

INTRO COMMENTS

1

- "MENTORING A PROJECT" = "SHOWING YOU SOME INTERESTING PHYSICS/MATH UNTIL WE RUN OUT OF TIME"
- BEST APPROACH FOR SIX-DAY FORMAT: DO SOME STATS/DATA ANALYSIS (CALCULATIONS, EXPLAIN AS MUCH OF THE PHYSICS AS TIME WILL ALLOW)
- BACKGROUND: CALCULUS? SOFTWARE? ~~to~~ (UNITS?)

[DISTRIBUTE MATH PROBLEMS HERE]

STANDARD COSMOLOGICAL MODEL: 6 PARAMETERS

- 3 EXPANSION HISTORY (Ω_b, Ω_m, H_0)
- 2 PERTURBATIONS (Δ_s, n_s)
- 1 NUISANCE (τ)

EXPANSION HISTORY

CONTENTS OF UNIVERSE (AS FRACTION OF ENERGY DENSITY ρ_{tot})

- 69% "DARK ENERGY" (A COSMOLOGICAL CONSTANT $\Lambda \neq 0$?)
- 26% "COLD DARK MATTER"
- 4.9% "BARYONIC MATTER" (PROTONS + NEUTRONS + ELECTRONS)

STANDARD NOTATION

$$\Omega_b = \frac{\rho_{bary}}{\rho_{tot}} = 0.049$$

$$\Omega_c = \frac{\rho_{cdm}}{\rho_{tot}} = 0.26$$

$$\Omega_\Lambda = \frac{\rho_{DE}}{\rho_{tot}} = 0.7$$

} ONLY 2 INDEPENDENT PARAMETERS
SINCE $\sum \Omega_i = 1$

EXPANSION RATE (OR HUBBLE PARAMETER) TODAY

(2)

$$H_0 = 68 \text{ km s}^{-1} (\text{MEGA PARSEC})^{-1}$$

WEIRD UNITS! ORIGINAL INTERPRETATION (HUBBLE'S LAW)

$$\left[\begin{array}{l} \text{RECESSIONAL VELOCITY OF} \\ \text{DISTANT GALAXY} \end{array} \right] = H_0 \times \left[\begin{array}{l} \text{DISTANCE} \\ \text{TO GALAXY} \end{array} \right]$$

MODERN INTERPRETATION: FRACTIONAL EXPANSION OF UNIVERSE PER UNIT TIME

$$H_0 = 0.069 (\text{GIGA YEAR})^{-1}$$

THERE IS A SET OF DIFFERENTIAL EQUATIONS WHICH CAN BE SOLVED TO GET THE ~~THE~~ COMPLETE EXPANSION HISTORY (FRIEDMAN EQS)

$\Rightarrow H_0, \Omega_i$ TODAY DETERMINE H, Ω_i FOR ALL TIMES

PERTURBATIONS (2 PARAMS)

- AT EARLY TIMES, A FEW SECONDS AFTER THE BIG BANG, THE FLUCTUATIONS IN MATTER DENSITY ARE RANDOM, GAUSSIAN, WITH A POWER LAW SPECTRUM.

WE'LL SPEND THE NEXT 2 DAYS DEFINING ~~WHAT THESE~~ THESE TERMS!

FOR NOW WE'LL JUST SAY THAT A COMPLETE STATISTICAL DESCRIPTION OF THE INITIAL CONDITIONS IS GIVEN BY TWO PARAMETERS

$$\Delta_S = 4.7 \times 10^{-5}$$

$$n_S = 0.97$$

TYPICAL
FRACTIONAL SIZE OF INITIAL FLUCTUATION

"SPECTRAL INDEX" OR
SCALE DEPENDENCE

\odot DEFINED SO THAT $n_S = 1$

\Rightarrow SCALE INVARIANT

NUISANCE PARAMETER τ [OPTICAL DEPTH TO CMB]

(3)

BACKGROUND:

- UNIVERSE IS IONIZED (OPAQUE) FOR THE FIRST $\sim 400,000$ YEARS AFTER THE BIG BANG (UNTIL $z \sim 1100$)
- THEN BECOMES TRANSPARENT, PHOTONS FREESTREAM AND ARE OBSERVED TODAY AS THE CMB
- AT $z \sim 6$, SOMETHING ELSE HAPPENS: HIGH ENERGY PHOTONS FROM THE FIRST STARS RE-IONIZE THE UNIVERSE, NOT PERFECTLY TRANSPARENT ANYMORE

τ = PROBABILITY THAT A ~~PHOTON~~ CMB PHOTON SCATTERS BETWEEN $z = 1100$ AND NOW

$$= 0.06 \quad [6\% \text{ PROBABILITY}]$$

FORES

~~END OF~~

WE'VE NOW DESCRIBED THE PARAMETERS OF THE STANDARD COSMOLOGICAL MODEL. A FEW COMMENTS (ABRIDGED IN NOTES, TO BE ELABORATED IN LECTURE)

- NON-PARAMETERS: MANY THINGS WE LOOK FOR BUT DON'T SEE (SPATIAL CURVATURE, EXTRA NEUTRINO SPECIES, GRAVITY WAVES, AND MANY MORE)

(4)

- SEVENTH "CLOCK-LIKE" PARAMETER IS REALLY PART OF THE MODEL BUT NOT LISTED EXPLICITLY BY CONVENTION
- OPTICAL DEPTH τ IS STRICTLY SPEAKING NOT INDEPENDENT OF THE OTHERS IN PRINCIPLE, BUT IS AN INDEPENDENT PARAMETER IN PRACTICE
- COSMOLOGY IS A STATISTICAL THEORY, PREDICT PROBABILITY OF OBSERVING A GIVEN CMB REALIZATION GIVEN MODEL PARAMS
- THE FUN PART IS TRYING TO INVERT THIS: GIVEN A SET OF OBSERVATIONS, WHAT ARE THE COSMOLOGICAL PARAMETERS?
- TODAY + TOMORROW WE'LL TALK MOSTLY ABOUT RANDOM VARIABLES AND MATH PRELIMINARIES

RANDOM VARIABLE X : ANYTHING THAT HAS A PROBABILITY DISTRIBUTION $p(x)$

⑤



$$\text{PROB}(x_0 \leq X \leq x_0 + dx) = p(x) dx$$

$$\text{PROB}(x_0 \leq X \leq x_1) = \int_{x_0}^{x_1} dx p(x)$$

$$\text{NOTE THAT } \int_{-\infty}^{\infty} p(x) dx = 1 \text{ ALWAYS}$$

DEFINITIONS

$$\bar{x} = \langle x \rangle = \text{MEAN (EXP. VALUE) OF } X$$

$$\text{VAR}(x) = \langle (x - \bar{x})^2 \rangle = \text{"VARIANCE" OF } X$$

[SQUARE OF "TYPICAL" FLUCTUATION
AROUND MEAN]

WHEN DOING CALCULATIONS WITH EXPECTATION VALUES, WATCH
OUT FOR THIS PITFALL:

$$\langle x_1 + x_2 \rangle = \langle x_1 \rangle + \langle x_2 \rangle \quad \text{ALWAYS}$$

$$\langle c x \rangle = c \langle x \rangle$$

ALWAYS, IF c IS A CONSTANT
(I.E., NOT A RANDOM VARIABLE)

(*) BUT $\langle x_1 x_2 \rangle = \langle x_1 \rangle \langle x_2 \rangle$

ONLY IF THE RANDOM VARIABLES
 x_1, x_2 ARE INDEPENDENT

RELATED: IF $X = X_1 + \dots + X_N$ THEN

(6)

$$\bar{X} = \sum_i \bar{X}_i \quad \text{ALWAYS}$$

$$\text{VAR}(X) = \sum_i \text{VAR}(X_i) \quad \text{NEEDS IF } X_i \text{ ARE INDEPENDENT}$$

(NOT OBVIOUS BUT FOLLOWS FROM (*)
+ SOME ALGEBRA)

CENTRAL LIMIT THEOREM: EASIEST TO EXPLAIN BY PICTURE,
SEE SLIDES!

