EngSci Year 3 Fall 2022 Notes

Brian Chen

Division of Engineering Science University of Toronto https://chenbrian.ca

brianchen.chen@mail.utoronto.ca

Contents

I	ECE349: Introduction to Energy Systems	1
1	Admin stuff	1
	1.1 Lecture 1	1
	1.1.1 Mark breakdown	1
2	AC Steady State Analysis	1
	2.1 Lecture 2	1
	2.1.1 TODO	1
	2.2 Lecture 3	3
3	AC Power	4
	3.1 Lecture 4	5
	3.1.1 Root Mean Squared (RMS) Values	6
II	ECE352: Computer Organization	7
4	Admin stuff	o
4	4.1 Lecture 1	8
	4.1.1 Mark breakdown	8
5	Preliminary	8
	5.1 Lecture 2: Using binary quantities to represent other things	8
	5.1.1 Floating Point Numbers	9
6	NIOS II	10
	6.1 Lecture 3: Behavioural Model of Memory	10
	6.1.1 Memory	10
	6.1.2 Physical Interface	11
	6.2 Lecture 4: NIOS II Programming Model	12
	6.2.1 Adding Two Numbers 6.2.2 Adding two numbers using memory	13
	6.2.2 Adding two numbers using memory	15
II	I ECE355: Signal Analysis and Communication	17
7	Admin and Preliminary	17
	7.1 Lecture 1	17
	7.1.1 Mark Breakdown	17

8	Transformations	18
	8.1 Lecture 2	18
	8.2 Lecture 3	18
	8.2.1 General Continuous Complex Exponential Signals	19
	8.3 Lecture 4: Step and Impulse Functions	20
IV	V ECE360: Electronics	22
9	Admin and Preliminary	22
	9.1 Lecture 1	22
	9.1.1 Mark Breakdown	22
	9.1.2 Diodes	22
10	0 Diodes	23
	10.1 Lecture 2	23
	10.2 Lecture 3	26
11	1 Lecture 4: Forward conducting diodes	27
	11.0.1 Small-Signal Model	29
\mathbf{v}	ECE358: Foundations of Computing	30
12	2 Admin and Preliminary	30
	12.1 Lecture 1	30
	12.1.1 Mark Breakdown	31
13	3 Complexities	31
	13.1 Lecture 2	31
	13.2 Lecture 3: Logs & Sums	33
	13.2.1 Functional Iteration	33
	13.3 Lecture 4: Induction & Contradiction	35
VI	T MAT389: Complex Analysis	35
14	4 Complex Numbers	35
	14.1 Lecture 1	35
	14.2 Lecture 2	37
	14.2.1 Functions on complex planes	38
	14.2.2 Exponential Functions	39
VI	TI ECE444: Software Engineering	41
15	5 Preliminary	41
	15.1 Lecture 1, 2	41

16 Project Management	41
16.1 Lecture 3	42
16.1.1 Agile	43
16.2 I dropped this course	43
VIII ESC301: Seminar 17 Preliminary 17.1 Seminar 1	43 44 44

ECE349: Introduction to Energy Systems

Section 1

Taught by Prof. P. Lehn

Admin stuff

Subsection 1.1

Lecture 1

First lecture was logistical info + a speil about how power systems are one of the great modern wonders. Course will cover sinusoidal AC power systems (1, 3 phase), power systems (dc-dc, dc-ac conversion), and magnetic systems (transformers, actuators, and synchronous machines)

1.1.1 Mark breakdown

- 50 % Final
- 25 % Midterm
- 5 % Quiz
- 15 % Labs
- 5 % Assignments

Section 2

AC Steady State Analysis

Subsection 2.1

Lecture 2

2.1.1 TODO

• Review Thomas 669-600

What we have learnt prior for differential equations enables us to arrive at analytical solutions to linear stable AC systems with phasors. A homogeneous and particular solution will be produced. If there's a stable homogeneous solution, $\to 0$ as $t \to \infty$. The full solution would be the addition of the two via super position.

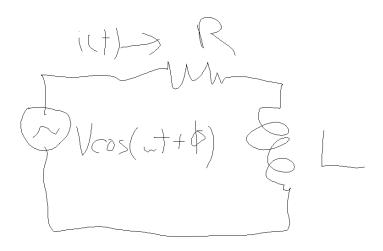
We generally use this approach to solve circuits since it's an efficient way to solve circuits and make them into essentially DC circuits.

Recall, for a general phasor \hat{P}

•
$$\frac{d\hat{P}}{dt} = jw\hat{P}$$

•
$$\int \hat{P} = \frac{1}{jw}\hat{P}$$

AC Steady State Analysis Lecture 2 2



$$Ri + L\frac{di}{dt} = V\cos(\omega t + \phi) \tag{2.1}$$

But this is a pain to solve. It can be made simpler by applying phasors

$$V\cos(\omega t + \phi) = Re\{Ve^{j(\omega t + \phi)}\}\tag{2.2}$$

Take the real part of \hat{I} :

$$R\hat{I} + L\frac{d\hat{I}}{dt} = Ve^{j(\omega t + \phi)}$$
 (2.3)

And therefore by inspection the solution is of format $\hat{I}e^{j\omega t}$, where \hat{I} is a phasor. Noting that \hat{I} contains only amplitude and phase,

$$\begin{split} R\hat{I}e^{j\omega t} + L\frac{d}{dt}(\hat{I}e^{j\omega t}) &= Ve^{j\omega + \phi} \\ R\hat{I} + L\hat{I}j\omega &= Ve^{j\phi} \end{split} \tag{2.4}$$

And now reconstructing:

$$\hat{I} = \frac{V}{\sqrt{R^2(\omega L)^2}} e^{j(\omega t + \phi - \tan^{-1} \frac{wL}{R})}$$

$$i(t) = Re\left\{\hat{I}\right\}$$
(2.5)

And therefore

$$\hat{I} = \frac{V}{\sqrt{R^2(\omega L)^2}} \cos\left(\omega t + \phi - \tan^{-1}\left(\frac{wL}{R}\right)\right)$$
 (2.6)

Notation: $Xe^{j\phi} \leftrightarrow X\lfloor \phi$

The steps to solving a phasor problem are:

- Define phasor: $V\cos{(\omega t + \phi)} \leftrightarrow V\lfloor \phi$
- Map L,C into phasor domain; find impedences

–
$$v=L\frac{di}{dt}\leftrightarrow \hat{V}=j\omega L\hat{I}$$

AC Steady State Analysis Lecture 3 3

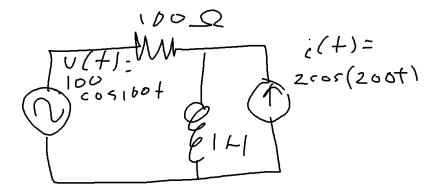
$$-i = C \frac{dv}{dt} \leftrightarrow \hat{V} = \frac{1}{i\omega C} \hat{I}$$

- Do mesh analysis to find $\hat{I};\hat{I} = \frac{\hat{V}}{\sum \text{impedances}}$
- Reconstruct i(t) from \hat{I}

Subsection 2.2

Lecture 3

Phasors allow us to solve circuits with multiple sources of differing frequencies.



To find the current i(t) over the inductor we can find it's response due to the voltage and current sources and then apply superposition.

•
$$I_1 = \frac{100 \lfloor 0}{1oo + j100} = 0.707 \lfloor -45^o \rightarrow i_1(t) = 0.707 cos(100t - 45^o)$$

•
$$I_2 = \frac{100}{100+j200} 2\lfloor 0 = 0.894 \lfloor -65^o \rightarrow i_2(t) = 0.894 cos(200t - 63^o)$$

•
$$i(t) = i_1(t) + i_2(t) = 0.707\cos(100t - 45^\circ) + 0.894\cos(200t - 63^\circ)$$

Non-sinusoidal stimulus may be solved by decomposing the signal with Fourier transforms. For example, square waves:

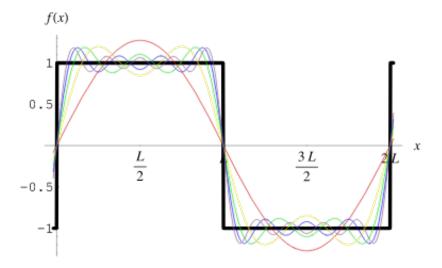


Figure 1. square waves with Fourier series superimposed

AC Power

The general form of a Fourier transform is given as:

$$v_{equiv}(t) = a_o + \sum_{n=1}^{\infty} a_k cos(nw_o t) + b_k cos(nw_o t)$$
(2.7)

Where:

$$a_{o} = \frac{1}{T} \int_{0}^{T} v(t)dt$$

$$a_{k} = \frac{2}{T} \int_{0}^{T} v(t)cos(nw_{o}t)dt$$

$$b_{k} = \frac{2}{T} \int_{0}^{T} v(t)sin(nw_{o}t)dt$$

$$(2.8)$$

Armed with Fourier series and superposition we may now model a non-sinusoidal signal as a superposition of an infinite sum of sources. About half the work can be cut in half by recognizing that sin lags cos by 90^o , so

$$a_o = \frac{1}{T} \int_0^T v(t)dt$$

$$a_k = \frac{2}{T} \int_0^T v(t)cos(nw_o t)dt$$

$$b_k = \frac{2}{T} \int_0^T v(t)cos(nw_o t - 90^o)dt$$
(2.9)

Section 3

AC Power

Instantaneous Power: $p(t) = v(t) \times i(t)[W, \frac{J}{s}]$

Example

For a circuit with a voltage source, $v(t) = V cos(\omega t)$ and a resistor Ω , $i(t) = I cos(\omega t)$, $p(t) = V I cos^2(\omega t) = \frac{VI}{2}(1 + cos(2\omega t))$

Definition 2

Definition 1

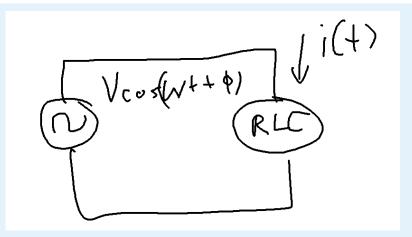
Average Power over Cycle: $P(t) = \frac{1}{T} \int_o^T p(t) dt = \frac{VI}{2}$

If we were to plot the instantaneous power we see that due to the sinusoidal response there are times where 0 power is supplied. This will always be true for a single phase power supply; real-world supplies always have multiple phases; this is why computer PSUs always contain a ton of capacitors.

