

Principles of Digital Communication

April 21, 2021

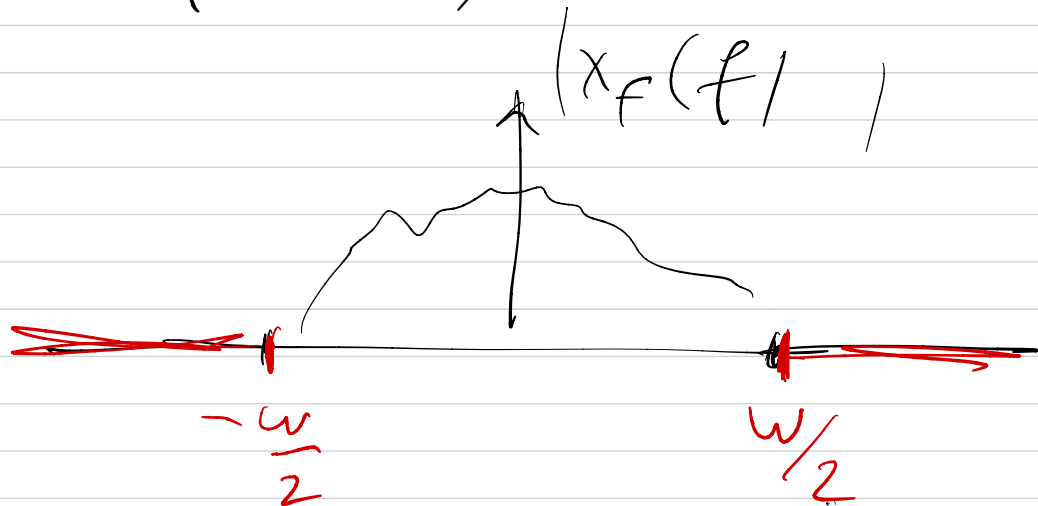


Reading was 4.4.7 - Summary of Ch 4.

a physical meaning of the signal-space dimension " N ".

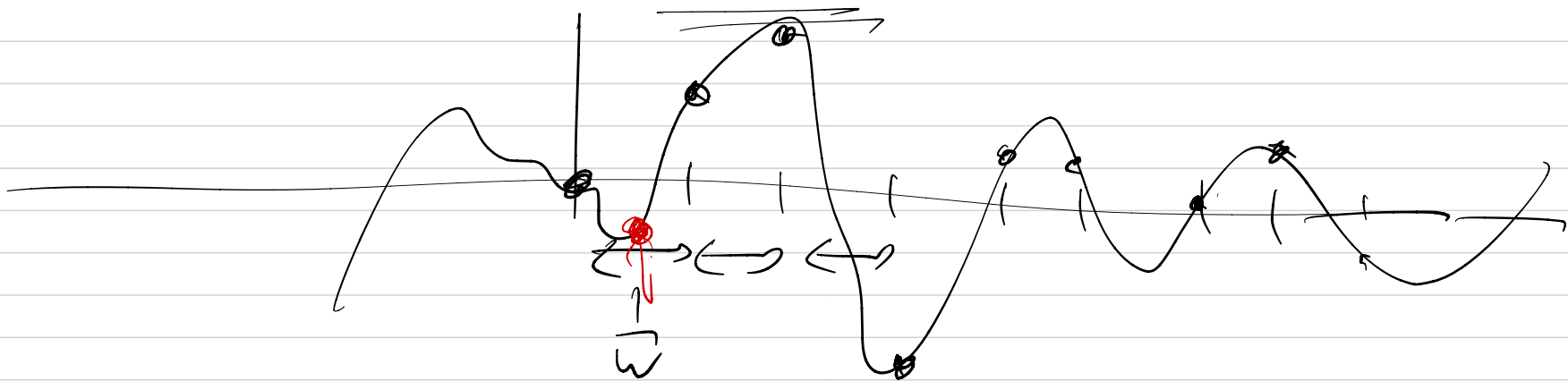
Recall the Nyquist Sampling Theorem:

