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# **Lattice Coding and its Applications in Communications**

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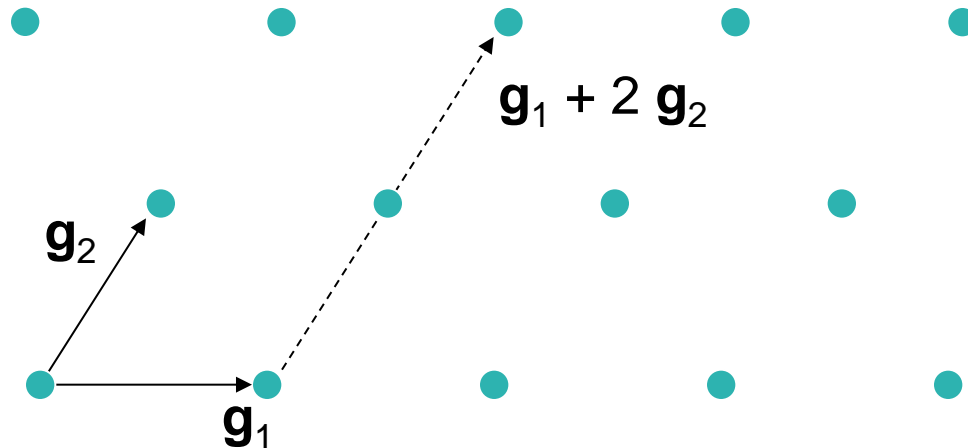
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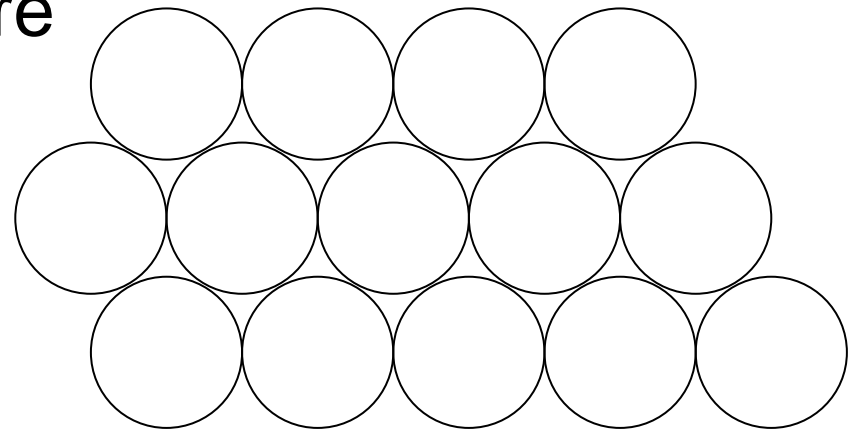
- Introduction to lattices
  - Definition; Sphere packings; Basis vectors; Matrix description
- Codes and lattice codes
  - Shaping region; Nested lattices
- Lattice constructions
  - Construction A/D, LDLC codes; construction from Gaussian/Eisenstein integers
- Lattice encoding and decoding
  - Problems of shaping; LDLC decoding; Construction A decoding
- Lattices in multi-user networks: Compute and forward

- A **lattice** is defined as:
  - the (infinite) set of points in an  $n$ -dimensional space given by all linear combinations with integer coefficients of a **basis** set of up to  $n$  linearly independent vectors
- It can be defined in terms of a **generator matrix**  $\mathbf{G}$ , whose columns are the basis vectors:

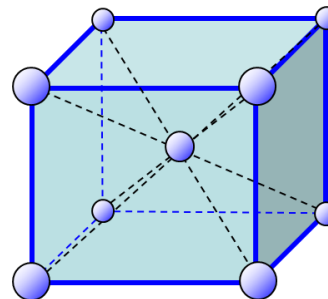
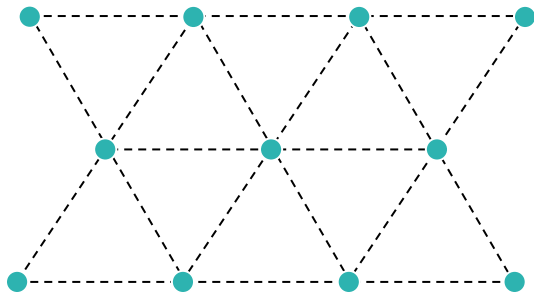
$$\Lambda = \{ \lambda = \mathbf{G} \mathbf{x} : \mathbf{x} \in \mathbb{Z}^n \}$$



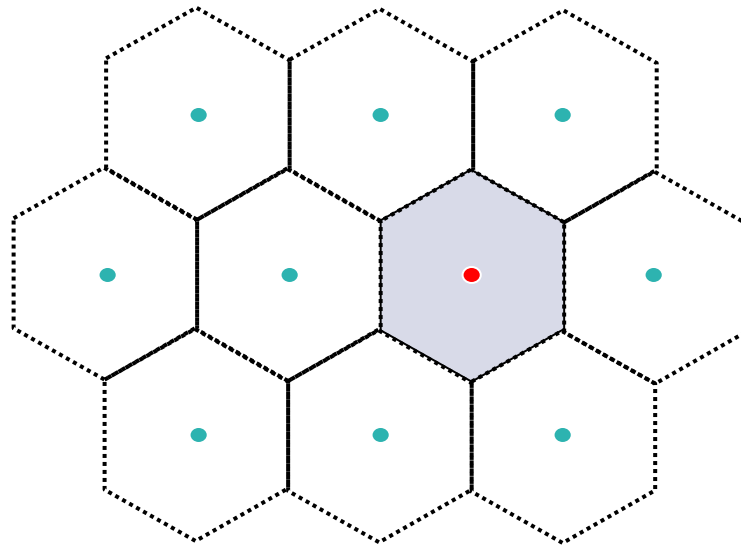
- A ***sphere packing*** is an arrangement of non-overlapping ***hyperspheres*** of equal radius in  $N$ -dimensional space
- We are often interested in the ***packing density***  $\eta$  or  $\delta_n$  of a packing
  - the proportion of space occupied by spheres
- Dense sphere packings are often ***lattice packings***
  - have sphere centres on lattices



Dimensions	Lattice	Packing density	Kissing number
2	Hexagonal	$\frac{1}{6}\pi\sqrt{3}=0.91$	6
3	BCC/FCC/HCP	$\frac{1}{6}\pi\sqrt{2}=0.74$	12
4	D4	$\frac{1}{16}\pi^2 = 0.62$	24
8	E8	$\frac{1}{384}\pi^4 = 0.25$	240
24	E24 (Leech)	$\frac{\pi^{12}}{12!} = 0.0019$	196 560



- The **Voronoi region** of a lattice point is the region of the  $N$ -dimensional space closer to that point than to all other lattice points
- Voronoi region of red point shown shaded



- Introduction to lattices
- **Codes and lattice codes**
  - **Shaping region**
  - **Nested lattices**
- Lattice constructions
- Lattice encoding and decoding
- Lattices in multi-user networks: Compute and forward

- i.e. **forward error-correcting** (FEC) codes
- A **code** is a finite set of **codewords** of length  $n$ 
  - Code contains  $M$  codewords – encodes  $\log_2(M)$  bits
- where a codeword is a sequence of  $n$  **symbols**, usually drawn from a finite **alphabet** of size  $q$ 
  - we will often assume the alphabet is a Galois field ( $\mathbb{F}_q$  or  $\text{GF}(q)$ ) or a ring ( $\mathcal{R}(q)$ )
- In a communication system the codewords must be translated into **signals** of length  $nT$ 
  - representing the variation in time of some quantity, such as electromagnetic field strength
- Each code symbol is typically **modulated** to some specific real or complex value of this variable



Message:

01111001

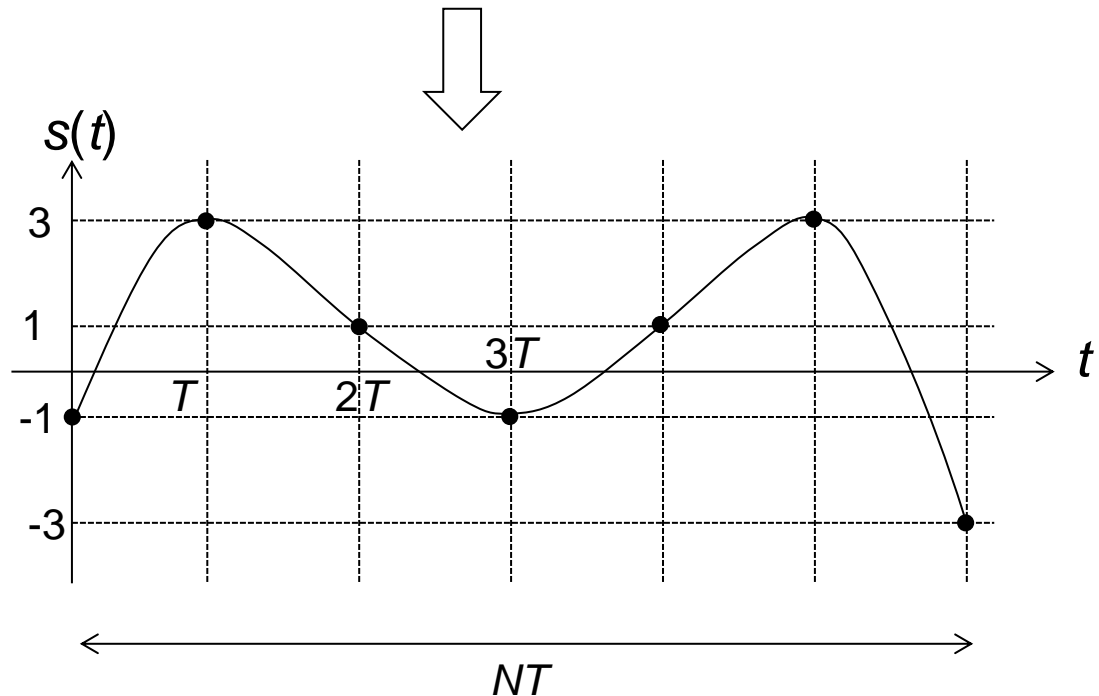
*Encode*

Codeword:

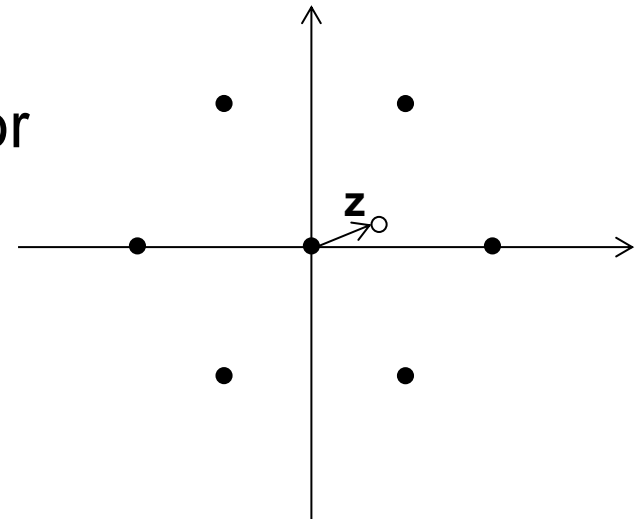
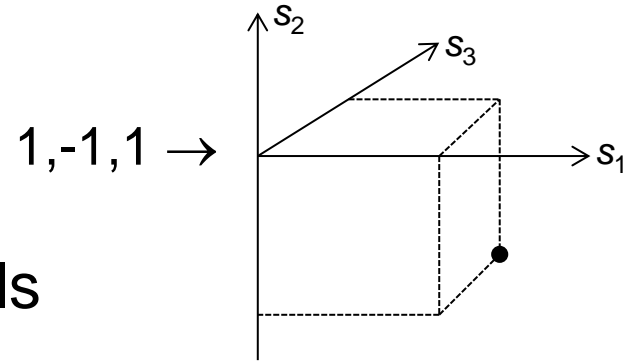
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*Modulate*

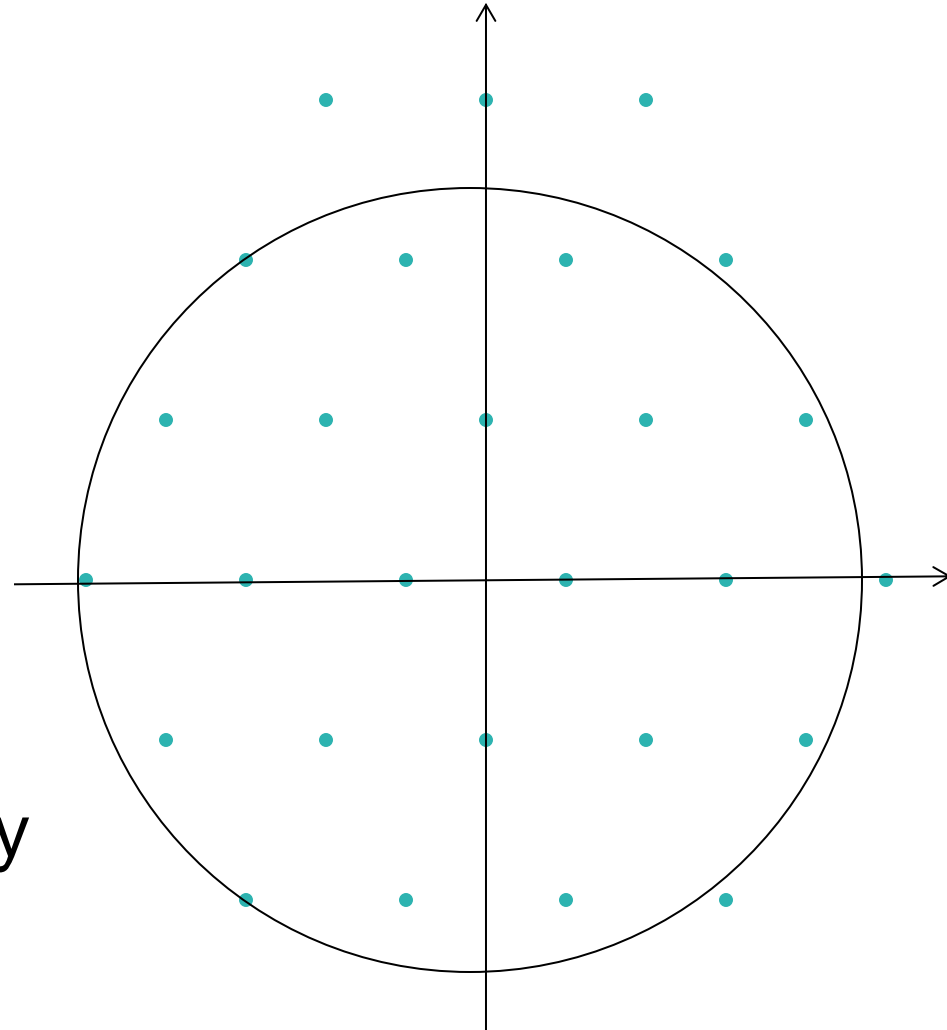
Signal:



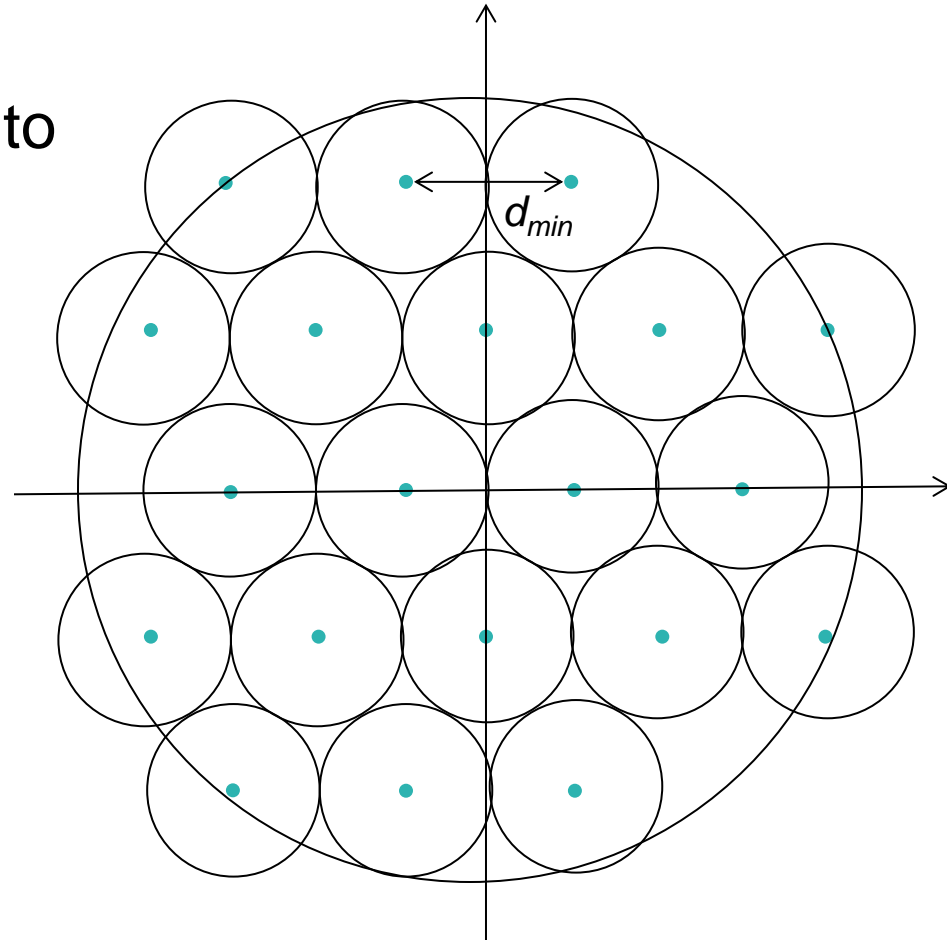
- Each coded signal can then be represented as a point in  $N$ -D **signal space**
  - where modulated values of symbols provide the  $n$  coordinate values
- Code is represented by ensemble of points in signal space
- Noise on channel equivalent to vector  $z$  in signal space
- Decoder chooses closest point
- Error probability determined by **minimum Euclidean distance** between signal space points