Definition 3

Reactive Power: Q

AC Power Lecture 4 5



If $\phi_i = 0$ and taking $\phi = \phi_v$,

 $\phi = \phi_v - \phi_i$

$$p(t) = V \cos(\omega t + \phi) * I \cos(\omega t)$$

$$= \frac{VI}{2} \cos(\phi) + \cos(2\omega t + \phi)$$

$$= \underbrace{\frac{VI}{2} \cos(\phi)(1 + \cos(2\omega t) - \underbrace{\frac{VI}{2} (\sin \phi \sin 2\omega t)}_{\text{stored and released back to source}}$$
(3.1)

Taking the average power of the reactive power we get

$$P_{avg} = \frac{VI}{2}\cos\phi \tag{3.2}$$

Another quantity, reactive power, can be defined with regards to the energy sloshing back and forth:

$$Q = \frac{VI}{2}\sin\phi\tag{3.3}$$

Subsection 3.1

Lecture 4

Definition 4

Displacement factor

 $DF \equiv \cos \phi$

Where ϕ is the angle measured from the \hat{V} to the \hat{I} phasors. A DF=1 means the system is transferring the most power possible.

• Lagging DF: ϕ is -'ve

- Leading DF: ϕ is +'ve'

We can also write P, Q from phasors.

AC Power Lecture 4 6

$$P = \frac{\hat{V}\hat{I}}{2}\cos(\phi_v - \phi_i)$$

$$= \frac{1}{2}\hat{V}\hat{I}Re\{e^{j\phi_v - \phi_i}\}$$

$$= \frac{1}{2}Re\{Ve^{j\phi_v} \cdot Ie^{-j\phi_i}\}$$

$$= \frac{1}{2}Re\{\hat{V}\hat{I}^*\}$$
(3.4)

$$Q = \frac{\hat{V}\hat{I}}{2}\sin(\phi_v - \phi_i)$$

$$\vdots$$

$$= \frac{1}{2}Im\{\hat{V}\hat{I}^*\}$$
(3.5)

The expressions for Q and P are basically the same so we define a new quantity, complex power, which incorporates both the imaginary and real components.

Reference: Thomas 16.1, 16.2, 16.3

Definition 5

Complex Power

$$S = \frac{1}{2}\hat{V}\hat{I} * = P + jQ \tag{3.6}$$

3.1.1 Root Mean Squared (RMS) Values

A RMS value measures the average power of a periodic signal. Consider a circuit with an AC voltage source with v(t) flowing through a resistor.

The power is:

$$p(t) = v(t)i(t) = v(t)\frac{v(t)}{R} = \frac{1}{R}v(t)^{2}$$
(3.7)

The average power is therefore

$$P = \frac{1}{R} \int_{0}^{T} v(t)^{2} dt$$
 (3.8)

Evaluating this becomes easier by defining an useful quantity, RMS voltage

$$v_{rms} = \sqrt{\frac{1}{T} \int_0^T v(t)^2 dt} \tag{3.9}$$

And using v_{rms} we get a nice expression for average power,

$$P = \frac{1}{R}v_{rms}^2 \tag{3.10}$$

More generally speaking we can define RMS values of sinusoidal signals

Definition 6

$$v_{rms} = \sqrt{\frac{1}{2\pi} \int_0^{2\pi} (\hat{V}\cos wt)^2 dt} = \frac{1}{\sqrt{2}} \hat{V}$$
 (3.11)

Plugging this into the expressions for complex power:

$$S = V \cdot I^* \tag{3.12}$$

Where \underline{V} , $\underline{I^*}$ are RMS phasors at a given common frequency.

And more generally yet, the RMS values of non-sinusoidal signals can be found with help of a Fourier expansion

Definition 7

Let $v(t) = \hat{v}_0 + \hat{v}_1 \cos(wt + \phi_1) + \dots$ Then.

$$v_{rms} = \sqrt{\hat{v}^2 + \frac{\hat{v}_1}{\sqrt{2}}^2 + \frac{\hat{v}_2}{\sqrt{2}}^2 \dots} = \sqrt{v_0^2 + \sum_{n=1}^{\infty} (\frac{V_m^2}{2})}$$
(3.13)

And if V_m is the rms value,

$$v_{rms} = \sqrt{v_0^2 + \sum_{n=1}^{\infty} V_m^2} \tag{3.14}$$

One thing to watch out for is that we could have a system with high voltage but near-zero current transferring little to no power. To account for this we look at the power factor 'PF'

Definition 8

Power Factor
$$PF \equiv \frac{\text{average power}}{\text{rms voltage} \cdot \text{rms current}}$$
(3.15)

For a sinusoidal V, I:

$$PF = \frac{\frac{1}{2}\hat{V}\hat{I}\cos\phi}{\frac{\hat{V}}{\sqrt{2}}\cdot\frac{\hat{I}}{\sqrt{2}}}$$

$$= \cos\phi$$
(3.16)

If signals are not harmonics PF = DF. This can be a source of confusion.

For non-sinusoidal systems this becomes more difficult because, unlike pure signals, they may contain harmonics. In these systems either V, I, or both may contain harmonics. Generally in household power V is clean but I contains harmonics.

The effect of harmonics in currents is that $I_{\text{harmonics}}$ causes a higher I_{rms} ; there is more current but no higher power! This reduces PF and causes PF < DF.

To summarize,

- In ideal systems, PF = DF if all V and I are at one frequency.
- PF = 1 means no energy sloshing between load and source.

Harmonics are bad because they reduce the usefulness of the system There are very tight standards for how many harmonics one is allowed to inject into the system via generators or loads. See: textbook 16.6

ECE352: Computer Organization

PART

Taught by Prof. Andreas Moshovos

Preliminary 8

Section 4

Admin stuff

Subsection 4.1

Lecture 1

- Lecture recordings on YouTube
- Online notes: https://www.eecg.utoronto.ca/ moshovos/ECE352-2022/
- Course will cover the following:
 - C to assembly
 - How to build a processor that works
 - Intro to processor optimizations
 - Peripherals
 - OS support (Maybe)
 - (Maybe) Arithmetic circuits
 - Use NIOS II and cover a little bit of RISC-V

4.1.1 Mark breakdown

- Labs 15%
- project 5%
- midterm 30%
- Final 50%
- All exams will be open notes/book/whatever except another person/service helping you.

Section 5

Preliminary

Subsection 5.1

Lecture 2: Using binary quantities to represent other things

Computers can represent information in bits; 0/1. Though they don't necessarily know or care what bits are, we may assign our own arbitrary meaning to them – usually numbers with the help of positioning; the LSB represents 2^0 and so forth.

C types

• int: 32b (word)

· char: 8b (byte)

• short: 16b (half word)

• long: 32b (word)

· long long: 64b

Signed numbers may be represented in a number of ways.

Or, just #include <stdint.h>...

- Sign bit (make MSB represent positive or negative numbers and then the remaining n-1bits represent the number. Con: hardware impl sucks because requires if/else
- Two's complement ¹. Pro: only need to implement adders on hardware and then negative numbers will work just like any other except must be interpreted differently. Positive numbers would always start with a 0 and negatives would start with 1. So the range of possible values becomes $-(2^{n-1}-1)$, $+2^{n-1}-1$

¹ Flip bits, add one. Intuition; in 3 bit system, adding 7 to 1 would result in 8 which would get truncated to 0.

Adding together binary numbers can also cause overflow; $(A + B) \ge A$, $(A + B) \ge B$ may not always be true. Also, when we work with these types we always use all the bits. This has implications when working with values of different lengths.

- char b = -1(111111111)
- short int c = -1 (0000 0000 0000 0001))
- a = b + c0000 0001 0000 0000
- In order to deal with this we must cast the char to a short int. This is done via sign extension which prepends 0s or 1s ² to the char so that math can be done on it.

² two's complement

5.1.1 Floating Point Numbers

Whereas fixed point numbers i.e. \$5.25 can be represented just as how an integer would be represented but with the understanding that the user would interpret it as having a decimal point somewhere that indicates the position of 2^0 . This decimal point would be the same for all numbers of that type, i.e. we could have a six bit number that has places $2^22^12^02^{-1}2^{-2}$ This is common in embedded systems and how it is formatted isn't super clearly standardized.

Lemma 1 | Reference: What Every Computer Scientist Should Know About Floats

Definition 9

IEEE 754 Floating Point

This is a single precision 32 bit float

The most significant S bit is the sign bit, bits 30 through 23 E form the exponent which is an unsigned integer, and 22 through 0 form the (M)antissa. The number being represented can be found using the following:

$$(-1)^S \times 2^{(E-127)} \times 1$$
.Mantissa (5.2)

Example

For example, given the following float:

 $1 \quad 10000001 \quad 10000000000000000000000$

So S = 1, E = 10000001 = 129 and Mantissa = 1000000000000000000000. The number is therefore

IEEE754 also defines 64 bit floating-point numbers. They behave the same except for now having an 11 bit exponent, the bias being 2047³, and the mantissa having 52 bits. A few special cases are also available to represent other quantities

- If E=0, M non-zero, value= $(-1)^S \times 2^{-126} \times 0.M$ (denormals)
- If E=0, M zero and S=1, value=-0
- If E=0, M zero and S=0, value=0

fins@adfolo@at is a 32 bit float and double is 64

NIOS II 10

- If E=1...1, M non-zero, value=NaN 'not a number'
- If E=1...1, M zero and S=1, value=-infinity
- If E=1...1, M zero and S=0, value=infinity

Floating-point numbers are inherently imprecise. Addition and subtract are inherently lossy; the mantissa window may not be large enough to capture the decimal points. Multiplication and division just creates a ton of numbers.