if $x(t)$ is a waveform, bandlimited

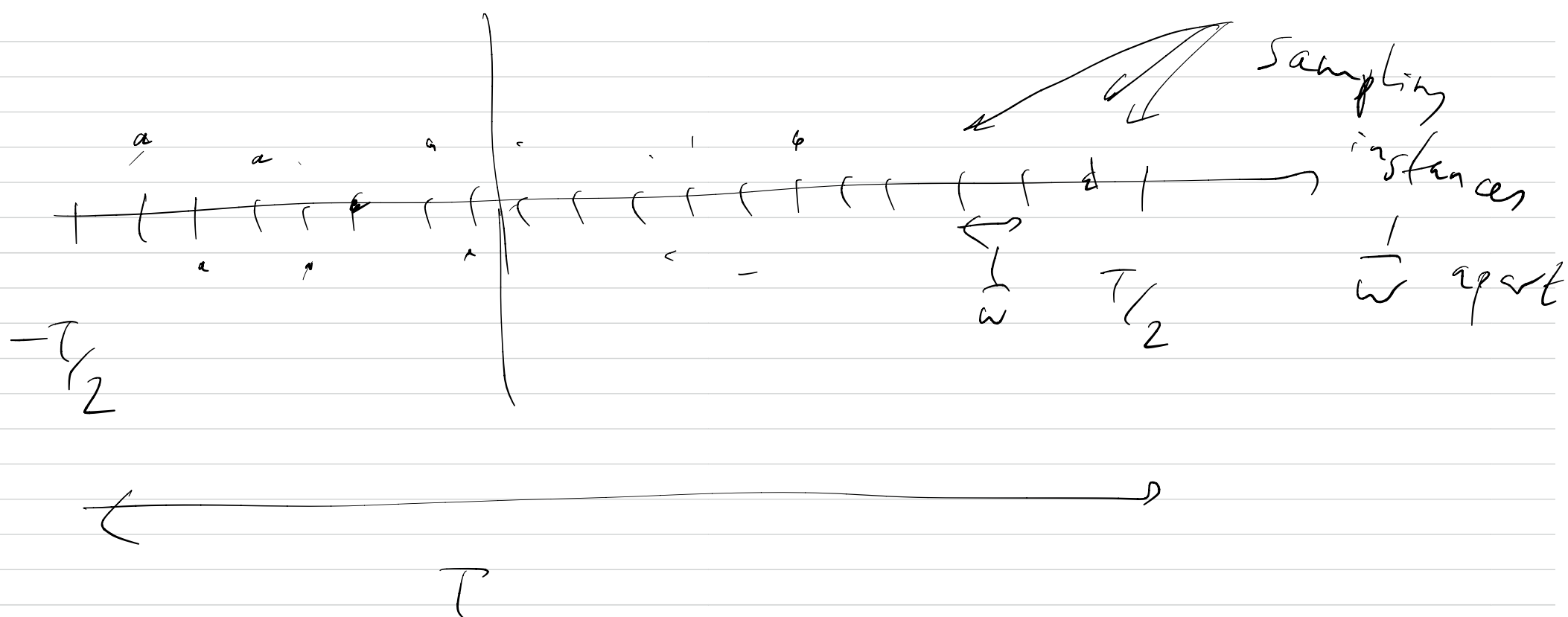


then $x(t)$ can be reconstructed from its
"Nyquist samples":

$$x_n = x\left(\frac{n}{W}\right) \quad n = \dots, -2, -1, 0, 1, 2, \dots$$



$$x(t) = \sum_{n=-\infty}^{\infty} x_n \text{sinc}(Wt - n)$$
$$\text{sinc}(Wt - n) = \begin{cases} 1 & Wt = n \\ 0 & Wt = m \neq n \end{cases}$$



we have wT samples in this interval
 we would "expect" that

$\boxed{(x(t) : -\frac{T}{2} < t < \frac{T}{2})}$ is determined by the

samples x_n taken inside the interval

(not really true) [wT samples]