FORMAL STATEMENT: IF X IS THE SUM OF MANY INDEPENDENT,
IDENTICALLY DISTRIBUTED RANDOM VARIABLES

$$X = X_1 + \dots + X_N$$

THEN THE PROB. DISTRIBUTION OF X IS WELL-APPROXIMATED BY
A UNIVERSAL FORM ("GAUSSIAN DISTRIBUTION")

$$p(x) \approx \frac{1}{\sigma(2\pi)^{1/2}} \exp\left(\cancel{-\frac{(x-\mu)^2}{2\sigma^2}} - \frac{(x-\mu)^2}{2\sigma^2}\right)$$

WHERE $\mu = \langle X \rangle$ AND $\sigma^2 = \text{VAR}(X)$

~~PROOF OMITTED~~

- PROOF OMITTED (TOO MUCH OF A DIGRESSION)
- GAUSSIAN RANDOM VARIABLES ARE COMMON SINCE COMPLICATED RV'S
ARE OFTEN BUILT UP FROM MANY INDEPENDENT CONTRIBUTIONS
(EXAMPLES: REPEATED COIN FLIPS, VELOCITY OF A THERMAL PARTICLE)

⑦

- ONCE A RANDOM VARIABLE HAS "GAUSSIANIZED", ONLY ITS MEAN μ AND VARIANCE σ^2 MATTER (GAUSSIAN PROB. DISTRIBUTION ONLY DEPENDS ON THESE 2 PARAMS)
- FUN FACT: YOU CAN GET "e" AND "PI" BY REPEATED COIN FLIPPING
- THE CMB IS GAUSSIAN, IN A STRING SENSE, BUT NOT BECAUSE OF THE CENTRAL LIMIT THEOREM
(ELABORATE IN LECTURE)

2D GAUSSIAN RANDOM VARIABLES

(8)

NOW LET'S CONSIDER A RANDOM VARIABLE X WITH TWO COMPONENTS: $X = (X_1, X_2)$

E.G. X_1, X_2 MIGHT BE CMB TEMPERATURES AT TWO POINTS ON THE SKY WITH SOME FIXED SEPARATION

NOW WE HAVE INDICES, E.G.

RANDOM VARIABLE X_i 2-VECTOR

MEAN $\langle X_i \rangle = \bar{X}_i$ 2-VECTOR

COVARIANCE $C_{ij} = \langle (X_i - \bar{X}_i)(X_j - \bar{X}_j) \rangle$ 2-By-2 MATRIX

NOTE THAT $C_{ij} = C_{ji}$, I.E. C IS A SYMMETRIC MATRIX

TO GIVE SOME INTUITION:

$$C_{11} = \text{VAR}(X_1)$$

$$C_{22} = \text{VAR}(X_2)$$

C_{12} IS ZERO IF X_1, X_2 ARE UNCORRELATED

$= \pm \sqrt{C_{11}C_{22}}$ IF X_1, X_2 ARE PERFECTLY (ANTI)CORRELATED

THE CENTRAL LIMIT THEOREM HAS A 2D ANALOGUE

[SHOW SLIDES]

FORM OF THE GAUSSIAN PROB. DISTRIBUTION IN 2D IS
(ASSUMING $\bar{X}_i = 0$ FOR SIMPLICITY)

$$p(X_1, X_2) = (\text{CONST.}) \exp \left[\frac{-aX_1^2 - bX_2^2 - cX_1X_2}{2} \right]$$

~~TO EXPLAIN HOW TO GET VALUES~~

I'LL GENERALLY USE THE FIRST NOTATION (MOVES EXPLICIT ⑨)
BUT SUMMATIONS (IMPLICIT)

LET'S SOLVE THE SYSTEM, BUT FOR MORE GENERALITY
WE'LL LET THE RHS BE A GENERAL VECTOR Y_i :

$$2x_1 + x_2 = y_1$$

$$x_1 - x_2 = y_2$$

$$\Rightarrow 3x_1 = y_1 + y_2 \quad [\text{SUM}]$$

$$\Rightarrow x_1 = \frac{1}{3}y_1 + \frac{1}{3}y_2$$

$$\Rightarrow x_2 = \frac{1}{3}y_1 - \frac{2}{3}y_2 \quad [\text{BACKSUBSTITUTE}]$$