Converting real numbers to IEEE754 floats, here using 37.64 as an example, can be done as follows

- Repeatedly divide the part of the number > 0 by 2 and get the remainders, i.e. 37/2 = 18, rem = 1 -> 18/2 = 0, rem = 0 -> 4/2 = 2, rem = 0, 2/2 = 1, rem = 0, 1/2 = 0, rem = 1; E = 100101
- Do the same for the part of the number past the decimal, but multiplying by two and checking if > 1: 0.64 * 2 = 1.28 -> 1, 0.28 * 2 = 0.56 -> 0, 0.56 * 2 = 1.12 -> 1 ... and so forth. At some point we will hit a cycle but we'll just take the $N_{\rm mantissa}$ of digits.

So the full number is 01000010000101101000111101011111

Section 6

NIOS II

Subsection 6.1

Lecture 3: Behavioural Model of Memory

Computers can be described as a set of units, each of which interact with each other and the outside world in a specified way. For example, modern computers tend to have memory units, processing units, display units, and so forth. Each unit has a set of inputs and outputs, and a set of rules that govern how the unit behaves. This gives the manufacturer flexibility in how they want to implement a unit, as long as the unit behaves as specified. When designing these operational units it is important to strike a balance between functionality and specificity; if the unit is too specific it will be difficult to implement, but if it is too general it will be difficult to use.

6.1.1 Memory

Memory is a unit that stores information and is usually represented as a vector of elements, usually a byte (8 bites). Each element, or memory location, contains a binary quantity and has an associated *address*. The address is a number that uniquely identifies the location of the element in the vector, and is **permanently fixed** at time of manufacture. Most systems are byte-addressable, meaning that there is an unique address for each byte in the memory. The collection of all addresses is called the **address space** of the memory, which is typically a power of two. Modern systems tend to be 32 or 64 bit, meaning that the address space is 2^{32} or 2^{64} elements long.

For each memory location there there are two operations available

- Read: Read the value stored at a given address
- Write: Write a value to a given address

Typical memory behaviour models define that the order of memory operations matters, i.e.

There are more floating point formats introduced by nvidia and google such as a halfprecision or 8-bit float designed to reduce memory use for machine learning

Specification is the description of what an unit should do, and **implementation** is how it actually does it. For example, an OR gate can be specified as a truth table and then implemented via transistors or a person in a box.

- 1. store 0x10, 0xf0
- 2. store 0x20, 0xf0

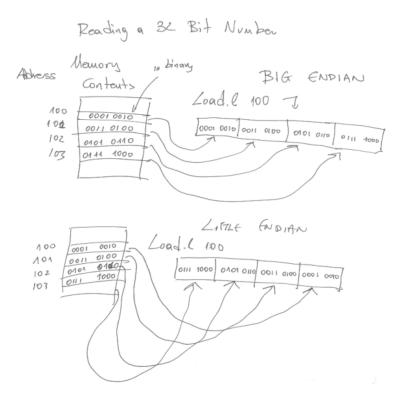
We would see that 0xf0 contains the value 0x20 and not 0x10 due to the sequential execution model. Memory that adheres to the sequential model offers operations that are **atomic**; the operations are performed on its own with no interaction or overlap with anything else.

In this case it is convenient to draw memory as an array where each row comprises four consecutive bytes.

0x00 0000	0x11	0x22	0x33	0x44
0x00 0004	0xff	0x88	0x62	0x51
0xff ffff				

Systems are generally also addressable by words, halfwords, and bytes. Different architectures have different constraints on allowing unaligned access⁴

Endianness refers to the order in which bytes are stored in memory. Though some processors are big-endian, most modern processors are little-endian. The NIOS II used for this course is little-endian.



6.1.2 Physical Interface

What physical interfaces would be necessary to implement this behavioural model? Given a summary of requirements as follows:

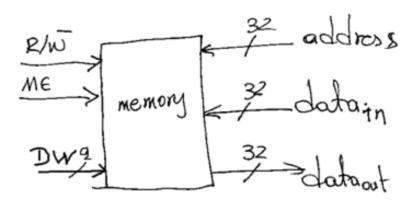
- 1. Read and write operations
- 2. Addressable by byte, word, longword
- 3. 24 bit address
- 4. 32 bit for writing

⁴ Aligned access only means to allow [only] reads or writes for a data size i.e. halfword to an address divisible by the size of said data type. For example an longword access on our development board would be at an address divisible by 4

- 5. 32 bit for reading
- 6. signal for do nothing

A single bit signal can be used to indicate whether the memory is reading or writing, and a two bit signal can be used to specify if we're interested in addressing by byte, word, or longword. The address is 24 bits, so we need 24 address lines. As for reading/writing data, we have the option of having two 32 bit data lines, or multiplexing a single 32 bit line. A single bit signal can be used to indicate to do nothing or not.

One way of multiplexing the data lines is to use a tri-state buffer, which is a buffer that can be enabled or disabled. When enabled, the buffer acts as a normal buffer, but when disabled, the output is disconnected from the input. On the other hand this means that our memory chip would not support simultaneous reads or writes.



The use of a single bit signal to indicate 'do nothing' is necessary because a physical device won't be able to change all signals instantaneously, so we use it to tell the memory to wait until these transient effects die off

Subsection 6.2

Lecture 4: NIOS II Programming Model

The NIOS II assumes a 32-bit address space where each address holds a single byte. Each byte is addressable, and three data types are supported. Halfword and word accesses must be aligned.

• Byte: 8 bits

• Halfword: 16 bits

• Word: 32 bits

The NIOS II also has a set of registers

- 32 general purpose 32 bit registers
 - r0 is always zero
- · 6 control registers, 32 bits each
- Program counter (PC), 32 bits

Many operations can be synthesized using another operation involving zero, i.e. assignment A=B can be implemented as A = B + 0

There are certain conventions for the use of registers, which are as follows:

Register	Name	Function	Register	Name	Function
r0	zero	0x00000000	r16		
r1	at	Assembler Temporary	r17		
r2		Return Value	r18		
r3		Return Value	r19		
r4		Register Arguments	r20		
r5		Register Arguments	r21		
r6		Register Arguments	r22		
r7		Register Arguments	r23		
r8		Caller-Saved Register	r24	et	Exception Temporary
r9		Caller-Saved Register	r25	bt	Breakpoint Temporary (1)
r10		Caller-Saved Register	r26	gp	Global Pointer
r11		Caller-Saved Register	r27	sp	Stack Pointer
r12		Caller-Saved Register	r28	fp	Frame Pointer
r13		Caller-Saved Register	r29	ea	Exception Return Address
r14		Caller-Saved Register	r30	ba	Breakpoint Return Address (1)
r15		Caller-Saved Register	r31	ra	Return Address

6.2.1 Adding Two Numbers

As an exercise, let's see how we can implement the following piece of code in NIOS II assembly

```
unsigned int a = 0 \times 00000000000;
unsigned int b = 0 \times 000000001;
unsigned int c = 0 \times 0000000002;
a = b + c;
```

Register-only version

```
addi r9, r0, 0x1
addi r10, r0, 0x2
add r9, r10, r11
```

addi stands for 'add intermediate', the only difference being that the second operand is a number. It is used to set a constant

In general, most instructions take the form of operation destination, source1, source2.

Breaking it down even further we can see that these assembly instructions actually perform a number of steps

```
addi r9, r0, 0x1
        ; 1. read r0
        ; 2. Add value read in step 1 with 0x1
        ; 3. Write result of step 2 to r9
        ; 4. increment PC to next instruction
addi r10, r0, 0x2
       ; 1. read r0
        ; 2. Add value read in step 1 with 0x2
        ; 3. Write result of step 2 to r10
        ; 4. increment PC to next instruction
add r9, r10, r11
        ; 1. read r10
        ; 2. read r11
        ; 3. Add values read in steps 1 and 2
       ; 4. Write result of step 3 to r8
       ; 5. increment PC to next instruction
```

What about 32 bit constants? An unfortunate quirk is that addi only supports 16 bit constants, so we need to use ori to set the upper 16 bits of the register.

```
movhi r9, 0x1122
        ; Sets the upper 16 bits of r9 to 0x1122
        ; and the lower 16 bits to zero
ori r9, r9, 0x3344
        ; bitwise OR the value in r9 with 0x3344
        ; which will set the lower 16 bits to 0x3344
```

This is a PITA so NIOS II offers a few pseudo-instructions to make this easier

```
movi rX, Imm16
       ; sets rX to the sign-extended (signed) 16 bit immediate
movui rX, Imm16
       ; sets rX to a zero-extended unsigned 16 bit immediate
movia rX, Imm32
        ; sets rX to a 32 bit immediate
```

[•] Footnote1: movia does not use the movhi and ori instructions to create a 32-bit immediate but rather a movhi and a addia. addi will sign extend it's 16-bit field so some adjustment might be needed for whatever is being passed to movhi.

[•] Footnote2: movhi r9, %hi(0x11223344) is equivalent to movhi r9, 0x1122. Ori r9, %lo(0x11223344) is equivalent to ori 9, 0x3344. That is, %hi(Imm32) returns the upper 16-bits of Imm32 and %lo(Imm32) the lower 16 bits.