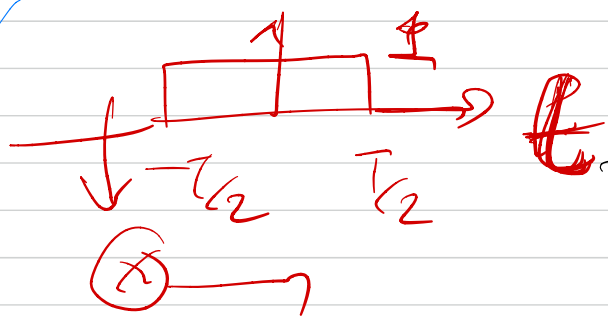
$$\equiv (x(t) : |t| < T/2) \leftrightarrow \mathbb{R}^{(wT)}$$

So intuitively speaking, from the Nyquist theorem we expect that

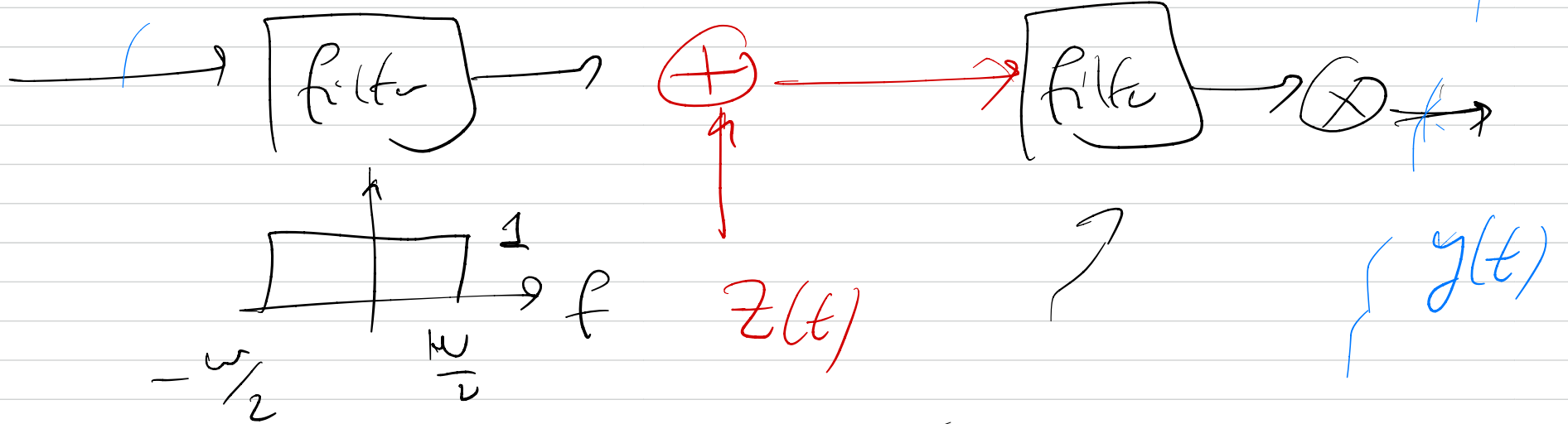
{ functions of duration T & Fourier
 footprint $\underline{w}$ } form a wT
 dimensional space.

This engineering intuition is made precise
in section 9.5. (due to a theorem by
D. Stepan, 1970s)

we try to design a communication system
for a bandlimited additive gaussian noise
channel.



