the perceived signal also looks like a sine ٠ wave at 1 Hz! We check the predicted aliasing frequency by using the above equation: Since

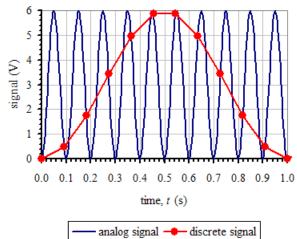
$$\frac{2}{3}f = \frac{2}{3}10 = 6.66666...$$
, which is less than $f_s = 9$ Hz, the

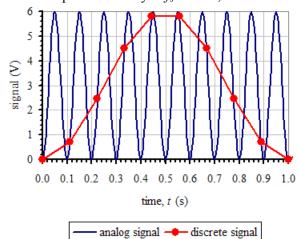
perceived frequency or aliasing frequency is $f_a = |f_s - f| = |9 - 10| = 1$ Hz. This agrees with the observed perceived frequency on the plot to the right (red line).

Comparing the plots for $f_s = 11$ Hz and $f_s = 9$ Hz, they • look similar at first glance, since the perceived frequency is 1 Hz in both cases. However, careful inspection reveals that the sampled data points occur at

different phases of the signal in the two plots. Furthermore, the 11-Hz plot contains 12 data points, while the 9-Hz plot contains only 10 data points between t = 0 and t = 1.0 s.







The sampling rate theorem and prediction of aliasing frequency

- The sampling rate theorem, also called the Nyquist theorem, helps us avoid aliasing. The sampling rate theorem is stated as follows:
 To avoid aliasing, the sampling frequency must be greater than twice the highest frequency component of the analog signal.
- The Nyquist criterion is $f_s > 2f$. The Nyquist criterion must be met in order to avoid aliasing.
- To determine if there is aliasing, and, if so, to calculate the aliasing frequency, we apply the following rules, depending on the relative values of signal frequency f and sampling frequency f_s :
 - If $f_s > 2f$, then there is no aliasing. In words, if the sampling frequency is greater than twice the signal frequency, there is *no* aliasing.
 - If $\frac{2}{3}f < f_s < 2f$, then $f_a = |f_s f|$. In words, if the sampling frequency is greater than two-thirds the

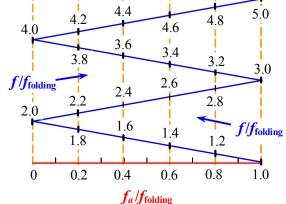
signal frequency but less than twice the signal frequency, there *is* aliasing, and the aliasing frequency is equal to the absolute value of the difference between the sampling frequency and the signal frequency.

• If
$$f_s < \frac{2}{3}f$$
, then $f_a = \left(\frac{f_a}{f_{\text{folding}}}\right)f_{\text{folding}}$, where f_{folding} is the *folding frequency*, defined as $f_{folding} = \frac{f_s}{2}$, and

the ratio f_a/f_{folding} is determined from the *folding diagram*. In words, if the sampling frequency is less than two-thirds of the actual frequency, the aliasing frequency must be calculated from the folding diagram. A summary of the procedure is given below:

- Calculate the folding frequency, $f_{\text{folding}} = f_s/2$.
- Locate f/f_{folding} on the folding diagram, as plotted on the right. *Note*: For values of f/f_{folding} greater than 5.0, the folding diagram can easily be extended, following the obvious pattern.
- Read straight down from the value of f/f_{folding} to obtain the value of f_a/f_{folding} on the bottom (horizontal) axis.
- <u>Finally, calculate the aliasing frequency</u>,

$$f_a = \left(\frac{f_a}{f_{\text{folding}}}\right) f_{\text{folding}}$$



- The folding frequency is *half* of the sampling frequency because of the Nyquist criterion – you must sample at a frequency at least *twice* the signal frequency in order to avoid aliasing.
- A general equation [Shaparenko, B. and Cimbala, J. M., *Int. J. Mech. Engr Education*, Vol. 39, No. 3, pp. 195-199, 2012] is available to determine the perceived frequency of *any* signal frequency *f* when sampled at

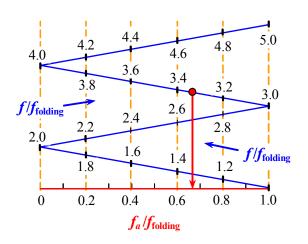
any sampling frequency f_s , whether there is aliasing or not: $\int_{\text{perceived}} = \int f - f_s \cdot \text{NINT}$

not: $J_{\text{perceived}} = \left| J - J_s \cdot \text{NIN I} \left(\frac{-}{f_s} \right) \right|$

- NINT is the "nearest integer" function.
- In Excel, use ROUND(x,0) to round real number x to the nearest integer.