[•] Footnote3: movhi r9, %hiadj(0x11223344) followed by addi r1, %lo(0x11223344) is the correct way of creating a 32-bit immediate using movhi and addi. %hiadj(Imm32) returns the upper 16 bits of the immediate as-is or incremented by 1 if bit 15 is 1. Think why this is necessary based on footnote 1.

[•] Footnote4: %hi(), %lo(), and %hiadj() are macros supported by the assembler. They are not NIOS II instructions. They get parsed during compile time.

6.2.2 Adding two numbers using memory

NIOS II is a load/store architecture which means that all data manipulation happens only in registers.

```
; read b from memory into r9
movhi r11, 0x0020
ori
     r11, r11, 0x0004
ldw
      r9, 0x0(r11)
; read c from memory into r10
movhi r11, 0x0020
     r11, 0x0008
ori
ldw
     r10, 0x0(r11)
; add, then store into r8
add r8, r9, r10
; store r8 into memory
movhi r11, 0x0020
      r11, r11, 0x0000
ori
stw
      r8, 0x0(r11)
```

The new instructions introduced here are

```
ldw rX, Imm16(rY) ;; 'load word' from memory
;; rX, rY registers, Imm16 is a 16 bit immediate
;; TLDR; Rx = mem[rY + sign-extended(Imm16)]
; 1. read rY
; 2. sign-extend Imm16 to 32bits
; 3. adds the result of step 1 and 2
; 4. reads from memory a word (32 bit) using the result of step 3 as the address
; 5. write the result of step 4 to rX
stw rX, Imm16(rY) ;; 'store word' to memory
;; rX, rY registers, Imm16 is a 16 bit immediate
;; TLDR; mem[rY + sign-extended(Imm16)] = rX
; 1. read rY
; 2. sign-extend Imm16 to 32bits
; 3. adds the result of step 1 and 2
; 4. write to memory rX using the result of step 3 as the address
```

This can be simplified using the movia macro

```
movia r11, 0x200004
      r9, 0x0(r11)
movia r11, 0x200008
ldw
      r10, 0x0(r11)
add
      r8, r9, r10
movia r11, 0x200000
      r8, 0x0(r11)
stw
```

In this lecture so far we have seen three addressing modes

- 1. Register addressing, i.e. rX
- 2. Immediate addressing, i.e. Imm16
- 3. Register indirect addressing with displacement, i.e. Imm16(rY). This is how we calculate the referenced memory address. Register indirect refers to using a register's value to refer to memory, and 'displacement' refers to adding a constant prior to using the register value to access memory. Register indirect addressing is where we use a displacement of 0.

We can exploit register indirect addressing with displacement.

```
movhi r11, 0x0020
ori
      r11, r11, 0x0004
      r9, 0x0(r11)
ldw
;; can be replaced with
movhi r11, 0x0020
ldw r9, 0x4(r11)
```

Note that the value of r11 does not change since the subsequent operations use an offset to that value.

Generally when we want to read memory from A we can use

```
movhi r11, (upper 16 bits of A)
      r9, r11, (lower 16 bits of A)
```

Care must be taken when the 16th bit of A is 1 since the addition that ldw performs will sign extend it to be a negative number, i.e.

```
movhi r11, 0x0020
ldw r9, 0x8000(r11)
;; this is incorrect because
;; will extend to 0xFFFF8000, which would result
;; in a final address of 0x001F800
```

This is where the macros %hiadj (Imm32) and %lo(Imm32) come in handy, since they will add 1 to the values if bit 15 of Imm32 is 1. This results in code that looks like this:

```
movhi r11, %hiadj(0x208000)
ldw r9, %lo(0x2080000)(r11)
;; will extend to 0xFFFF8000, which would result
;; in a final address of 0x001F800
```

PART

Ш

Taught by Prof. Sunila Akbar

ECE355: Signal Analysis and Communication

Section 7

Admin and Preliminary

Subsection 7.1

Lecture 1

- CT and DT signals
- A ton of LTI (Linear time invariant) systems
- Processing of signals via LTI systems
- · Fourier transforms
- Sampling

7.1.1 Mark Breakdown

Table 1. Mark Breakdown

·	abic 1. Mark Dicakuow				
	Homework	20			
	MT1	20			
	MT2	20			
	Final	40			

- Continuous enclose in (), independent is t
- Discrete: enclose in [], independent is n

Theorem 1

Energy for Complex Signals

$$E_{[t_1,t_2]} = \int_{t_1}^{t_2} |x(t)|^2 dt \tag{7.1}$$

$$E_{[t_1,t_2]} = \sum_{n=n_1}^{n_2} |x(t)|^2 dt$$
 (7.2)

Average Power for Complex Signals

In many systems we are interested in power and energy of signals over an infinite time interval; $-\infty < \{t, n\} < \infty$

Transformations 18

$$P_{avg,[t_1,t_2]} = \frac{1}{t_2 - t_1} \int_{t_1}^{t_2} |x(t)|^2 dt$$
 (7.3)

$$P_{avg,[t_1,t_2]} = \frac{1}{n_2 - n_1 + 1} \sum_{n=n_1}^{n_2} |x(n)|^2$$
(7.4)

SECTION 8

Transformations

Subsection 8.1

Lecture 2

Most of this lecture was review. When applying transforms just note to always scale, *then* shift, i.e.

1.
$$y(t) = x(\alpha t)$$

2.
$$y(t) = x(\alpha t + \frac{\beta}{\alpha})$$

Definition 10

Fundamental Period

$$x_t = x(t + mT), m \in \mathbb{Z} \tag{8.1}$$

The fundamental period, T_o is the smallest positive value of T for which this holds true

Definition 11

Even signals

$$x(t) = x(-t) \tag{8.2}$$

Definition 12

Odd signals

$$x(t) = -x(t) \tag{8.3}$$

Theorem 2

Any signal can be broken into an even and odd component

$$x(t) = Ev \{x(t)\} = \frac{1}{2} [x(t) + x(-t)]$$

$$x(t) = Od \{x(t)\} = \frac{1}{2} [x(t) + -x(-t)]$$
(8.4)

Subsection 8.2

Lecture 3

Again, most of this lecture was review from the waves portion of PHY293 from last year or some other course prior.

A complex exponential and sinusoidal system can be represented as

$$x(t) = Ce^{at} (8.5)$$

Where C, a are complex numbers.

Two cases may occur.

Transformations Lecture 3 19

If a imaginary and C is real we have, depending on ω , either a constant signal or a periodic sinusoidal system.

$$x(t_{=}e^{j\omega_0 t}) \tag{8.6}$$

- Important property: this is periodic, i.e. $Ce^{j\theta_0t}=Ce^{j\theta(t+T)}$
- Implies that $e^{j\omega_0T}=1$
- Implies that for $\omega \neq 0 \rightarrow T_0 = \frac{2\overline{n}}{|\omega_0|}$

On the other hand, if a imaginary and C complex, we have a periodic signal with $T=\frac{2\overline{n}}{\omega_0}$

$$x(t) = Ce^{j\omega_0 t} = |C|e^{j\omega_0 t + \phi} = |C|\cos(\omega_0 + \phi) + j|C|\sin(\omega_0 + \phi)$$
(8.7)

The energy of the signal is given by (7.1), or

$$E_{period} = \int_0^{T_0} |e^{j\omega_0 t}|^2 dt = \int_0^{T_0} 1 dt = T_0$$
 (8.8)

$$P_{period} = \frac{E_{period}}{T_0} = 1 (8.9)$$

Recall: implication that $e^{j\omega_0T}=1$, therefore the quantity inside the integral evaluates to 1

8.2.1 General Continuous Complex Exponential Signals

The most general case of a complex exponential can be represented as a combination of the real exponential and the periodic complex exponential;

$$C = Ce^{at} (8.10)$$

and

$$a = r + j\omega_0 \tag{8.11}$$

can be combined to give

Definition 13

$$Ce^{at} = |C|e^{rt}e^{j\omega_0 t + \theta} \tag{8.12}$$

Euler's relation can be used to simplify this to

$$Ce^{at} = |C|e^{rt}\left(\cos(\omega_0 t + \theta) + j\sin(\omega_0 t + \theta)\right) \tag{8.13}$$

By inspection we can see that the signal has the following properties:

- 1. r = 0: real and imaginary parts of sinusoidal
- 2. r > 0: sinusoidal signal with exponential growth
- 3. r < 0: sinusoidal signal with exponential decay

I will be skipping notes on the discrete case as it is essentially the same as the continuous case, but with the following differences

Table 2. Comparison of continuous and discrete complex exponential signals			
$e^{j\omega_0 t}$	$e^{j\omega_0 n}$		
Distinct signals for distinct ω_0	identical signals for distinct $\omega_0 \in$		
	$\{\omega_0 \pm 2\pi i, i \in \mathbb{Z}\}$		
Periodic for any ω_0	Periodic only if $\omega_0 = \frac{2\pi m}{N}$ for inte-		
	gers $N > 0, m$		
Fundamental frequency ω_0	Fundamental frequency $\frac{\omega_0}{m}$		
Fundamental period $\omega_0 = 0 \rightarrow \text{unde-}$	Fundamental period $\omega_0 = 0 \rightarrow \text{unde-}$		
fined, otherwise $T_0 = \frac{2\pi}{\omega_0}$	fined, otherwise $T_0=m rac{2\pi}{\omega_0}$		
	Since unique ω does not mean unique		
	signal, pick $0 \le \omega_0 \le 2\pi$ or $-\pi \le$		
	$\omega_0 < \pi$		

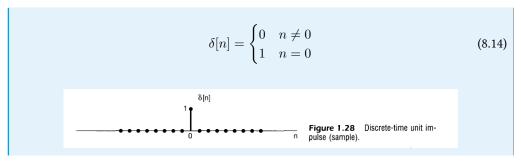
Subsection 8.3

Lecture 4: Step and Impulse Functions

One of the simplest discrete-time signals is the **unit impulse**⁵ function, $\delta[n]$

