

Matric Theory
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Lecture No. 03
Vector Spaces

(Refer Slide Time: 0:14)

Handwritten notes on a grid background, likely from a presentation slide. The notes are written in blue and red ink. At the top, it says $(AB)^T = B^T A^T$ with a red arrow pointing to 'Show!'. Below that, it defines the trace of a square matrix $A \in \mathbb{R}^{n \times n}$ as $\text{tr}(A) = \sum_{i=1}^n A_{ii}$, which is the 'Sum of diag entries'. Then, it lists properties of the trace: $\text{tr}(A+B) = \text{tr}(A) + \text{tr}(B)$, $\text{tr}(\lambda A) = \lambda \text{tr}(A)$, $\text{tr}(A^T) = \text{tr}(A)$, and $\text{tr}(AB) = \text{tr}(BA)$ for 'compatible matrices' (with a red arrow pointing to 'Show!'). It then gives two viewpoints of matrices: (a) Rectangular array of scalars, and (b) Linear transformation between two vector spaces. The section 'Vector Space (VS)' is underlined. The second slide continues with the same two viewpoints, with 'vector spaces' circled in red. It then defines a 'Field (scalar field) F ' as a set of scalars with '+' and '·'. It lists properties: closed under '+' and '·', '+' and '·' are associative and commutative, there is an identity element, every element has an additive inverse, all elements except the additive identity (0) have a multiplicative inverse, and multiplication is distributive over addition. It notes 'Real' and 'Complex' as examples of fields. Finally, it defines a 'VS' (Vector Space) S or V over a field F as a set of elements with two operations: 'Elementwise addition' and 'Scalar mult.'. It gives an example: $x, y \in S \Rightarrow x+y \in S$.

So, there are, as I mentioned to you a little earlier there are two ways of looking at matrices. The first is that it is a rectangular array of scalars; that is the simple way to introduce matrices that is why I started with the definition. But as I mentioned the more useful viewpoint is that it represents a linear transmission between two vector spaces.

So, in order to understand that we need to know what are vector spaces and that is what I am going to define next. Are there any questions so far?

Student: Sir, you should we should perceive matrix multiplication as a linear transform, so I mean, in lower dimension it is very easy to say that, I mean, when we do a linear transformation what happens, but in higher dimension how should we ensure that this is a linear transform?

Professor: Can you repeat your question please?

Student: Yes, sir. I was saying that you said that matrix multiplication, we can perceive as a linear transformation.

Professor: Yes.

Student: So, I was asking that how can we even visualize it? Like in lower dimension it is very easy to see that this is a linear transform, but in higher dimension let us say doing from (\mathbb{R}^n) how can we prove it?

Professor: So, it turns out, I mean, we cannot visualize more than three dimensions. So, if it is two dimensions or three dimensions I can kind of draw things or I can show you what happens in three dimensions and so you can visualize it. But there is no hope of visualizing a linear transform from say 6 dimensional space to 8 dimensional space or 16 dimensional space down to 14 dimensional space and things like that.

You cannot visualize it. So, it is a mathematical construction and you have to take it as such, but that is what it is doing. It is taking a vector from 14 dimensional space and then mapping it to say 23 dimensional, something like that, so that is what it is doing. You cannot visualize it.

Student: I wanted to understand that, I mean, what does it distinguish that it is a linear transform and it is a non-linear transform, so how can we distinguish between these transforms?

Professor: So, this refers to how do we define linearity? So, I will come to that in a little bit, for that I need you to understand this concept of vector spaces and how we define a linear transform between vector spaces.

Student: Okay sir.

Professor: So, you do need to understand, we have to cover vector spaces before I can formally define what a linear transform is. But for now I am just saying that there are two

ways to visualize or view matrices, one is a rectangular array of scalars, the other is that a matrix represents a linear transform between a pair of vector spaces.

And the key point is that any linear transform, so I need to define vector spaces, so there is an object which is called a vector space and if I define two vector spaces and if I define a linear transform between two vector spaces, that can be represented as one and only matrix, so there is a unique mapping or a one-to-one mapping between linear transformations between two vector spaces and the space of matrices.

Student: Okay, sir.

Professor: So, we will come to that shortly. So, let us start with vector spaces. So, in order to define a vector space we have to start with a field. A field is a set of scalars and for the purpose of this course we are only going to essentially focus on real or complex field. So, that is the set of all real numbers or the set of all complex numbers.

So, in the back of your mind, even though I write F here, think of it as a short notation to say it is either real or it is complex. It is a set of scalars with two operations defined on it, plus and dot, and it is closed under plus and dot; that is you take any two scalars and add them together, you will another scalar which belongs to this field F and you take any two scalars and multiply them together, that is this dot symbol.