- A **lattice code** is then defined by the (finite) set of lattice points within a certain region
  - the **shaping region**
  - ideally a hypersphere centred on the origin
  - this limits the maximum signal energy of the codewords
- Lattice may be offset by adding some vector



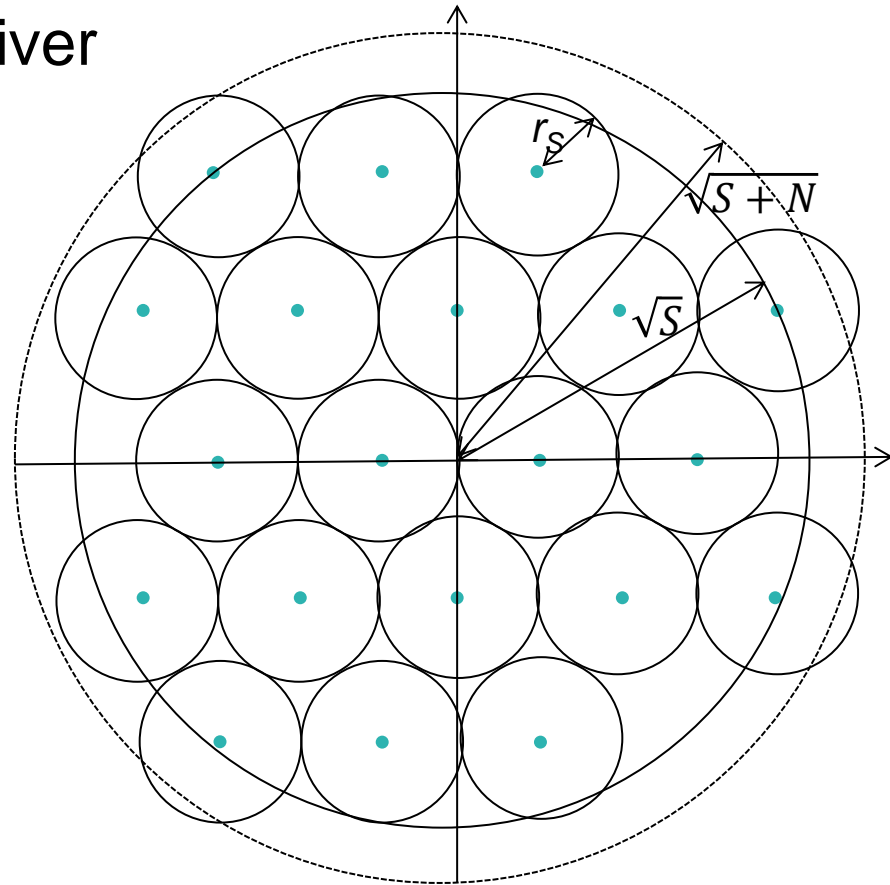
- If the lattice is viewed as a sphere packing, then the minimum Euclidean distance must be twice the sphere radius
- Signal power  $S$  proportional to  $\text{radius}^2$  of shaping region
- The greater the packing density, the greater  $M$  for given signal power
- $\text{Radius}^2$  of packed spheres proportional to maximum noise power



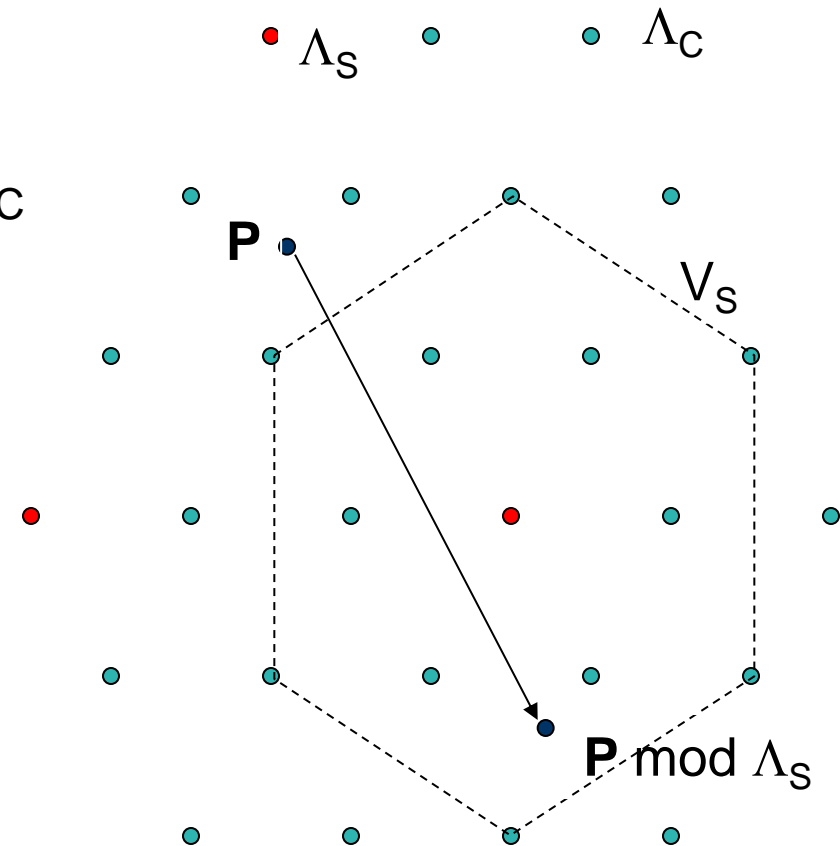
- Hence for low error probability, noise power  $N \leq r_s^2$
- Radius of signal space at receiver containing signal plus noise is  $\sqrt{S + N}$
- Volume of  $n$ -D sphere of radius  $r$  is  $V_n r^n$
- Hence max. no. of codewords in code

$$M \leq \frac{V_n (S + N)^{n/2}}{V_n r_s^{N/2}} \leq \left( \frac{S + N}{N} \right)^{n/2}$$

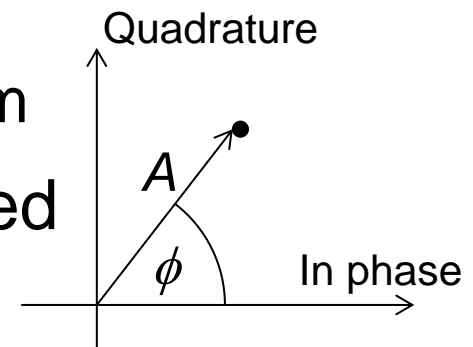
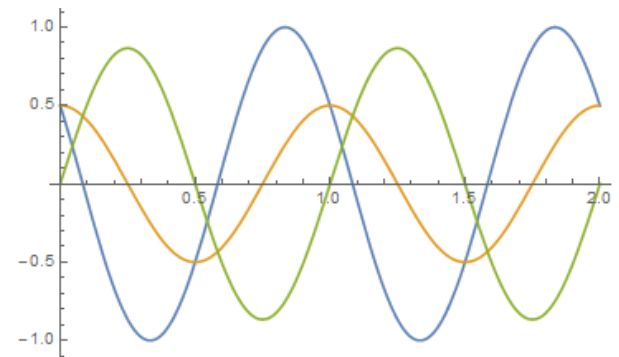
$$\frac{\log_2 M}{n} \leq \frac{1}{2} \log_2 \left( 1 + \frac{S}{N} \right)$$



- Define fine lattice  $\Lambda_C$  for the code
  - plus a **coarse lattice**  $\Lambda_S$  which is a sub-lattice of  $\Lambda_C$
- Then use a Voronoi region  $V_S$  of the coarse lattice as the shaping region
- Modulo- $\Lambda_S$  operation
  - for any point  $\mathbf{P} \notin V_S$  find  $\mathbf{P} - (\lambda \in \Lambda_S) \in V_S$



- Wireless signals consist of a sine wave **carrier** at the transmission frequency (MHz – GHz)
- Sine waves can be modulated in both amplitude and phase
  - hence the signal corresponding to each modulated symbol is 2-D
  - also conveniently represented as a complex value
  - typically represented on a **phasor** diagram
- Hence wireless signals can be represented in  $2n$  dimensions
  - or  $n$  complex dimensions



- Introduction to lattices
- Codes and lattice codes
- **Lattice constructions**
  - **Constructions A and D,**
  - **LDLC codes**
  - **Construction from Gaussian and Eisenstein integers**
- Lattice encoding and decoding
- Lattices in multi-user networks: Compute and forward



- For practical purposes in communications, we require lattices in very large numbers of dimensions
  - typically 1000, 10 000, 100 000...
- Lattices of this sort of dimension most easily constructed using FEC codes such as LDPC and turbocodes
- Most common constructions encountered are called Constructions A and D (Conway and Sloane)
  - Construction A based on a single code
  - Construction D is multilevel, based on a nested sequence of codes

- Start with a  $q$ -ary linear code  $\mathcal{C}$  with generator matrix  $\mathbf{G}_C$
- The set of vectors  $\lambda$  such that  $\lambda \bmod q$  is a codeword of  $\mathcal{C}$  form a Construction A lattice from  $\mathcal{C}$ :