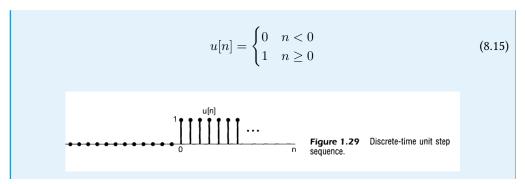
⁵ or unit sample

Definition 14



Another basic signal is the **unit step** function, u[n]

Definition 15



The unit impulse function is the first difference of the discrete time step function, i.e.

$$\delta[n] = u[n] - u[n-1] \tag{8.16}$$

And the unit step function is the running sum of the unit impulse function, i.e.

$$u[n] = \sum_{m = -\infty}^{n} \delta[m] \tag{8.17}$$

This can be rewritten with k = n - m to make a more convenient expression for moving the function along $-\infty \dots 0 \dots \infty$

$$u[n] = \sum_{k=0}^{\infty} \delta[n-k] \tag{8.18}$$

Theorem 3

The unit impulse function $\delta[n-n_0]$ can be used to sample a function at a specific $n=n_0$ since the impulse function will take on the value 0 for all values of $n \neq n_0$

$$x[n]\delta[n - n_0] = x[n_0]\delta[n - n_0]$$
(8.19)

The continuous equivalents of the unit impulse and unit step functions are defined similarly

Definition 16

$$u(t) = \begin{cases} 0 & t < 0 \\ 1 & t > 0 \end{cases}$$
 (8.20)

Likewise, the continuous unit step function is a running sum integral of the continuous unit impulse function

$$u(t) = \int_{-\infty}^{t} \delta(\tau)d\tau \tag{8.21}$$

A relationship analogous to the discrete case can be found for the continuous case; the continuous unit impulse function can be thought of as the first derivative of the continuoustime unit step function

Definition 17

$$\delta(t) = \frac{du(t)}{dt} \tag{8.22}$$

(8.22) is discontinuous at t=0 so it is non-differentiable. We can address this by considering an approximation of (8.22) for a Δ short enough to not matter for any practical purpose

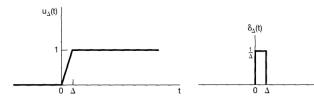


Figure 1.33 Continuous approximation to the unit step, $u_{\Delta}(t)$.

Figure 1.34 Derivative of

(8.21) can be rewritten as follows to make it more convenient to use along $\sigma \in$ $-\infty \dots 0 \dots \infty$.

$$u(t) = \int_{0}^{\infty} \delta(t - \sigma) d\sigma \tag{8.23}$$

Theorem 4

And by the same argument as for the discrete case, the continuous impulse function has an important sampling property.

For any arbitrary point t_0 ,

$$x(t)\delta t(t-t_0) = x(t_0)\delta(t-t_0)$$
 (8.24)

ECE360: Electronics

IV

Khoman

Taught by Prof.

Phang

PART

Section 9

Admin and Preliminary

Subsection 9.1

Lecture 1

9.1.1 Mark Breakdown

Table 3. Mark Breakdown

Test 1	15
Test 2	20
Homework	10
Labs	12
Final	43

9.1.2 Diodes

Diodes are an electronic valve which causes current to only flow in one direction. An ideal diode is an open circuit in the closed direction and a closed circuit in the other, so the current is always in the direction of the arrow (+'ve @ arrow base, -'ve at arrow point) 6 .

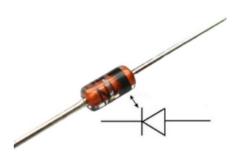


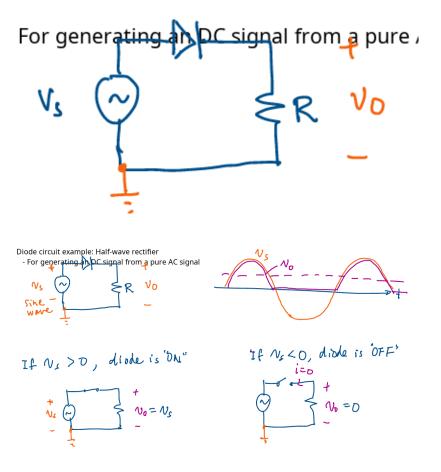
Figure 2. A diode and its symbol

An example of a diode circuit is the half-wave rectifier which turns an AC signal to a DC signal

Can take oscilloscope over resistor to see that a pure DC signal has been generated

 $^{^{6}}$ recall: passive sign convention

Diodes 23



Section 10

Diodes

Subsection 10.1

Lecture 2

More formally, off/on for diodes should be referred to as:

- Off \leftrightarrow reverse bias
- On \leftrightarrow forwards bias

DIODES Lecture 2 24

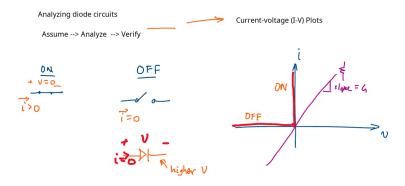


Figure 3. General steps for analyzing non-linear circuits. Note plotting out expected response

Analysis Examples					101 111
Example 1: Find output voltage Vo assuming inp What is the function of this circuit?	ut voltages	V_1 and V_2 are ϵ	either 0V or	5V.	0' "'
DN	$\underline{\vee}_{\mathbf{l}}$	<u>D</u> ,	Vz	Dz	Vo
D.	51	ON	SV	ON	SV
EV A	5	01	D	OFF	51
V Vo	0	OFF	S	ON	50
WY (+)	0	OFF	D	DFF	DV
OR gate					

An example of how this is used in circuit design is to manage two power sources. Consider an Arduino that could be powered by an AC adapter or by a computer's USB port. This circuit would choose the higher voltage source and prevent backflow into the other power source due to any potential power differentials. It is also effectively an OR gate

DIODES Lecture 2 25

In this example the initial assumption was incorrect. Let's try another analysis with D_1 off and D_2 on:

Second attempt

Step 1: Assume

$$D_1 \circ FF$$
, $D_2 \circ DN$
 SED
 SED

If we were to do this brute force we'd have to consider 4 cases, so it's important to build up some sort of intuition for the circuit.

DIODES Lecture 3 26

Subsection 10.2

Lecture 3

Today we're going to look at the characteristics of real diodes.

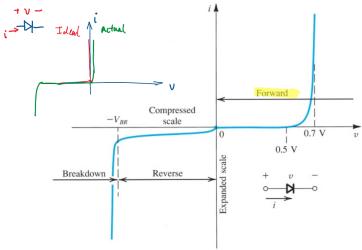


Figure 4.8 The silicon diode i-v relationship with some scales expanded and others compressed in order to reveal details.

Real diodes have a little bit of leakage current and also encounter a breakdown point where they're no longer able to block the current.

Theorem 5

Forward Bias

$$i = I_s(e^{\frac{V}{V_T} - 1}) \tag{10.1}$$

Where:

$$V_T = \frac{kT}{q} \quad [V] \tag{10.2}$$

Most of the time we can assume that the circuit is at room temperature and that $v_T=25mV$. Note that this value explodes when $V>V_T$ which is the breakdown point. When encountering a reverse bias $V_s<0$, the -1 term comes in and causes $i\approx I_s$. The scale current is just a general constant which varies in range from $10^{-9}to10^{-15}A$ and scales with temperature, doubling with every approximately 5^oC increase in temperature. Note: the ideal diode equation can be rearranged to find an expression for voltages

$$V = V_T \ln\left(\frac{i}{I_s}\right) = \ln(10)V_T \log_{10}\left(\frac{i}{I_s}\right)$$
 (10.3)

These expressions turns out to be quite reliable for reasonable diodes to reasonable voltages.

Using the ideal diode equation we can find the relationship between voltages and currents as they pass through the diode.

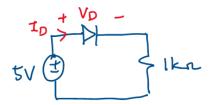
$$\frac{i_2}{i_1} = \frac{I_s e^{\frac{V_2}{V_T}}}{I_s e^{\frac{V_1}{V_T}}} = e^{\frac{V_2 - V_1}{V_T}} \tag{10.4}$$

$$V_2 - V_1 = V_T \ln\left(\frac{i_2}{i_1}\right) \xrightarrow{\text{room temperature}} 60mV \log_{10} \frac{i_2}{i_1}$$
(10.5)

k is Boltzmann's constant, T is temperature in Kelvins, q is the charge of an electron. I_s is the scale current which is usually $\approx 1pA$, which doesn't change much until the breakdown point.

Example:

Calculate the diode voltage and current in the circuit below. Assume that the diode voltage is 0.7V at 1mA and $V_T = 25mV$.



Example

Recall (10.1). Plugging in the given values gives us the scale current.

$$1mA = I_s e^{\frac{0.7V}{25mV}}, I_s = 6.9 \cdot 10^{-16} A \Rightarrow I_o = I_s e^{v_o/v_T}$$
(10.6)

Ohm's law can then be applied at the resistor

$$V_r = IR = I_o R = 5V - V_D \Rightarrow 5 - V_D = I_o R$$
 (10.7)

 V_D is the voltage across the diode

So we have two equations and two unknowns (since we know $v_T = 25mV$ but v_o was used at first just to find I_s) Solving for the unknowns gives us:

- $V_o = 0.736V$
- $I_D = 4.264mA$

Section 11

Lecture 4: Forward conducting diodes

The exponential model accurately describes the diode outside of the breakdown region, though it's nonlinear behaviour makes it difficult to use.