Then you will another element that belongs to this field F . Both plus and dot are associative and commutative. There exist an identity element both for addition and multiplication and every element has an additive inverse. So, given any a belonging to F , there is a minus a which also belongs to F . And all elements except the additive identity, which is typically denoted by 0 , have a multiplicative inverse and multiplication is distributive over addition.

Again this is a very formal sounding definition, but like I said for the purposes of this course just keep in mind that are thinking about the real line or the complex plane and the multiplication defined in the real line or the complex plane or multiplication of complex numbers. So, there is nothing here, but there is a formal way to define these things. I am not going to deal with these too much, but this is, I put these down mainly for the sake of completeness so that you know where these things come from.

(Refer Slide Time: 6:06)

Multiplication is distributive over addition

VS: S or V over a field F :

- 1) $x, y \in S \Rightarrow x + y \in S$
- 2) $x \in S \text{ \& } c \in F \Rightarrow cx \in S$

Elements of S are vectors.

$+$ and \cdot satisfy 8 axioms, not listed here.

Linear Combinations (LC):

Vectors v_1, v_2, \dots, v_n , $v_i \in \mathbb{R}^m$, $i=1, \dots, n$

Scalars $c_i \in \mathbb{R}$, $i=1, \dots, n$

Then $y = \sum_{i=1}^n c_i v_i$ Linear Comb.

$y = \begin{bmatrix} | & | & \dots & | \\ v_1 & v_2 & \dots & v_n \\ | & | & \dots & | \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$

$m \times n$ $n \times 1$

Linear independence (LI):

Then $y = \sum_{i=1}^n c_i v_i$ Linear Comb.

Linear independence (LI):

v_1, \dots, v_n LI when $\sum_{i=1}^n c_i v_i = 0$

iff $c_1 = c_2 = \dots = c_n = 0$.

Linearly dependent (LD) if not LI.

Span: $\text{Span}\{v_1, \dots, v_n\} = \{y \in \mathbb{R}^m \mid y = \sum_{i=1}^n c_i v_i, c_i \in \mathbb{R}\}$

$=$ a VS. \leftarrow show

$A = [v_1 \dots v_n] \in \mathbb{R}^{m \times n}$

$\mathcal{R}(A) = \text{Range space of } A = \{y \in \mathbb{R}^m \mid y = Ac, c \in \mathbb{R}^n\}$

$=$ a VS.

Subspace: Subset of a VS over F

Now, vector space - A vector space I am going to use either S or capital S or capital V to denote a vector space. It is defined over a field F and it satisfies two core properties. If I take x and y belonging to this vector space S , then their sum x plus y also belongs to this vector space S and this sum is defined as element wise addition.

If I take x belonging to this vector space S and any c belonging to this field F , then c times x also belongs to S and this scalar multiplication is defined as multiplying every entry of this x and these elements of S are called vectors. And addition and multiplication which I have used here satisfy some, a set of 8 axioms which I am not going to list here.

But again for the purposes of this course, just think of it as element wise addition and multiplying every entry of this vector x by this scalar. So, this x plus y as defined here is

actually taking a simple linear combination of these two vectors x and y , and more general linear combination; if you are given vectors V_1 through V_n in each V_i belonging to \mathbb{R}^m , this is the m dimensional real space.

That is a set of vectors with m real valued entries in them and i is 1 to n , those are the n set of vectors. And if you are given scalars i equal 1 to n c_i , then if I define a vector y which is equal to the summation i going from 1 to n , c_i times V_i , that is called a linear combination of these vectors V_1 through V_n . We also write it often by stacking these vectors V_1 to V_n as a matrix, then this matrix will be of size m by n .

Because of each of these vectors are m dimensional vectors and we stack the entries of the elements of this c_i as a column vector, c_1 through c_n , so this is n by 1 and then you take this product of this, this matrix vector product as I defined earlier, then that is exactly the same as doing summation i equal to 1 to n c_i times V_i .

The moment we define linear combinations we can define linear independence, so a set of vectors V_1 through V_n are linearly independent when summation i equal to 1 to n $c_i V_i$ equals 0; if and only if c_1 equals c_2 equals, et cetera equals c_n equals 0. It is important to take a minute and digest this definition.

Again this is something you would have seen in your undergraduate course, but one important thing I want to point out here is the 'if and only if' condition, the 'if' part is trivial here, of course, if c_1, c_2 , up to c_n are equal to 0, then summation $c_i V_i$ is always going to be equal to 0. 0 times a vector is a 0 vector and so when you add up all the 0 vectors you will another 0 vector. So, this is a 0 vector here, so the 'if' part is trivial.