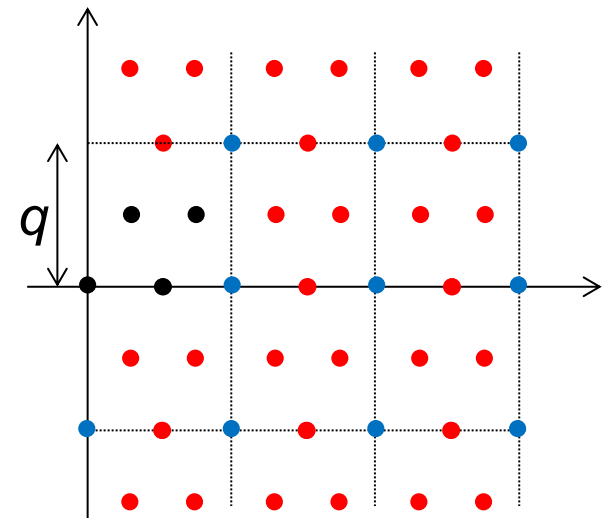
$$\Lambda = \{ \lambda : \lambda \bmod q \in \mathcal{C} \}$$

- Alternatively we can write:

$$\Lambda = q\mathbb{Z}^n + \mathcal{C}$$

- The generator matrix of the lattice:

$$\mathbf{G} = \begin{bmatrix} & \mathbf{0} \\ \mathbf{G}_C & q\mathbf{I}_{n-k} \end{bmatrix}$$

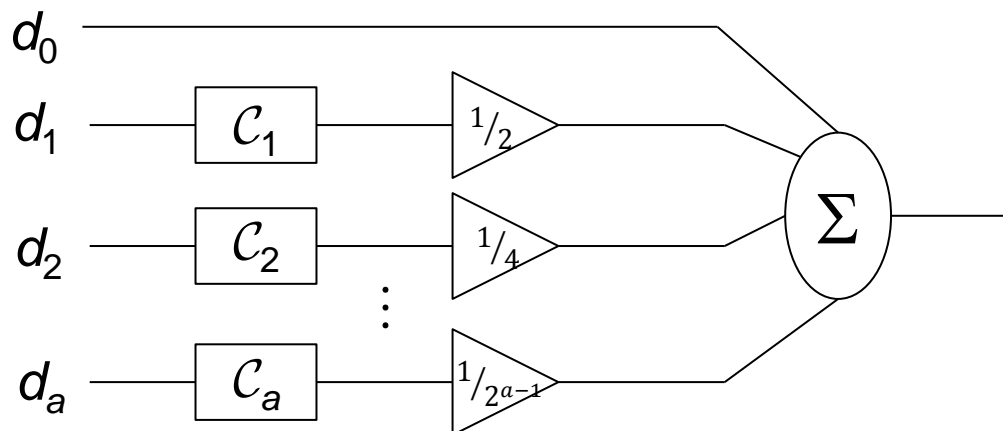


- Note that minimum distance is limited by  $q$

- Let  $\mathcal{C}_0 \subseteq \mathcal{C}_1 \subseteq \mathcal{C}_2 \dots \subseteq \mathcal{C}_a$  be a family of linear binary codes
  - where  $\mathcal{C}_0$  is the  $(n, n)$  code and  $\mathcal{C}_\ell$  is an  $(n, k_\ell)$  code
- Then the lattice is defined by:

$$\Lambda = \left\{ \lambda : \lambda = \mathbf{z} + \sum_{l=1}^a \sum_{j=1}^{k_l} d_j^l \frac{\mathbf{c}_{j,l}}{2^{l-1}} \right\}$$

- where  $\mathbf{z} \in 2\mathbb{Z}^n$ ,  $\mathbf{c}_{j,\ell}$  is the  $j^{\text{th}}$  basis codeword of  $\mathcal{C}_\ell$ , and  $d_j^\ell \in \{0,1\}$  denotes the  $j^{\text{th}}$  data bit for the  $\ell^{\text{th}}$  code



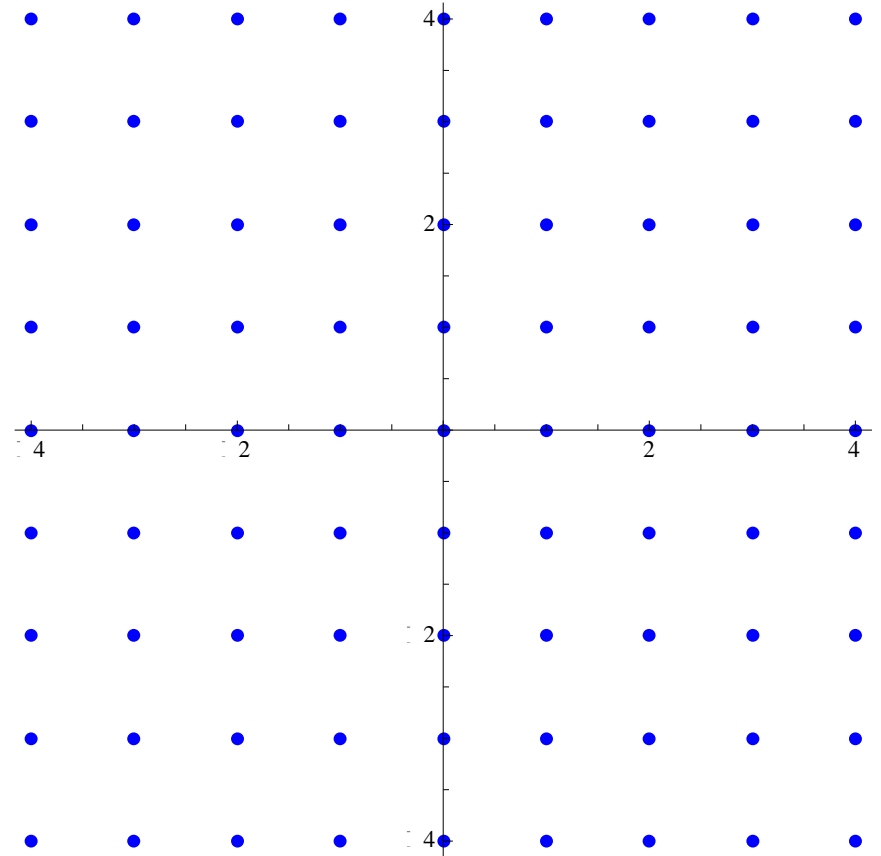
- Uses the principle of LDPC codes:
  - Define generator matrix such that its inverse  $\mathbf{H} = \mathbf{G}^{-1}$  is sparse
  - Then decode using sum-product algorithm (message passing) as in LDPC decoder
- However elements of  $\mathbf{H}$  and  $\mathbf{G}$  are reals (or complex) rather than binary
  - Messages are no longer simple log-likelihood ratios
- Ideally use nested lattice code
  - i.e. shaping region is Voronoi region of a coarse lattice

- Construction A/D and LDLC result in real lattices
  - can exploit Gaussian/Eisenstein integers to construct complex lattices
- Gaussian and Eisenstein integers form the algebraic equivalent in complex domain of the ring of integers
- Can construct complex constellations from them which form complex lattices

- Gaussian integers are the set of complex numbers with integer real and imaginary parts, denoted

$$\mathbb{Z}[i] = a + bi, a, b \in \mathbb{Z}$$

- They form a ring on ordinary complex arithmetic
- Hence operations in the ring exactly mirror operations in signal space
- Also form a lattice