For
$$V_{DD} > 0.5V$$

 V_{DD} denotes a DC power supply

$$I_D = I_S e^{V_D/V_T} \tag{11.1}$$

Where

- I_S is the diode parameter
- V_T is the thermal voltage

Another equation may be produced via Kirchhoff's law

$$I_D = \frac{V_{DD} - V_D}{R} \tag{11.2}$$

The unknown quantities I_D and V_D may be solved for via graphical analysis or iteration.

Example | This simple circuit is used to demonstrate the exponential model of the diode.

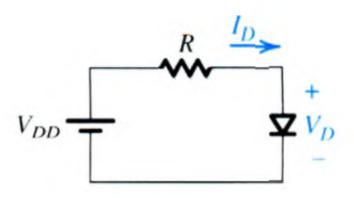
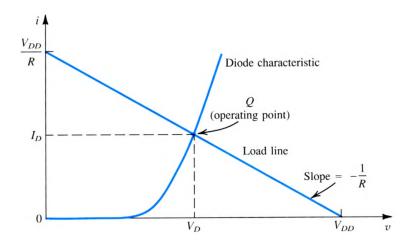


Figure 4. Simple example circuit with diode

Plots of the diode characteristics and Kirchhoff's relation are plotted, the intersection of which gives the solution.



An iterative procedure may also be applied to solve for the unknowns, the procedure for which will be illustrated through an example

Example

Find I_D,V_D for the circuit in the previous example (Fig. 4). $V_{DD}=5V,R=1k\Omega$, and at $V_D0.7V,I_D=1mA$

1. Assume $V_D=0.7V$, then use (11.2) to find I_D .

$$I_D = \frac{5V - 0.7V}{1k\Omega} = 4.3mA \tag{11.3}$$

2. Use the diode equation (11.1) to get a better estimate for V_D .

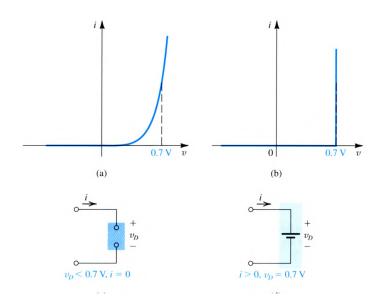
$$V_2 - V_1 = 2.3V_T \log \frac{I_2}{I_1} \Rightarrow V_2 = V_1 + 0.06 \log \frac{I_2}{I_1}$$
 (11.4)

substituting $V_1 = 0.7V, I_1 = 1mA, I_2 = 4.3mA$,

$$V_2 = 0.738V \Rightarrow I_D = 4.3mA, V_D = 0.738V$$
 (11.5)

- This states that for a decade⁷ change in current the diode voltage drop changes by $2.3(V_T \approx 60mV)$ which is negligibly small for v < 0.5V. The voltage at which the this behaviour becomes significant is called the **cut-in voltage**
- ⁷Factor of 10
- 3. Repeat steps 1 and 2 with the new values until the values more or less become stable

This iterative model is powerful and yields accurate results, but can be computationally expensive especially when calculating by hand. To address this we employ other models such as the *constant-voltage-drop* model which approximates the exponential characteristics via a piecewise linear model. The reason why this is possible is because forward conducting diodes exhibit a voltage drop that varies in a relatively narrow range.



Using the constant voltage drop model in our analysis looks the same as before, but with V_D directly taking on the value of 0.7V (as per the prior example) instead of being solved for with the diode equation.

In applications that involve voltages greater than the voltage drop (i.e. usually $\approx 0.6-0.8V$) we can neglect the diode voltage drop altogether while calculating the diode current.

$$V_D = 0V I_D = \frac{5-0}{1} = 5mA$$
 (11.6)

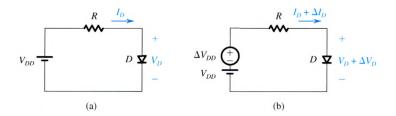
This is generally good enough for a first estimate, though the previous model isn't that much more work and gives more accurate results. The primary use of this model is to determine which diodes are on or off in a multi-diode circuit

11.0.1 Small-Signal Model

The small signal method is an alternative model used to describe the nonlinear diode's characteristics with greater accuracy than piecewise linear models.

Consider a small ΔV_{DD} applied to the diode, which would cause a small ΔI_D , ΔV_D . We want to find a quick way of determining the values of these incremental changes.

Similar methods will be applied to transistors in later chapters



Skipping a bunch of math⁸ the results are as follows:

⁸ It is 11:17pm and I have two more lectures to catch up to today

Definition 18

Small signal approximation

$$i_D(t) \approx I_D(1 + \frac{v_d}{V_T}) \tag{11.7}$$

This is valid for when variations in diode voltage $|v_d| \lesssim 5mV$ The quantity relating i_d to v_d is the diode small-signal resistance

$$r_d = \frac{V_T}{I_D} \tag{11.8}$$

The steps for calculating the small signal model are as follows:

- 1. Perform a dc analysis using the exponential, constant-voltage-drop, or piecewise-linear model.
- 2. Linearise the circuit. For a forward-based diode, find r_d by substitution I_D into (11.8). The small-signal equivalent circuit is found by eliminated all independent dc sources and replacing the diode with its small-signal resistance r_d
- 3. Solve the linearised circuit. In particular we would want to find ΔI_D , ΔV_D and check to see if it is consistent with our approximation, i.e. that $\Delta V_D \lessapprox 5mV$

Small signal analysis can be performed separately from the dc bias analysis because of the linearization of diode characteristics in the small-signal approximation since we already accounted

since we already accounted for them in step 1

ECE358: Foundations of Computing

PART

V

Section 12

Admin and Preliminary

Subsection 12.1

Lecture 1

Topics covered will include:

- · Graphs, trees
- · Bunch of sorts
- · Fancy search trees; red-black, splay, etc

Taught by Prof. Shurui Zhou

Complexities 31

- · DP, Greedy
- Min span tree, single source shortest paths
- Maximum flow
- NP Completeness, theory of computation
- · Blockchains??
- ⊖

Solutions will be posted on the window of SF2001. Walk there and take a picture.

12.1.1 Mark Breakdown

Table 4. Mark Breakdow			
Homework x 5	25		
Midterm (Open book)	35		
Final (Open book)	40		

Section 13

Complexities

Subsection 13.1

Lecture 2

This lecture we talked about big O notation. For notes on this refer to my tutorial notes for ESC180, ESC190: https://github.com/ihasdapie/teaching/

Definition 19

Big O notation (upper bound)

g(n) is an asymptotic upper bound for f(n) if:

$$O(g(n)) = \{ f(n) : \exists c, n_0 \quad s.t. \quad 0 \le f(n) \le c \cdot g(n), \forall n \ge n_o \}$$
 (13.1)

Proof

What is the big-O of n!?

$$n! \le n \cdot n \cdot n \cdot n \dots n = n^n \Rightarrow n! \in O(n^n)$$
 (13.2)

Definition 20

Big Ω notation (lower bound)

h(n) is an asymptotic lower bound for f(n) if:

$$\Omega(h(n)) = \{f(n): \exists \quad c, n_0 > 0 \quad s.t. \quad 0 \le c \cdot h(n) \le f(n), \forall n \ge n_o\}$$
 (13.3)

Proof

Find Θ for $f(n) \sum_{i=1}^{n} i$.

For this we will employ a technique for the proof where we take the right half of the function, i.e. from $\frac{n}{2} \dots n$ and then find the bound

Complexities Lecture 2 32

$$f(n) = 1 + 2 + 3 \dots + n$$

$$\geq \lceil \frac{n}{2} \rceil + (\lceil \frac{n}{2} \rceil + 1) + (\lceil \frac{n}{2} \rceil + 2) + \dots n \quad n/2 \text{ times}$$

$$\geq \lceil \frac{n}{2} \rceil + \lceil \frac{n}{2} \rceil + \lceil \frac{n}{2} \rceil + \dots \lceil \frac{n}{2} \rceil$$

$$\geq \frac{n}{2} \cdot \frac{n}{2}$$

$$= \frac{n^2}{4}$$
(13.4)

And therefore for $c=\frac{1}{4}$ and n=1 , $f(n)\in\Theta(n^2)$

Definition 21

Big Θ notation (asymptotically tight bound)

$$\Theta(g(n)) = \{ f(n) : \exists c_1 c_2, n_0 \text{ s.t. } 0 \le c_1 g(n) \le f(n) \le c_2 g(n), \forall n \ge n_o \}$$
 (13.5)

Proof

Prove that

$$\sum_{i=1}^{n} i^k = \Theta(n^{k+1}) \tag{13.6}$$

First, prove $O(f(n)) = O(n^{k+1})$

$$f(n) = \sum_{j=1}^{n} i^{k} = 1^{k} + 2^{k} + \dots n^{k}$$

$$\leq n^{k} + n^{k} + \dots n^{k}$$

$$= n^{k+1}$$
(13.7)

Next, prove $\Omega(f(n)) = \Omega(n^{k+1})$

$$f(n) = \sum_{j=1}^{n} i^{k} = 1^{k} + 2^{k} + \dots n^{k}$$

$$= n^{k} + (n_{1})^{k} + \dots 2^{k} + 1^{k} = \sum_{i=1}^{n} (n - i + 1)^{k}$$

$$\geq \frac{n^{k}}{2} * n \geq \frac{n^{k+1}}{2^{k}} = \Omega(n^{k+1})$$
(13.8)

Therefore
$$f(n) = \Theta(n^{k+1})$$

Note that we may not always find both a tight upper and lower bound so not all functions have a tight asymptotic bound.