CONCLUSION: IF $A_{ij}x_j = y_i$, THEN $x_i = B_{ij}y_j$
WHERE B IS THE MATRIX

$$B = \begin{pmatrix} \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & -\frac{2}{3} \end{pmatrix}$$

THE MATRICES A AND B ARE "INVERSES" AND WE
WOULD NORMALLY WRITE $B_{ij} = A^{-1}_{ij}$.

$$A_{ij}x_j = y_i \Leftrightarrow x_i = A^{-1}_{ij}y_j$$

MULTIPLYING BY A^{-1}_{ij} "UNDONES" MULTIPLICATION BY A_{ij}

(10)

QUESTION: HOW ARE THE VALUES OF $\{a, b, c\}$ RELATED TO THE COVARIANCE MATRIX C_{ij} ?

TO ANSWER THIS QUESTION WE'LL NEED A LINEAR ALGEBRA DIGRESSION

LINEAR ALGEBRA DIGRESSION

CONSIDER A 2-BY-2 SYSTEM OF LINEAR EQS, E.G.

$$2x_1 + x_2 = 6$$

$$x_1 - x_2 = -3$$

LET'S WRITE THIS IN "INDEX" FORM

$$\sum_{j=1}^2 A_{ij} x_j = y_i \quad (*)$$

WHERE $y_j = \begin{pmatrix} 6 \\ -3 \end{pmatrix}$ $A_{ij} = \begin{pmatrix} 2 & 1 \\ 1 & -1 \end{pmatrix}$

AN EQUATION LIKE (*) WITH ONE "FREE" (I.E. UNSUMMED) INDEX ON EACH SIDE (i) STANDS FOR TWO EQUATIONS, ONE FOR EACH VALUE OF i .

~~ALTERNATE~~ ALTERNATE NOTATIONS FOR (*)

$$A_{ij} x_j = y_i$$

EINSTEIN SUMMATION CONVENTION:
REPEATED INDICES (E.G. j) IMPLICITLY SUMMED

$$\underline{A} x = y$$

MATRIX NOTATION: ALL INDICES AND SUMMATIONS IMPLICIT
MATRIX-VECTOR PRODUCT

MORE FUN WITH MATRICES: THERE IS AN "IDENTITY MATRIX" δ_{ij} (SOMETIMES WRITTEN I_{ij}) DEFINED BY

$$\delta_{ij} = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

IT HAS THE PROPERTY THAT

$$\delta_{ij} x_j = x_i \quad \delta_{ij} A_{jk} = A_{ik} \quad \text{ETC.}$$

ONE DEFINITION OF THE INVERSE MATRIX:

$$A_{ij}^{-1} A_{jk} = \delta_{ik} \quad [\text{DEFINES } A^{-1} \text{ GIVEN } A]$$

END OF THE LINEAR ALGEBRA DISCUSSION -- BUT THERE WILL BE A LOT MORE LINEAR ALGEBRA IN THE NEXT FEW DAYS!

NOW WE CAN EXPLAIN HOW THE COEFFS (a, b, c) IN THE 2D GAUSSIAN PROB. DISTRIBUTION

$$p(x_1, x_2) = (\text{CONST.}) \exp \left[\frac{-a x_1^2 - b x_2^2 - c x_1 x_2}{2} \right] \quad (*)$$

ARE RELATED TO THE 2-BY-2 COVARIANCE MATRIX C_{ij} .

THE GENERAL STATEMENT IS THAT

$$p(x_1, x_2) = (\text{CONST.}) \exp \left[-\frac{1}{2} x_i C_{ij}^{-1} x_j \right] \quad (**)$$

IF WE WRITE THE INVERSE COVARIANCE MATRIX EXPLICITLY AS

$$C_{ij}^{-1} = \begin{pmatrix} a & c/2 \\ c/2 & b \end{pmatrix}$$

THEN WE GET $(*)$. HOWEVER IT'S PROBABLY EASIER
TO THINK "IN MATRICES" AND USE THE ~~FORM~~ ALGEBRAIC
FORM $(**)$

WE'LL GIVE A DERIVATION OF $(**)$ LATER!