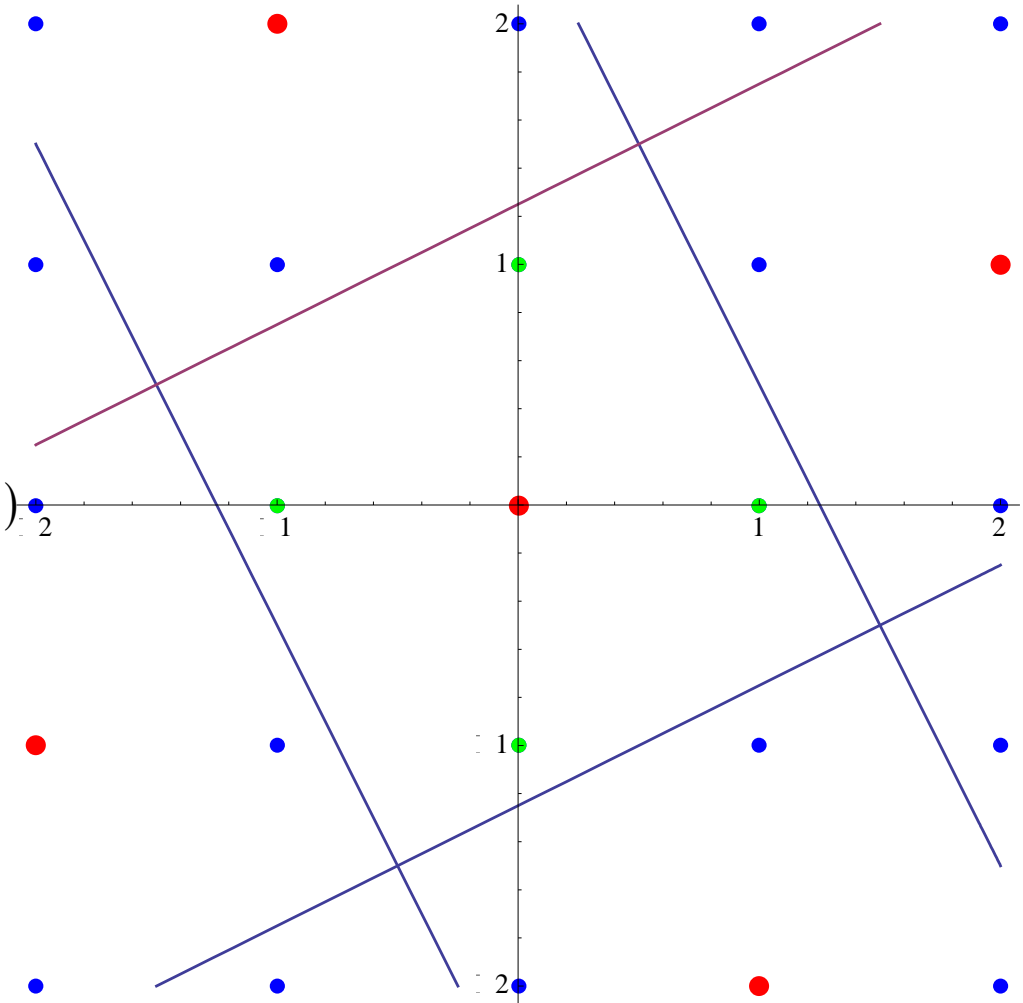


- Consider *fine* and *coarse* lattices,  $\Lambda_f$  and  $\Lambda_c$ , both based on Gaussian integers

$$\Lambda_c \subset \Lambda_f$$

- Here we assume that each point in the coarse lattice is a point in the fine multiplied by some Gaussian integer  $q$ 
  - i.e. the coarse is a scaled and rotated version of the fine
  - and the fine is just the Gaussian integers
- We then define our constellation as consisting of those Gaussian integers which fall in the Voronoi region of the coarse lattice

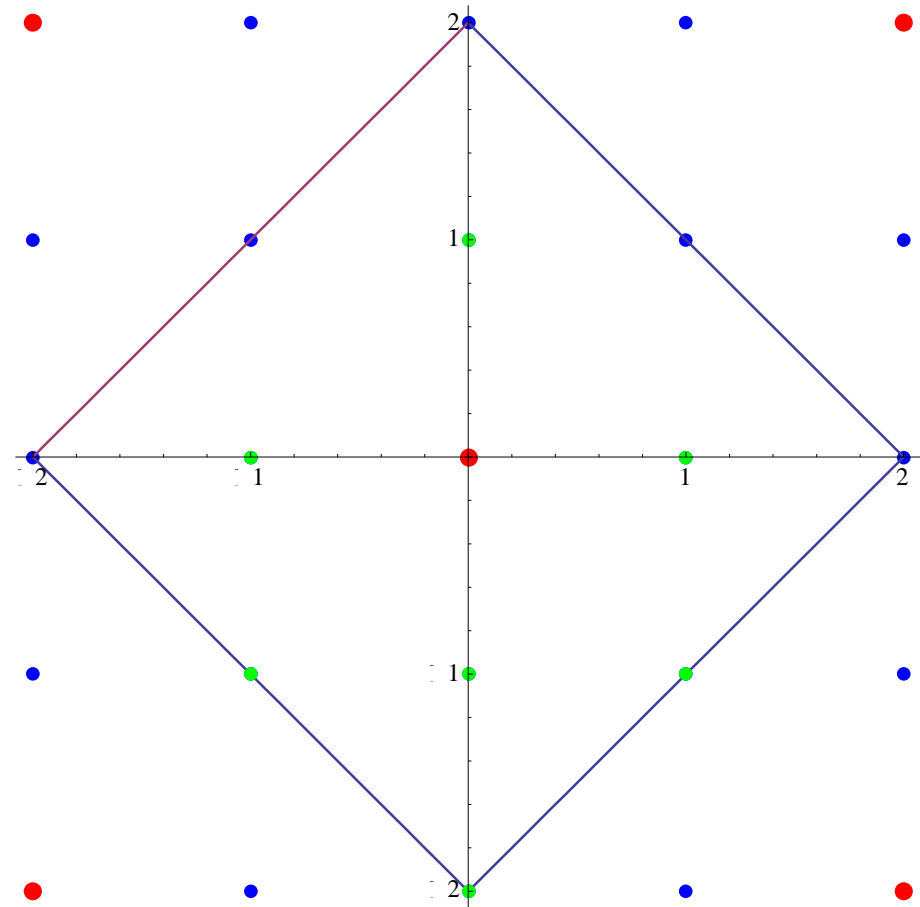
- e.g.  $q = 2 + i$
- Blue points are fine lattice
- Red points are coarse lattice
- Fundamental region  $V_c(0)$  is region closer to origin than any other coarse lattice point
- Hence constellation is green points, inc origin





- The fundamental region is surrounded by regions corresponding to  $q, qi, -q$  and  $-qi$
- We treat the boundaries of the latter two as belonging to the fundamental region
  - use this to allocate certain boundary points to constellation
- This also leads to an alternative definition of the fundamental region:

$$V_c(0) = \left\{ \lambda \in \mathbb{C} : \left( -\frac{|q|^2}{2} \leq \Re[\lambda] \Re[q] + \Im[\lambda] \Im[q] < \frac{|q|^2}{2} \right) \right. \\ \left. \& \left( -\frac{|q|^2}{2} \leq -\Re[\lambda] \Im[q] + \Im[\lambda] \Re[q] < \frac{|q|^2}{2} \right) \right\}$$



- We can establish **isomorphisms** between these constellations and either fields or rings
- An isomorphism is a one-to-one (or *bijective*, and hence invertible) mapping between the constellation  $\mathcal{C}$  and the ring  $\mathcal{R}$ 

$$\lambda = \mathcal{M}(s), \lambda \in \mathcal{C}, s \in \mathcal{R} \quad s = \mathcal{M}^{-1}(\lambda), \lambda \in \mathcal{C}, s \in \mathcal{R}$$
- such that the operations on the ring are equivalent to those on the constellation
 
$$\mathcal{M}(s_1 \otimes s_2) = \mathcal{M}(s_1) \mathcal{M}(s_2) \quad \mathcal{M}(s_1 \oplus s_2) = \mathcal{M}(s_1) + \mathcal{M}(s_2)$$
- It turns out that if  $q$  is a **Gaussian prime**, then the constellation is isomorphic to a field, otherwise it is isomorphic to a ring
- Size of field/ring is  $|q|^2$

- This isomorphism can be used to construct a complex lattice from a code based on the field or ring

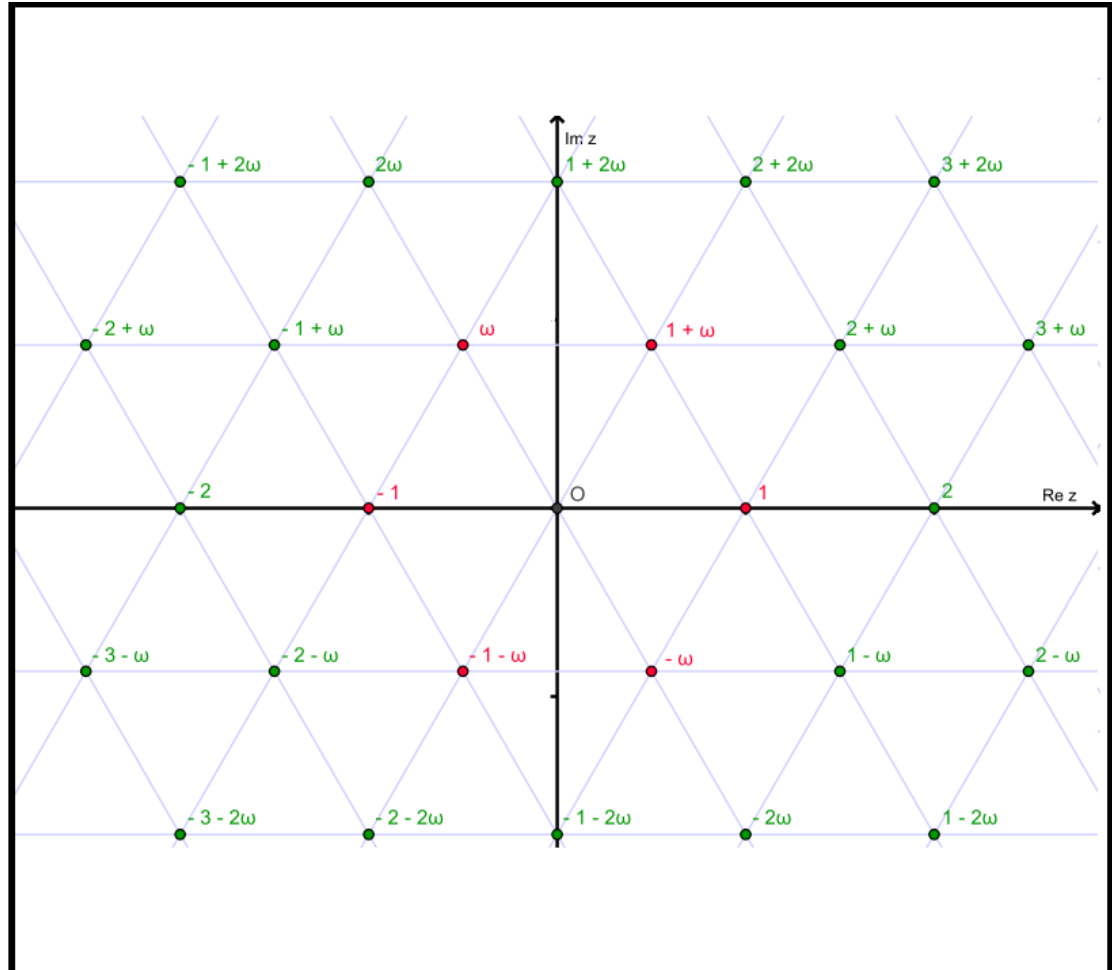
- in a manner equivalent to Construction A