Theorem 6

Properties of asymptotes:

Note: ∧ means AND

Transitivity 10

$$(f(n) = \Theta(g(n)) \land g(n) = \Theta(h(n))) \Rightarrow f(n) = \Theta(h(n))$$
(13.9)

¹⁰ The following applies to O, Θ, o, ω

COMPLEXITIES Lecture 3: Logs & Sums 33

Reflexivity¹¹

$$f(n) = \Theta(f(n)) \tag{13.10}$$

¹¹ The following applies to O, Θ

Symmetry

$$f(n) = \Theta(g(n)) \iff g(n) = \Theta(f(n))$$
 (13.11)

Transpose Symmetry

$$f(n) = O(g(n)) \iff g(n) = \Omega(f(n))$$

$$f(n) = o(g(n)) \iff g(n) = \omega(f(n))$$
(13.12)

Runtime complexity bounds can sometimes be used to compare functions. For example, f(n) = O(g(n)) is like $a \leq b$

- O ≈≤
- Ω ≈≥
- Θ ≈≈
- $o \approx <$; an upper bound that is **not** asymptotically tight
- ω > a lower bound that is **not** asymptotically tight

Note that there is no trichotomy; unlike real numbers where we can just do a < b, etc, we may not always be able to compare functions.

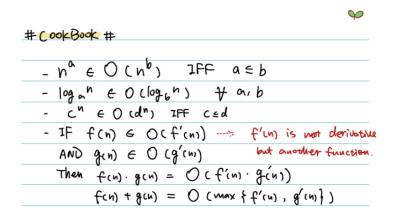


Figure 5. Complexity Cookbook

Subsection 13.2

Lecture 3: Logs & Sums

13.2.1 Functional Iteration

Recall:

 $f^{(i)}(n)$ denotes a function iteratively applied i times to value n. For example, a function may be defined as:

 $a = b^c \Leftrightarrow loq_b a = c$ (13.13)

$$f^{(i)}(n) = \begin{cases} f(n) & \text{if } i = 0\\ f(f^{(i-1)}(n)) & \text{if } i > 0 \end{cases}$$
 (13.14)

Given (13.14) we see that

- 1. f(n) = 2n
- 2. $f^{(2)}(n) = f(2n) = 2^2 n$
- 3. $f^{(3)}(n) = f(f^{(2)}(n)) = 2^3 n$:
- 4. $f^{(i)}(n) = 2^{i}n$

As an exercise we may look at an iterated logarithm function, 'log star'

$$lg^*(n) = \min\{i \ge 0 : lg^{(i)}n \le 1\}$$
(13.15)

This describes the number of times we can iterate log(n) until it gets to 1 or smaller.

- $log^*2 = 1$
- $log^*4 = 2 = log^*2^2 = 1 + log^*2 = 2$:
- for practical reasons log^* doesn't really get bigger than 5. This is one of the slowest growing functions around.

Summations & Series

Proof for a finite geometric sum: **PROOF**

$$\sum_{k=0}^{n} x^{k} = S$$

$$S = 1 + x + x^{2} \dots x^{n}$$

$$xS = x + x^{2} + x^{3} \dots x^{n+1}$$

$$S = \frac{1 - x^{n+1}}{1 - x}$$
(13.16)

$$\sum_{i=1}^{\infty} x^i = \frac{1}{1-x} \quad \text{if } |x| < 1 \tag{13.17}$$

$$\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2} \quad \text{if } |x| < 1 \tag{13.18}$$

Begin by differentiating both sides over xProof

$$\sum_{k=0}^{\infty} x^k = \frac{1}{(1-x)} \quad \text{if } |x| < 1 \tag{13.19}$$

$$\sum_{k=0}^{\infty} kx^{k-1} = \frac{1}{(1-x)^2} \tag{13.20}$$

And then multiply both sides by x, therefore (13.18) follows.

Telescoping Series

$$\sum_{k=1}^{n} a_k - a_{k-1} = a_n - a_0 \tag{13.21}$$

Proof

Write it out and cancel out terms

$$(a_1 - a_0) + (a_2 - a_1) \dots (a_n - a_{n-1}) = a_n - a_0$$
(13.22)

Therefore the sum telescopes

Another telescoping series may be proved similarly:

$$\sum_{k=1}^{n-1} \frac{1}{k(k+1)} \xrightarrow{math} \sum_{k=1}^{n-1} (\frac{1}{k} - \frac{1}{k+1}) = (1 - \frac{1}{2}) + \dots + (\frac{1}{n-1} - \frac{1}{n}) = a_o - a_n \quad (13.23)$$

Subsection 13.3

Lecture 4: Induction & Contradiction

MAT389: Complex Analysis

Section 14

Complex Numbers

Subsection 14.1

Lecture 1

Consider a 2-vector $\vec{x} = (x, y) \in \mathcal{R}$. As complex numbers correspond to two-vectors

$$\vec{x} = (x, y) \leftrightarrow z = x + iy, i^2 = -1 \tag{14.1}$$

z is, therefore, a complex variable. What are the benefits of a complex number like z?

Definition 22

Imaginary and Complex Numbers

i is an imaginary number such that

$$i^2 = -1 (14.2)$$

A complex number has the form:

$$z = x + iy \tag{14.3}$$

Definition 23

There are a number of operations we can perform on complex numbers. **Addition**

$$z + z' = (x + x') + i(y + y')$$
(14.4)

Multiplication

$$zz' = (x+iy)(x'+iy') = (xx'-yy') + i(xy'+x'y)$$
(14.5)

PART

VI

Taught by Prof. Sigil

This prof lectures at the speed of sound and talks *into* the board. Couldn't quite follow during this lecture, hopefully I get better about it in the following ones. Complex Numbers Lecture 1 36

Proof

Proof of (14.5):

$$zz' = (x + iy)(x' + iy')$$

$$= x + ixy' + iyx' + i^{2}yy'$$

$$= xx' - yy' + i(xy' + yx')$$
(14.6)

Magnitude

$$|z| = \sqrt{x^2 + y^2} \tag{14.7}$$

Conjugate

The complex conjugate has the properties:

•
$$\overline{z}z = |z|^2$$

•
$$\overline{(z+z')} = \overline{z} + \overline{z}'$$

•
$$\overline{z \cdot z'} = \overline{z} \cdot \overline{z}'$$

We can define a new operation

$$\forall \text{complex} z, \exists \text{ complementary number } w \text{ such that } zw = wz = 1$$
 (14.8)

Denote

$$w = \frac{1}{z} = z^{-1} \tag{14.9}$$

Proof

Proof of (14.9): Find w s.t. zw = 1

$$zw = 1$$

$$w\overline{z}z = \overline{z}z = |z|^2 > 0$$

$$|z|^2 w = \overline{z}$$

$$w = \frac{\overline{z}}{|z|^2} \to Z^{-1} = \frac{\overline{z}}{|z|^2}$$

$$\Box$$
(14.10)

Furthermore, there are operators that we can define on complex numbers.

Definition 24

Real and Imaginary Operators

Given z = x + iy, we can define the real and imaginary operators

$$x = Re\{z\} \tag{14.11}$$

$$y = Im\left\{z\right\} \tag{14.12}$$

Complex Numbers Lecture 2 37

Example

$$Im\left\{ (1 = \sqrt{2}i)^{-1} \right\} \tag{14.13}$$

By (14.9), we have

$$Im\left\{z^{-1}\right\} = \frac{-Im\left\{z\right\}}{|z|^2}$$
 (14.14)

And

$$Re\left\{z^{-1}\right\} = \frac{-Re\left\{z\right\}}{|z|^2}$$
 (14.15)

Using these, for example, we find that the $Im=\frac{-\sqrt{2}}{3}$ We can get the real component in a similar way.

Here is an enumeration of absolute value properties for complex numbers:

$$|z \cdot w| = |z||w| \tag{14.16}$$

$$|z + w| \le |z| + |w| \tag{14.17}$$

$$|\overline{z}| = |z| \tag{14.18}$$

$$|z+w|^2 = (\overline{x} + \overline{w})(z+w) = |z|^2 + |w|^2 + \overline{z}w + \overline{w}z$$
(14.19)

Proof

Note that $\overline{z}w+\overline{w}z=2Re\left\{ z\overline{w}\right\}$, by (14.19) And so

$$|z+w|^2 \le |z|^2 + |w|^2 + 2|z||w| = (|z|+|w|)^2$$
 (14.20)

Subsection 14.2

Lecture 2

Whereas a two-vector $\vec{x} \in \mathbb{Z}$, complex numbers exist in the complex plane, $z \in \mathbb{C}$

Theorem 7

Polar Decomposition

Complex numbers can be expressed in polar form as well

$$z = r(\cos\theta + i\sin\theta) \tag{14.21}$$

Where

$$r = |z|$$
 $x = rcos\theta$ $y = rsin\theta$ (14.22)

This has a number of useful properties

$$z \cdot z' = |z||z'|(\cos(\theta + \theta') + i\sin(\theta + \theta')) \tag{14.23}$$

$$\frac{z}{z'} = \frac{|z|}{|z'|} (\cos(\theta - \theta') + i\sin(\theta - \theta')) \tag{14.24}$$

Complex Numbers Lecture 2 38

Proof P

Proof for (14.23):

$$z \cdot z' = |z|(\cos(\theta + i\sin\theta)) \times |z'|(\cos\theta' + i\sin\theta')$$

$$= |z||z'|(\cos\theta\cos\theta' + i\cos\theta\sin\theta' + i\sin\theta\cos\theta' - \sin\theta\sin\theta')$$

$$= |z||z'|\left[\cos\theta\cos\theta' - \sin\theta\sin\theta' + i(\cos\theta\sin\theta' + \sin\theta\cos\theta')\right]$$
(14.25)

And the proof follows

Lemma 2

A corollary exists

$$z^2 = |z|^2(\cos 2\theta + i\sin 2\theta) \tag{14.26}$$

Theorem 8

Moivre's Theorem

$$z^{n} = |z|^{n}(\cos(n\theta) + i\sin(n\theta)) \tag{14.27}$$

So we may define z to be the n^{th} root of w which implies that

Lemma 3

Every complex number has a n^{th} root $\forall n$

Proof

Let
$$z = |w|^{\frac{1}{n}} \left(\cos\frac{\theta}{n} + i\sin\frac{\theta}{n}\right)$$
 (14.28)

Then

$$w = |w|(\cos\theta + i\sin\theta), \text{ then } z^n = w \tag{14.29}$$

This leads us to the conclusion that representations of complex numbers are not unique¹².