$$w_i(t) = 0 \text{ when } |t| > T/2$$



$$y(t) = \int K(t, \tau) x(\tau) d\tau + \tilde{z}(t)$$

$$z_i = \sum_j K_{ij} x_j$$

General theorem

$$\underline{K(t, \tau)} = \sum_e \sigma_e u_e(t) \underline{v_e(\tau)}$$

Matrices: Singular Value Decomposition

$$K = U D V^T$$

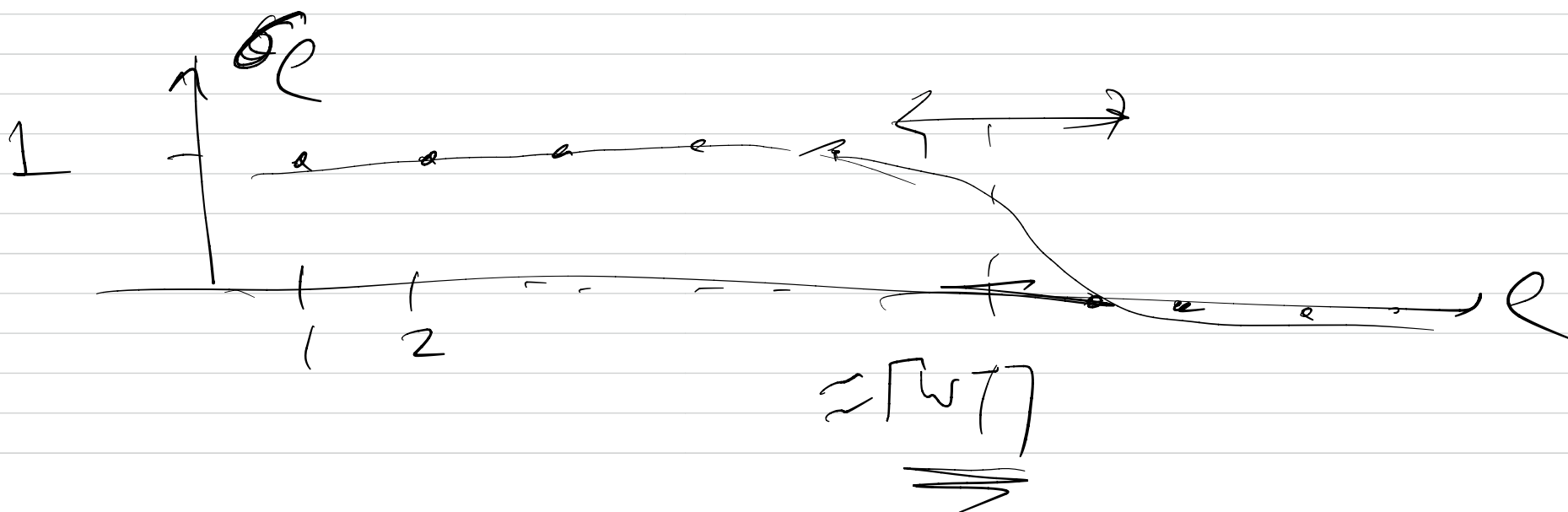
$$K = \sum_l \sigma_l (u_l v_l^T)$$

$$(a_1, \dots, a_n) \vdash a_i$$
$$(v_1, \dots, v_n) \perp w$$
$$I \subsetneq \mathbb{R}$$

If we write $w(\underline{e}) = \sum_e c_e v_e(\underline{e})$

$$\Rightarrow \text{output } y(t) = \sum_l \underbrace{e^{s_l t}}_{\text{homogeneous}} e^{u(t)} e^{\tilde{z}(t)}$$

Stepian: for the channel model above



Moral of the story:

Signal space is spanned by

$$\{v_1(t), v_2(t), \dots\}$$

but it makes sense to use only

$$v_1(t), v_2(t), \dots, v_n(t)$$

$n \approx \omega T$

4.4 (vs 4.6) trade-off between

k : # of data bits

n : # of dimen

$\mathcal{E}_b$: energy per bit, σ^2 : noise power.