$$\Lambda = \left\{ \lambda : \lambda = \mathbf{z} + \mathcal{M}(\mathbf{c}), \mathbf{z} \in q\mathbb{Z}[i]^n, \mathbf{c} \in \mathcal{C}\left(\mathbb{F}_{|q|^2}\right) \right\}$$

- that is, we encode a data sequence in the field  $\mathbb{F}_q$  using the code  $\mathcal{C}$  (over  $\mathbb{F}_{|q|^2}$ )
  - then map the resulting symbols to the complex constellation using the mapping based on the isomorphism
  - then combine with a lattice of Gaussian integers scaled by  $q$

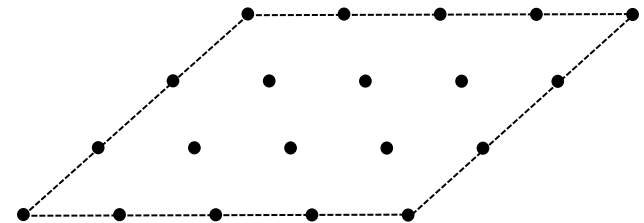
- Set of complex values with similar properties to Gaussian integers
- Hexagonal structure may result in denser lattices
- Note:

$$\omega = \frac{1 + i\sqrt{3}}{2} = e^{2\pi i/3}$$



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  - Problems of shaping
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- Ideally the shaping region should be as close as possible to a hypersphere
  - provides **shaping gain** up to 1.5 dB compared to hypercube shaping
- Nested lattice shaping gives a good approximation to this
- First multiply data vector by generator matrix
  - this may generate region of lattice of arbitrary shape
- Then apply modulo-lattice operation:
  - decode to coarse lattice, and subtract resulting coarse lattice vector
- In practice this decoding operation may be difficult
  - may use hypercube shaping as simpler alternative

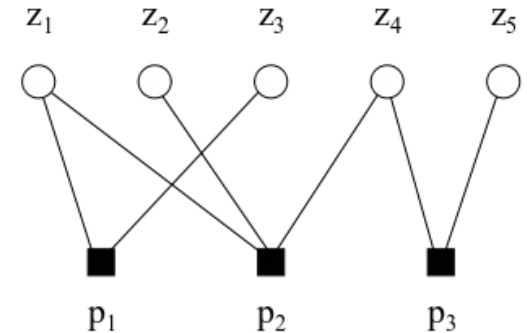


- Generally can be carried out with decoder for underlying code  $\mathcal{C}$
- Applying  $\text{mod}_q$  operation regenerates codeword of  $\mathcal{C}$ 
  - then decode this codeword
  - can then recover specific point in  $\mathbb{Z}^n$
- Note that in practice we use non-binary codes ( $q > 2$ )
  - because  $q = 2$  limits minimum distance and hence coding gain
- Typically use LDPC or turbocodes to achieve good performance
  - hence need non-binary sum-product or BCJR decoder
  - messages are probability distribution of  $q$  symbol values

- Use multilevel decoding approach based on component codes
  - decode codes  $\mathcal{C}_a, \mathcal{C}_{a-1}, \dots, \mathcal{C}_1$  in succession
- Component codes may usually be binary
- May require iterative approach
  - c.f. multilevel coded modulation

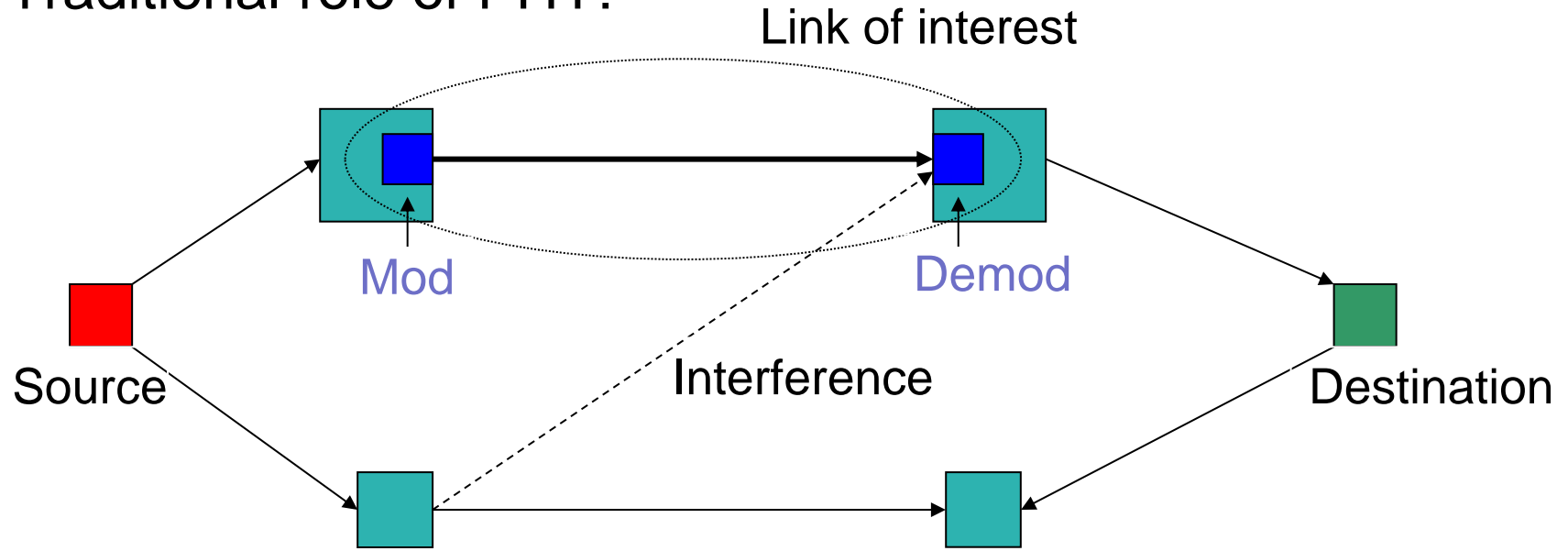


- Code structure designed for sum-product decoding, cf LDPC
  - using ***factor graph***
- However symbol values are now continuous variables (reals)
  - hence messages should be probability density functions
  - requires compact means of representing PDF in decoder
- May use Fourier or Karhunen-Loeve basis representation
  - or Gaussian mixture model



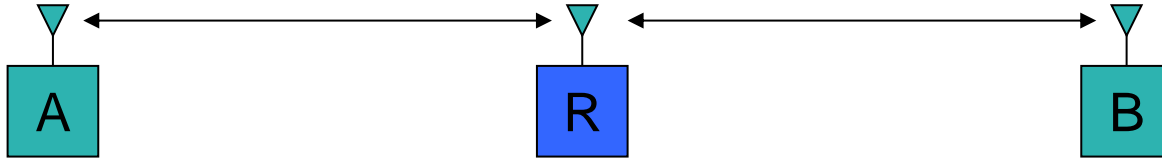
- Introduction to lattices
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- **Lattices in multi-user networks**
  - **Wireless physical-layer network coding**
  - **Compute and forward**

- Traditional role of PHY:

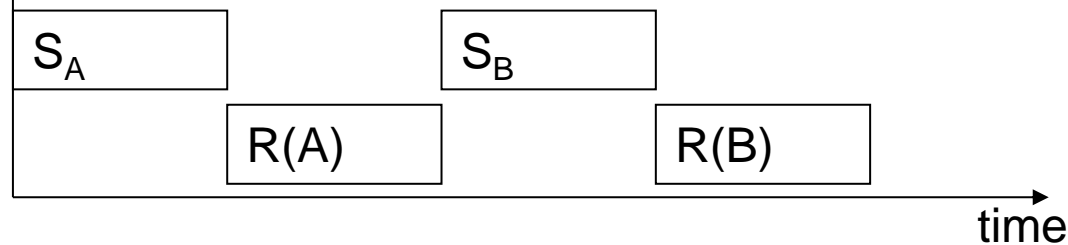


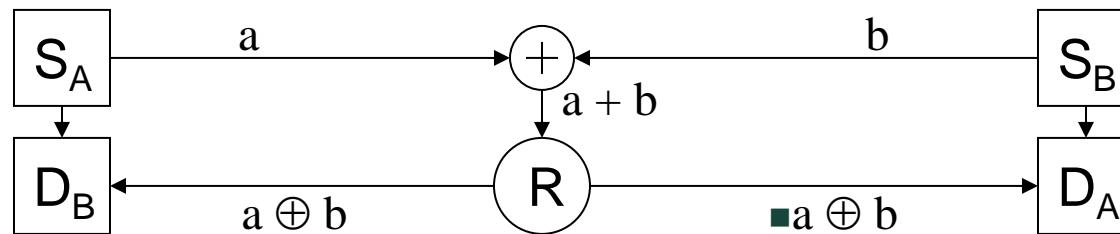
- signals from elsewhere in network treated as harmful interference
- however they may carry related information that can be exploited