12 They are part of a cyclic group

Intuition: define z to be $\frac{1}{n}$ and

then take the n^{th} power of

both sides to show that $z^n =$

Proof

If every z can be written as $z = r(\cos \theta + i \sin \theta)$, then it holds for $\theta + 2\pi n \forall n \in \mathbb{Z}$ since $\sin \theta = \sin(\theta + 2\pi n)$ and $\cos \theta = \cos(\theta + 2\pi n)$.

14.2.1 Functions on complex planes

Definition 25

Given a domain $\mathbb{D} \in \mathbb{C}$, a function f is a rule such that

$$z \in \mathbb{D} \xrightarrow{f} w \in \mathbb{D} \leftrightarrow w = f(z) \tag{14.30}$$

Definition 26

We may define $\mathbb D$ to be the domain of f

Likewise, range is defined as

$$Ran\{f\} = \{w \in \mathbb{C} : \exists z \in D : f(z) = w\}$$

$$\tag{14.31}$$

Example

$$f(z) = \frac{1}{z+i} \tag{14.32}$$

What is the maximum domain of f?

$$Dom\{f\} = \{z \in \mathbb{C} : |z| < -i\}$$
 (14.33)

What is the range of f?

$$\frac{1}{z+i} = w \tag{14.34}$$

COMPLEX NUMBERS Lecture 2 39

For which values of w can we solve this equation?

$$z = -i + \frac{1}{w} \tag{14.35}$$

So the range of the function is

$$Ran\{f\} = \{w \in \mathbb{C} : |w| \neq 0\}$$
 (14.36)

Example

$$f(z) = z^2 + 1 (14.37)$$

It is fairly clear that $Dom\{f\} = f \in \mathbb{C}$

The range can be found by solving for z in

$$z^2 + 1 = w ag{14.38}$$

And so

$$Ran\{f\} = \{w \in \mathbb{C}\}\tag{14.39}$$

14.2.2 Exponential Functions

Definition 27

Given z = x + iy,

$$e^{z} = e^{x}(\cos y + i\sin y) = e^{Re\{z\}}(\cos(Im\{z\}) + i\sin(Im\{z\}))$$
 (14.40)

1.
$$e^{z+w} = e^z e^w$$

1.
$$e^{z+w} = e^z e^w$$

2. $|e^z| = e^{Re\{z\}} \neq 0$

$$3. e^{z+i2\pi n} = e^z$$

- (1) follows from the product rule for complex numbers
- (2) follows by definition
- (3) follows by definition (recall: writing z w.r.t. sin, cos)

More properties:

- $Dom\{e^z\} = \mathbb{C}$
- $Ran\{e^z\} = \{\mathbb{C} \setminus \{0\}\}$
- $e^z = w$ if $w \neq 0$

13

arg, or argument is the angle from the real axis to that on the complex plane. Usually denoted by θ

 13 Note: ' \ ' denotes set exclusion

$$z = ln|w| + i \arg w$$

$$e^{z} = e^{ln|w| + i \arg w}$$

$$= e^{ln|w|} e^{i \arg w}$$

$$= |w| \cos(\arg w) + i \sin(\arg w)$$

$$= w$$
(14.41)

Polar representation Remark

$$w = |w|e^{i\arg w} \tag{14.42}$$

Complex Numbers Lecture 2 40

Example

Find polar coordinates of z = i + 1

 $r = |w| \quad \theta = \arg w \quad (14.43)$

$$|z| = \sqrt{1+i} = \sqrt{2}$$

$$\cos \theta = \frac{1}{\sqrt{2}} \to \theta = \frac{\pi}{4}$$

$$z = \sqrt{2}e^{i\pi/4}$$
(14.44)

Example

Find

$$(1+i)^{\frac{1}{3}} \tag{14.45}$$

Solution: $z = \sqrt{2}e^{\frac{i\pi}{4}} \to z^{1/3} = 2^{\frac{1}{6}}e^{i\pi/12}$

Definition 28

Trig functions for complex numbers

$$\cos z = \frac{1}{2} \left(e^{iz} + e^{-iz} \right) \tag{14.46}$$

Proof

$$\cos x = \frac{1}{2}(e^{ix} + e^{-ix}) = \frac{1}{2} \left(\cos x + i \sin x + \underbrace{\cos(-x)}_{\text{odd}; = \cos(x)} + \underbrace{i \sin(-x)}_{\text{even}; = -\sin(x)} \right) = \cos x$$
(14.47)

$$\sin z = \frac{1}{2} \left(e^{iz} - e^{-iz} \right) \tag{14.48}$$

And a similar proof follows for $\sin z$.

These have the following properties

$$\sin z|_{ImZ=0} = \sin x \tag{14.49}$$

$$\cos(z + 2\pi n) = \cos z \forall n \in \mathbb{Z}$$
(14.50)

$$\sin(z + 2\pi n) = \sin z \forall n \in \mathbb{Z} \tag{14.51}$$

Proof

$$\cos z + 2\pi n = \frac{1}{2} (e^{i(z+2\pi n)} + e^{-i(z+2\pi n)})$$

$$= \frac{1}{2} (e^{iz} e^{i2\pi n} + e^{-iz} e^{-i2\pi n})$$

$$= \frac{1}{2} (e^{iz} + e^{-iz})$$

$$= \cos z$$
(14.52)

The domain of $\{\cos z, \sin z\} = \mathbb{C}$

Range?

Solve $\cos z = w$ for z

$$\frac{1}{2}(e^{iz}+e^{-iz})=w$$
 ... \times $2e^{iz}$ on both sides
$$e^{2iz}-2we^{iz}+1=0$$
 ... Let $S=e^{iz}$
$$S^2-2ws+1=0$$

$$S=w\pm\sqrt{w^2-1}\equiv u$$
 (14.53)

Now we note that $e^{iz}=u$ can be solved for z for any $u\neq 0$

$$u = 0 \leftrightarrow w = \pm \sqrt{w^2 - 1} \tag{14.54}$$

$$w^2 = w^2 - 1 \text{ impossible for } u \neq 0 \tag{14.55}$$

Therefore:

$$Ran\{\cos z\} = Ran\{\sin z\} = \mathbb{C}$$
(14.56)

Remark

An intuitive way of interpreting this result is thinking of $\{\sin,\cos\}$ being a function that projects values from the complex domain to the real plane; though $\{\sin,\cos\}$ takes on a limited range of values in the real domain, in the complex domain it spans the entire plane. Think: mental model of a complex number spinning around and having that project onto a real line. More formally, see: the Little Picard Theorem

ECE444: Software Engineering

SECTION 15

Preliminary

Subsection 15.1

Lecture 1, 2

- Software engineering is different from what coding is; design, architecture, documentation, testing, etc v.s. just script kiddie-ing
- Vasa syndrome
- · Rockstar engineers are a myth

Section 16

Project Management

Subsection 16.1

Lecture 3

PART



Taught by Prof. Shurui Zhou

Project Management Lecture 3 42

Definition 29

Conway's law states that 'Any organization that designs a system (defined broadly) will produce a design whose structure is a copy of the organization's communication structure'.

The waterfall method is slow and costly and defects can be extremely costly, especially early on in the development lifecycle.

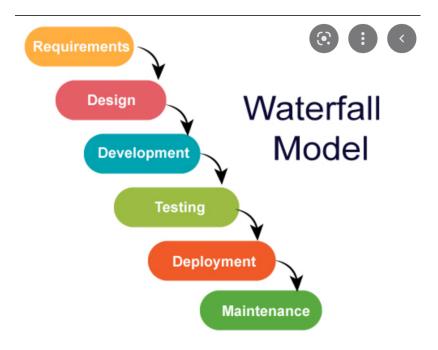


Figure 6. Waterfall method

In order to address this the V model was introduced which increases the amount of testing to reduce the possibility of having to rework everything

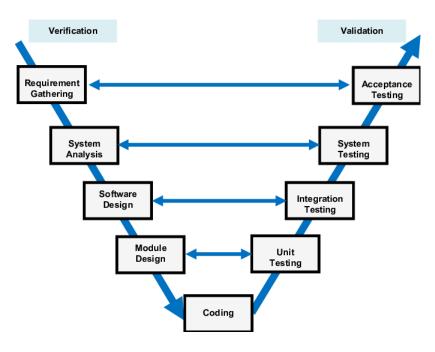


Figure 7. V model

Generally speaking the waterfall model isn't used much anymore due to the reality that software specifications change on a near daily basis.

16.1.1 Agile

Agile is a project management approach which, in most general terms, seeks to respond to change and unpredictability using incremental, iterative work (sprints). This allows for a balance between the need for predictability and the need for flexibility. Some agile methods include:

- Extreme programming: really really fast iteration (think days)
- Scrum: 2-4 week sprints with standups and backlogs; sticky notes for tasks, etc. Think kanban boards. Daily scrum meetings to unblock ASAP. Development lifecycle is therefore a series of sprints.
- On-site customer; frequent interaction with end users to figure out what exactly they need.

Subsection 16.2

I dropped this course

I decided to drop this course because the courseload was a little too much to handle between EngSci ECE, clubs, design teams, work, and trying to have a life.

ESC301: Seminar

Section 17

Preliminary

Recall: aUToronto Spring 2022 integration hell



Preliminary 44

Subsection 17.1

Seminar 1