So, really the crux of this definition lies in the 'only if' part; that is there is no other linear combination of these vectors V_1 to V_n that you can take and obtain the 0 vector. So, graphically the way to think about it is if I have a vector V_1 like this, another vector V_2 like this, then can I take a linear combination, scale this by c_1 , scale this by c_2 , add them together and then end up at the origin, get the 0 comma 0 vector.

If I can do that then these two vectors are linearly independent, if not they are linearly dependent. It turns out that these two vectors are linear independent and this is something that should be obvious to you. Instead if I take 3 vectors like this, now it turns out that I can always find a non-trivial linear combination of these 3 vectors, such that I will end up at the origin.

So, 3 vectors in the two dimensional plane are always going to be linear dependent. And so we say that a set of vectors are linearly dependent if they are not linearly independent. Again continuing with the theme of linear combinations, this span of a set of vector V_1 through V_n is a set of all wise which can be written as linear combinations of these V_1 to V_n .

It turns out that this is a vector space and again this is something that you can try to show. It is very easy to show this. The point is basically that if you take 2 vectors belonging to span to V_1 to V_n , then the first vector can be written as a linear combination of V_i like this and the second vector can also be written as a linear combination of these vectors and therefore their sum, so if the two vectors was y_1 and y_2 , y_1 plus y_2 can be written as sum of these vectors with different coefficients c_i and therefore that also lies in span of V_1 to V_n .

And similarly if you take, if you scale a vector y by some alpha then that is the same as scaling each of these coefficients by the same factor alpha and therefore alpha times y can also be written as a linear combination of these vectors and it belongs to this span. So, it satisfies the two properties we said that a vector space should satisfy and so the span of a set of vectors is actually a vector space.

A related object is the range space of a matrix A , which is the set of all y 's which can be written as linear combinations of the columns of A , so y can be written as A times c for some c in \mathbb{R} to the n , this is also a vector space. So, essentially the span of V_1 to V_n is the same as the range space of a matrix whose columns are V_1 to V_n and the range space of a matrix is a same as the span of its columns.

(Refer Slide Time: 13:17)

Linearly independent

Span: $\text{Span}\{v_1, \dots, v_n\} = \{y \in \mathbb{R}^m \mid y = \sum_{i=1}^n c_i v_i, c_i \in \mathbb{R}\}$
 $=$ a VS. \leftarrow show

$A = [v_1 \dots v_n] \in \mathbb{R}^{m \times n}$
 $\mathcal{R}(A) = \text{Range space}(A) = \{y \in \mathbb{R}^m \mid y = Ac, c \in \mathbb{R}^n\}$
 $=$ a VS.

Subspace: Subset of a VS over \mathbb{F}
Is a VS over \mathbb{F} . \mathbb{R}^2
 $\{y \in \mathbb{R}^2 \mid y_2 = 0\}$ subspace

$\{v_1, \dots, v_n\}$ span a VS S if $\text{Span}(v_1, \dots, v_n) = S$.

[Spanning set] Every $v \in S$ can be expressed as an LC of v_1, \dots, v_n .

Basis: $\{v_1, \dots, v_k\}$ is said to be a basis for a VS V if $\{v_1, \dots, v_k\}$ is a spanning set for V and is linearly independent.

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ly. dependent (LD) if not LI.

$$\text{Span}\{v_1, \dots, v_n\} = \left\{ y \in \mathbb{R}^n \mid y = \sum_{i=1}^n c_i v_i, c_i \in \mathbb{R} \right\}$$

= a VS. ← show

$$A = [v_1 \dots v_n] \in \mathbb{R}^{m \times n}$$

$$\mathcal{R}(A) = \text{Range space}(A) = \left\{ y \in \mathbb{R}^m \mid y = A c, c \in \mathbb{R}^n \right\} = \text{Column space}(A)$$

= a VS.

pace: Subset of a VS over \mathbb{F}

Is a VS over \mathbb{F} .

\mathbb{R}^2
 $\{y \in \mathbb{R}^2 \mid y_2 = 0\}$ output

v_1, \dots, v_n span a VS S if $\text{Span}(v_1, \dots, v_n) = S$.

spanning set] Every $v \in S$ can be expressed as an LC of v_1, \dots, v_n .

basis: $\{v_1, \dots, v_k\}$ is said to be a basis for a

A subspace of a vector space is basically a subset of a vector space, so you take a vector space and you throw out some of the vectors and you retain the others. But it should satisfy a property that this subset of vectors is itself a vector space over the same field, when it does that then we call it as subspace. So, for example if I take \mathbb{R}^2 , then the set of all vectors y belonging to \mathbb{R}^2 such that y_2 equals 0, that is the second entry of y is equal to 0.