- Two terminals want to exchange data via a relay:



- Conventionally this would require 4 time-slots:





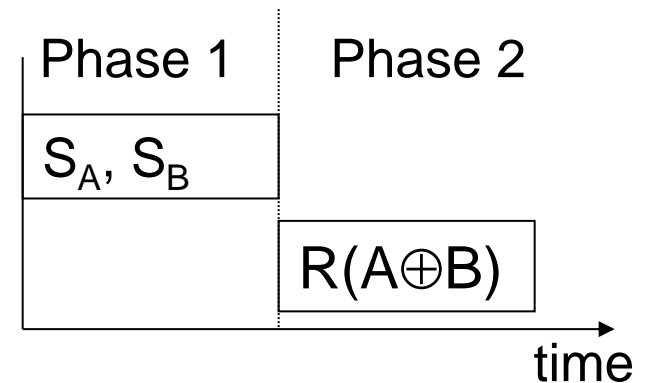
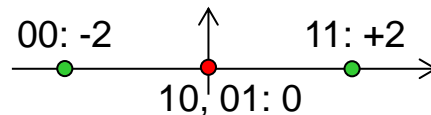
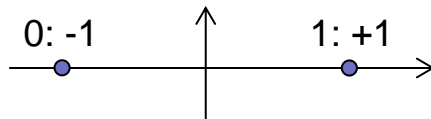
- We can do better using *Wireless Physical-layer Network Coding*

- using two phases

- Assume both sources transmit BPSK:

- map data symbol '1' to signal +1; '0' to -1

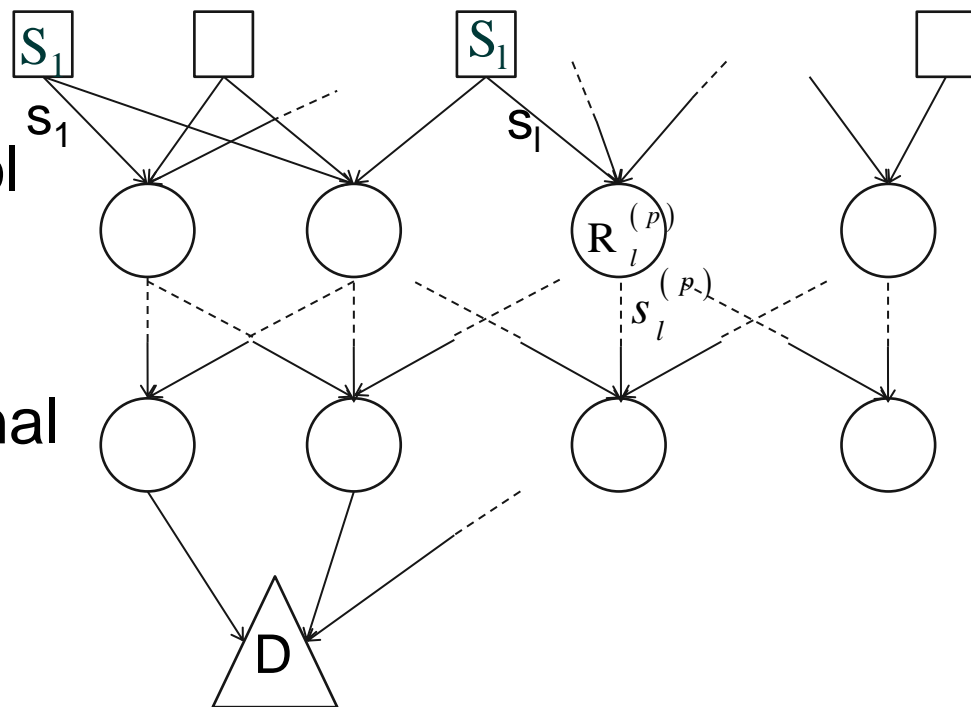
- At relay, map signals +2 and -2 to '0'; 0 to '1'



a	b	a+b	$a \oplus b$
0	0	-2	0
0	1	0	1
1	0	0	1
1	1	+2	0

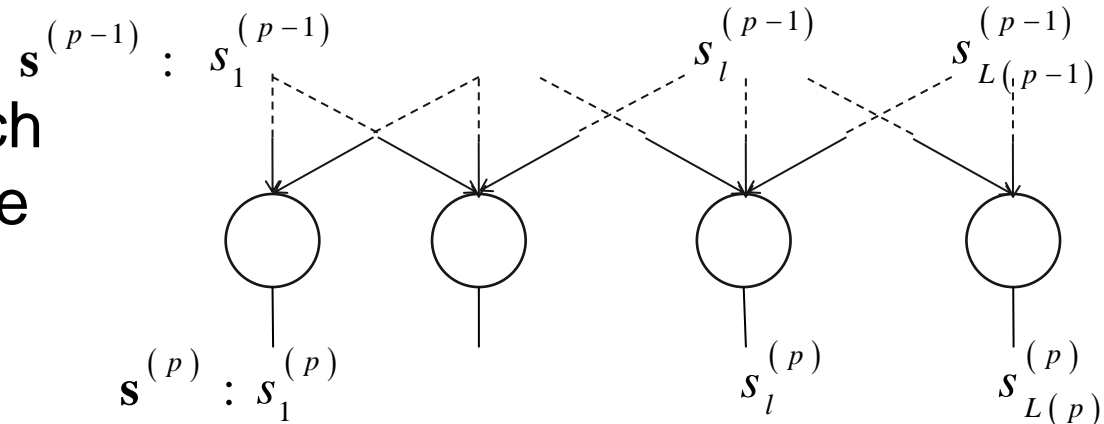
- Model a network with  $P$  layers of relays
- In general all nodes in a layer transmit simultaneously
- Each relay decodes a (linear) function of symbols from previous layer  

$$s_l^{(p)} = a_{1l}s_1^{(p-1)} + a_{2l}s_2^{(p-1)} + \cdots a_{Ll}s_L^{(p-1)} = \mathbf{a}_l \cdot \mathbf{s}^{(p-1)}$$
- based on the combined signals they receive
- Destination extracts symbol it is interested in from outputs of functions
- Lattices provide useful signal sets



- We can relate the vector of outputs of each layer to its inputs via the matrix **A**:

$$\mathbf{s}^{(p)} = \mathbf{A}^{(p)} \mathbf{s}^{(p-1)}$$



- We can combine these in cascade, so that:

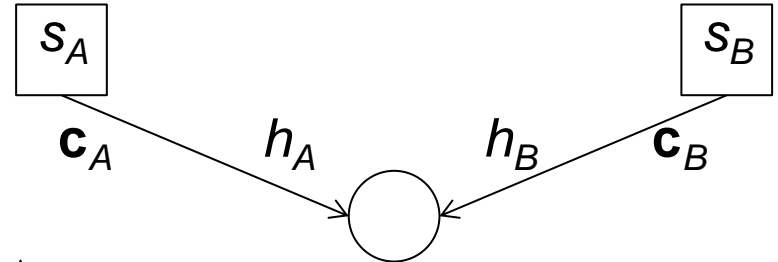
$$\mathbf{s}^{(p)} = \mathbf{A}^{(p)} \mathbf{A}^{(p-1)} \dots \mathbf{A}^{(1)} \mathbf{s}$$

- We can write this as a single matrix relating the vector of symbol  $\mathbf{s}^D$  at relays connected to the destination:

$$\mathbf{s}^D = \mathbf{B} \mathbf{s}$$

- We assume that the destination can (in principle) decode all symbols in its connection set

- Consider relay receiving from two sources via channel  $h_A, h_B$
- Sources transmit codewords  $\mathbf{c}_A, \mathbf{c}_B$  from the same fine lattice  $\Lambda_C$
- Received signal at relay is then:



$$\mathbf{x} = h_A \mathbf{c}_A + h_B \mathbf{c}_B + \mathbf{w}$$

- Now the sum of any integer multiples of two lattice points is another lattice point
  - hence if  $h_A, h_B$  were integers we could decode at the relay using the same lattice decoder
- Key idea is to scale received signal by scaling factor  $\alpha$  so that  $\alpha h_A$  and  $\alpha h_B$  are approximately integers



- Then:

$$\alpha \mathbf{X} = \alpha h_A \mathbf{c}_A + \alpha h_B \mathbf{c}_B + \alpha \mathbf{W} \approx a_A \mathbf{c}_A + a_B \mathbf{c}_B$$