Suppose we insist that our signals have

energy $\mathcal{E} = k \mathcal{E}_b$ so $\|c_i\|^2 \leq k \mathcal{E}_b$

if c_i is transmitted or received

$$y = c_i + z, \quad \|y\|^2 = \|c_i\|^2 + \|z\|^2 + 2\langle c_i, z \rangle$$

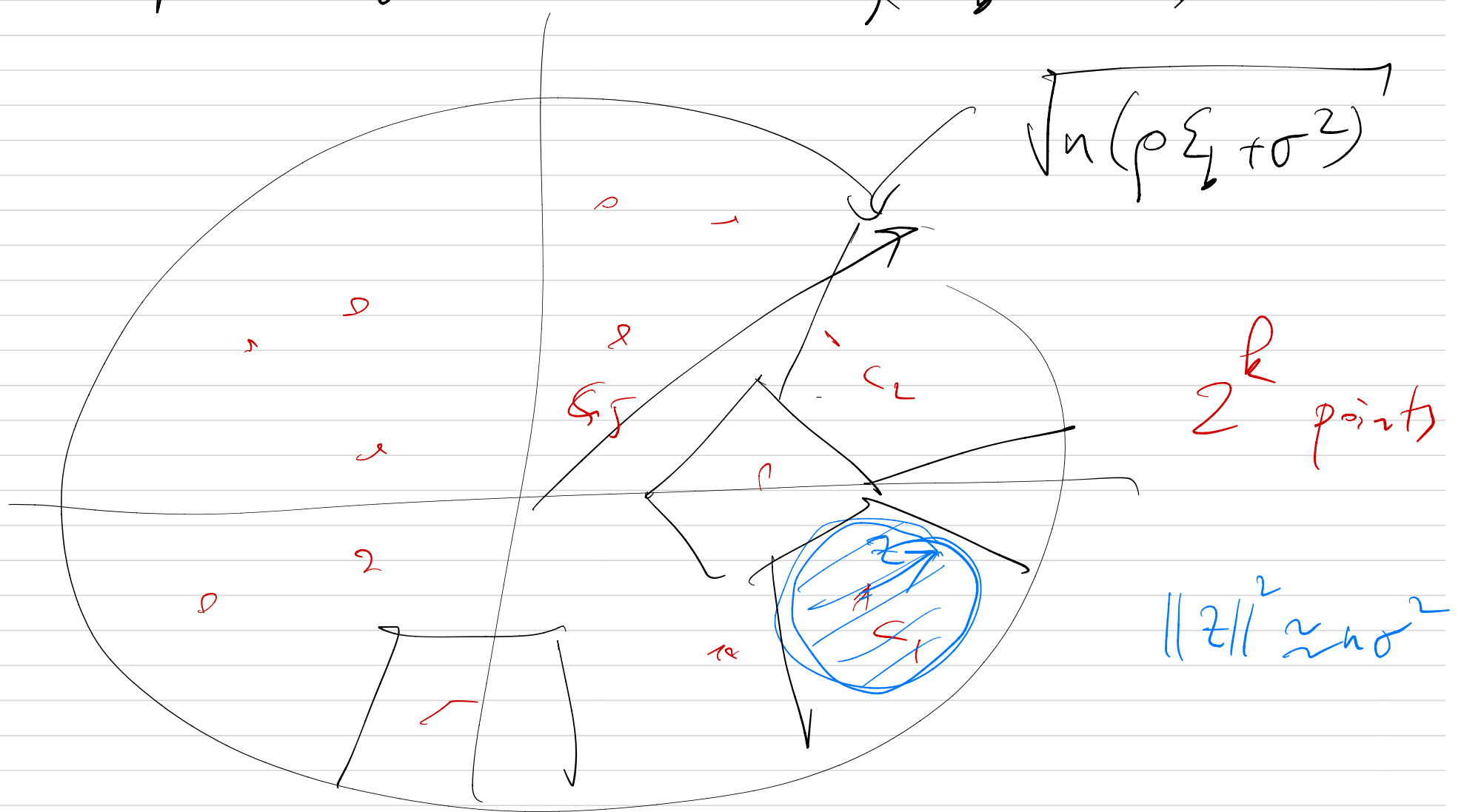
$\leq k \mathcal{E}_b$ $\sim n \sigma^2$ $\nearrow$

$N(0, \sigma^2 k \mathcal{E}_b)$

Suppose k & n are proportional to each other
 i.e. $\frac{k}{n} = \rho = \# \text{ of bits/dimension}$

$$\|y\|^2 \leq \underbrace{n\rho\xi_b + n\sigma^2}_{O(\sqrt{n})} + \underbrace{N(0, \sigma^2\rho n\xi_b)}_{O(\sqrt{n})}$$

y will, with probability ≈ 1 fall inside a
 sphere of radius $^2 = n(\rho\xi_b + \sigma^2)$.