<u>Example</u>:

- *Given:* A sine wave of frequency 10 Hz is sampled at a sampling frequency of 6 Hz.
- *To do:* Calculate the *perceived* frequency of the sampled signal.
- *Solution:* We follow the procedure outlined above.
- For f = 10 Hz, 2f/3 = 2(10 Hz)/3 = 6.666666...
- Since $f_s = 6$ Hz is *less than* 2f/3 = 6.66666..., the simple formula cannot be used. We use the folding diagram to calculate the aliasing frequency f_a .
- The folding frequency is $f_{\text{folding}} = f_s/2 = 6/2 = 3$ Hz.



- 0
- We calculate $f/f_{\text{folding}} = 10/3 = 3.33333...$ We locate this value of f/f_{folding} on the folding diagram (see above), and read down to the f_a/f_{folding} axis at 0 the bottom: At $f/f_{\text{folding}} = 3.33333..., f_a/f_{\text{folding}} = 0.666666...$ Finally, we calculate the aliasing frequency:
- 0

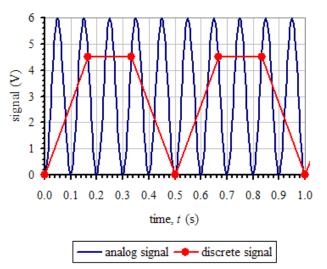
 $f_a = \left(\frac{f_a}{f_{\text{folding}}}\right) f_{\text{folding}} = (0.666666...)(3 \text{ Hz}) = 2 \text{ Hz}.$ The perceived signal will be aliased with an aliasing frequency of $f_a = 2$ Hz

Alternate Solution: We use the general equation for perceived frequency.

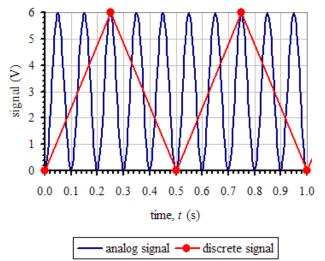
- For f = 10 Hz and $f_s = 6$ Hz, $f_{\text{perceived}} = \left| f f_s \cdot \text{NINT}\left(\frac{f}{f_s}\right) \right| = \left| 10 6 \cdot \text{NINT}\left(\frac{10}{6}\right) \right| = \left| 10 6 \cdot 2 \right| = 2$ Hz. 0
- We wee that the general equation yields the correct perceived frequency, i.e., the aliasing frequency in 0 this case, without having to calculate the folding frequency or use the folding diagram.

Discussion:

- Since f_s is less than 2f/3 in this example, we 0 cannot use the simple difference equation to calculate the aliasing frequency. In other words, $f_a \neq |f_s - f| = |6 - 10| = 4$ Hz. Rather, $f_a = 2$ Hz.
- We simulate this example using Excel, and a plot 0 of the actual (analog) signal and the perceived (discrete digital) signal is shown to the right. The perceived signal does not even *look* like a sine wave, but it does indeed have a frequency of 2 Hz, as predicted.
- You are encouraged to use the accompanying Excel . *spreadsheet to experience aliasing* – the spreadsheet is set up so that you can change the sampling frequency f_s and watch the plot change. The signal frequency is



- 10 Hz, but this, along with the amplitude, DC offset, and phase shift, can also be changed easily.
- For example, if f_s is changed to 4 Hz, a *triangular* wave pattern is seen as the perceived signal, with an • aliasing frequency of 2 Hz, as shown below. Try to predict this aliasing frequency using the above procedure.



Finally, we note that the folding diagram and the general equation can *always* be used, regardless of the • values of f and f_s . In other words, whether or not there is aliasing, and whether or not $f_s < 2f/3$, you can still calculate the perceived frequency by following the procedure outlined above, using either the folding diagram or the general equation for perceived frequency.