- where  $a_A$  and  $a_B$  are integers
- Approximation error is:

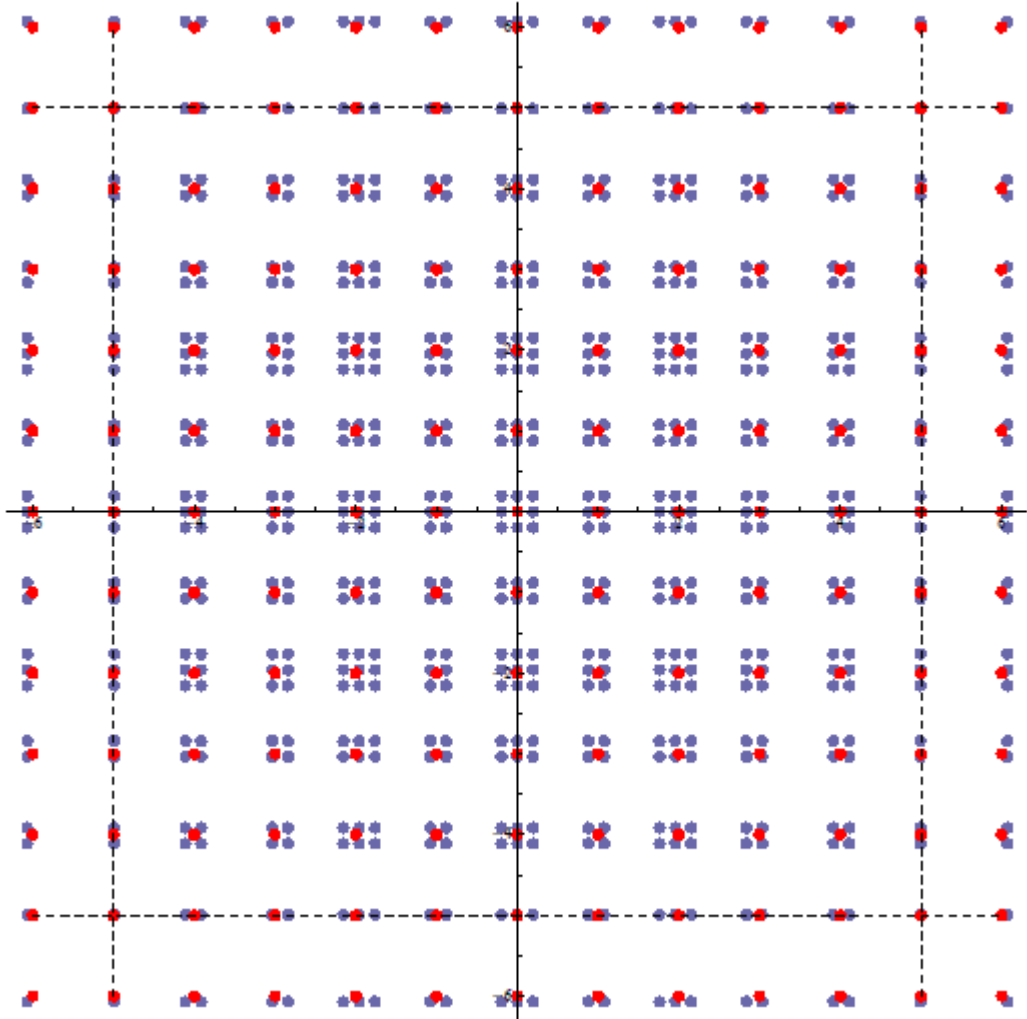
$$(\alpha h_A - a_A) \mathbf{c}_A + (\alpha h_B - a_B) \mathbf{c}_B + \alpha \mathbf{W}$$

- We can minimise this by choosing  $\alpha$ :

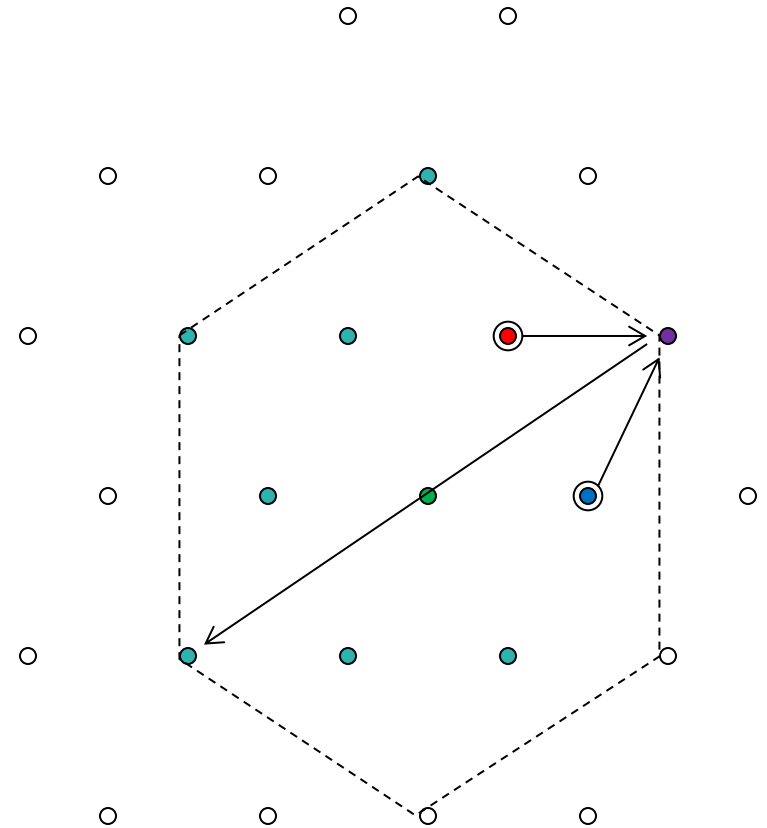
$$\alpha_{\text{MMSE}} = \frac{P \sum_i h_i a_i}{N + P \sum_i |h_i|^2}$$

- where  $P$  is signal power
- Also need to choose  $a_A$  and  $a_B$ 
  - could choose such that  $a_A/a_B = h_A/h_B$
  - but might require large  $\alpha$ , and hence increase noise

- $h_A = 0.55; h_B = 1.0$
- Choose:  
 $a_A = 1; a_B = 2;$   
 $\alpha = 1.95$
- Blue points are received signal
- Red are approximated lattice

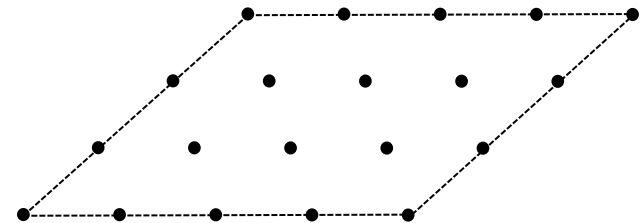
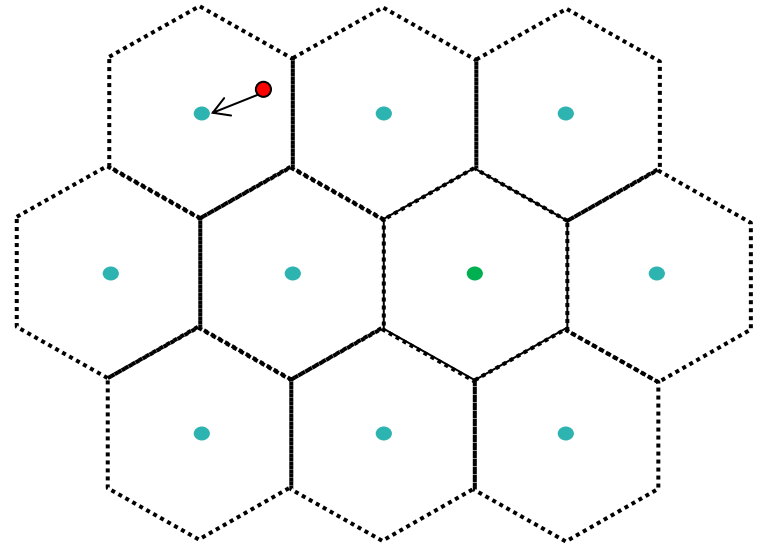


- Sum of two points from a lattice code may in general result in point outside shaping region
- Hence we apply modulo-lattice operation
  - returns a point in the original lattice code
  - so we can use the same decoder to recover sum point
- For lattice constellations isomorphic with field this operation can always be inverted



- Lattices can be extensively used in communications
  - especially for ***lattice coding***
- Can be shown to achieve capacity, as lattice dimension tends to infinity
- Practical lattice constructions are based on FEC codes
  - can provide high dimension lattices
  - with practical decoding algorithms
- For wireless channels use complex lattice constellations based on Gaussian/Eisenstein integers
- Important application is ***compute and forward***
  - applies to relay networks

- Lattice quantisation:
  - quantising signals to lattice points in high dimension can reduce mean square error
  - Applying modulo-lattice operation also allows Wyner-Ziv compression of correlated sources
- Lattice reduction aided MIMO detection
  - MIMO channel may distort received signal:
  - LRA treats as a different lattice



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