to have a reliable system ($P(\text{Error}) \approx 0$) we
 better have the decision region R_i includes
 the "blue ball" of radius $^2 = n\sigma^2$ around c_i

So we need each R_i has to have

volume $\geq$ volume of the blue ball

$$\propto (n\sigma^2)^{n/2}$$

total volume of $R_0, \dots, R_{m-1}$

$$\geq \propto 2^k (n\sigma^2)^{n/2}$$

but since these are all inside the large sphere

of radius $^2 = n(\epsilon_b \rho + \sigma^2)$ we find that

$$\cancel{(n\sigma^2)^{n/2}} \cdot 2^k \leq \cancel{\left[n(\epsilon_b \rho + \sigma^2)\right]^{n/2}}$$

$$\Rightarrow 2^{k/n} \leq \left(1 + \rho \frac{\epsilon_b}{\sigma^2}\right)^{1/2}$$

$$\Rightarrow 2^p \leq \left(1 + \rho \frac{\epsilon_b}{\sigma^2}\right)^{1/2}$$

$$\Rightarrow \frac{\epsilon_b}{\sigma^2} \geq \frac{2^{2p} - 1}{\rho}$$

Moral: if we aim for dimensional efficiency ρ

we need to use energy/bit at least

$$\sigma^2 \left[\frac{2^p - 1}{p} \right]$$

Also, notice that $\frac{2^p - 1}{p} \geq (2 \ln 2)$.

So, for a reliable system we need $\frac{E_b}{\sigma^2} \geq 2 \ln 2$

(if we read ~~44~~ in this light we see that:
443)

$p \approx 0$, but $\frac{E_b}{\sigma^2} \approx 2 \ln 2$.

Recalls $\frac{2^p - 1}{p} = \frac{e^{2p \ln 2} - 1}{p}$

$$e^x = 1 + x + \frac{1}{2} x^2 + \dots \quad \text{So}$$

$$e^x - 1 > x \Rightarrow \frac{2^p - 1}{p} > \frac{2p \ln 2}{p} = 2 \ln 2$$

if p is small.

Question about 4.4.3 — How to compute the error probability of the "orthogonal transmission".

$$c_0 = [1, 0, \dots, 0] \sqrt{E} \quad E = k E_b$$

$$c_1 = [0, 1, \dots, 0] \sqrt{E} \quad k = \log_2 m$$

$$c_{m-1} = [0, \dots, 0, 1] \sqrt{E}$$

$$y = c_i + z \quad \text{if } i \text{ is the message}$$

$$\begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \sqrt{E} + \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix}$$

Suppose $\bar{c} = 0$ is sent message

$$y_1 = \sqrt{E} + z_1$$

$$y_2 = z_2$$

$\vdots$

$$y_n = z_n$$

The RECV will decide correctly if and only if $y_1 > y_2, y_1 > y_3, \dots, y_1 > y_n$.

$$P(\text{correct} | \bar{c} = 0) = P(y_1 > y_2, y_1 > y_3, \dots, y_1 > y_n) \quad (\bar{c} = 0)$$

$$= \Pr(\underbrace{\sqrt{\Sigma} + z_1 > z_2, \sqrt{\Sigma} + z_1 > z_3, \dots, \sqrt{\Sigma} + z_1 > z_n})$$

$$n = m = 2^k$$

$$\Pr(\overset{\text{correct}}{\text{HYPH}} \mid z_1 = z)$$

$$= \Pr(\underbrace{z_2 < \sqrt{\Sigma} + z, z_3 < \sqrt{\Sigma} + z, \dots, z_n < \sqrt{\Sigma} + z})$$

$$= \left(1 - \Phi\left(\frac{\sqrt{\Sigma} + z}{\sigma}\right)\right)^{n-1}$$

$$\Rightarrow P(\text{correct}) = \int \left(\downarrow \right)^{n-1} \frac{e^{-z^2/2\sigma^2}}{\sqrt{2\pi}\sigma} dz$$