This is a subspace. Clearly, if I take two vectors whose second entry is 0 and I add them together, the second entry cannot suddenly become nonzero and so that also belongs to this set and if I take y which belongs to this set, and I scale it by some alpha, then the first entry will get scaled by alpha but the second entry which is 0 will remain equal to 0, so that will also lie in this subspace. We say that a set of vectors v_1 to v_n span a vector space S if the span of v_1 to v_n is equal to this vector space S .

Student: Sir, can you please once again elaborate on the subspace part?

Professor: So, a subspace is nothing but a subset of the vectors in a vector space with the additional property that it should itself be a vector space.

Student: Okay sir.

Professor: And a vector space is one which satisfies those two properties that I showed you earlier.

Student: Yes sir.

Professor: The sum of two vectors in a vector space should be in the vector space and scaling a vector by a scalar, you should continue to live in that vector space, you can never leave.

Student: Yes.

Professor: I often joke, in physical class I often joke that vector spaces are like Hotel California, you can never leave, whatever you do, these vectors however they interact with each other, you will always stay in that vector space. If V_1 to V_n span a vector space then span of V_1 to V_n is equal to the set S this vector space S , in other words any vector in this vector space can be written as a linear combination of V_1 to V_n and any linear combination of V_1 to V_n is lying in this space.

So, this is another small point I want to make about, see this span of V_1 to V_n is a set of vectors and S is also a set of vectors and we want to say 2 sets are equal, that is equivalent to saying if I take any vector in S , that belongs to a span of V_1 to V_n and likewise, if I take any vector which belongs to a span of V_1 to V_n that lies in this set S , so they are equals. When this happens we call V_1 to V_n as a spanning set.

Of course, it means like I said this equality means that every vector in S can be expressed as a linear combination of V_1 through V_n . So, I think we have reached here and the next concept I want to discuss is that of a basis, which we will do on Wednesday. Any more questions before we close the class.

Student: Sir, can you please explain this range space once again?

Professor: Range space is the same as the span, range space of a matrix A is the same as the span of the columns of A and mathematically it is defined like this. It is actually the same as this definition here, so say that y is in R to the m , where y can be written as a linear combination of V_i is the same as saying that y is equal to A times c where c is a vector in R to n , it has n entries, it has c_1 to c_n as its entries.

Student: Okay.

Student: Sir, is it equivalent to the column space of the matrix?

Professor: Yes. So, that is a good point, this is also called as column space.

Student: Sir?

Professor: Yes.

Student: Could you explain span?

(Refer Slide Time: 18:30)

Linear Combinations (LC):

Vectors v_1, v_2, \dots, v_n , $v_i \in \mathbb{R}^m$, $i=1, \dots, n$

Scalars $c_i \in \mathbb{R}$, $i=1, \dots, n$

Then $y = \sum_{i=1}^n c_i v_i$ Linear Comb.

$y = \begin{bmatrix} | & | & | & | \\ v_1 & v_2 & \dots & v_n \\ | & | & | & | \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$

$m \times n$ $n \times 1$

Linear independence (LI):

v_1, \dots, v_n LI when $\sum_{i=1}^n c_i v_i = 0$

iff $c_1 = c_2 = \dots = c_n = 0$.

Linearly dependent (LD) if not LI.

Span: $\text{Span}\{v_1, \dots, v_n\} = \{y \in \mathbb{R}^m \mid y = \sum_{i=1}^n c_i v_i, c_i \in \mathbb{R}\}$

VS. \leftarrow show

$A = [v_1 \dots v_n] \in \mathbb{R}^{m \times n}$

$y = Ac, c \in \mathbb{R}^n$

$y = \begin{bmatrix} | & | & | & | \\ v_1 & v_2 & \dots & v_n \\ | & | & | & | \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$

$m \times n$ $n \times 1$

$\sum_{i=1}^n c_i v_i = 0$

LI.

$\mathbb{R}^m \mid y = \sum_{i=1}^n c_i v_i, c_i \in \mathbb{R}$

VS. \leftarrow show

$\{y \in \mathbb{R}^m \mid y = Ac, c \in \mathbb{R}^n\} = \text{Column space (A)}$

$v_1 = \begin{bmatrix} v_{11} \\ v_{12} \\ v_{13} \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} v_{21} \\ v_{22} \\ v_{23} \\ 0 \end{bmatrix}$

$c_1 v_1 + c_2 v_2 = \begin{bmatrix} c_1 v_{11} + c_2 v_{21} \\ c_1 v_{12} + c_2 v_{22} \\ c_1 v_{13} + c_2 v_{23} \\ 0 \end{bmatrix}$

v_1, v_2 span the x-y plane.

\mathbb{R}^2

$A \in \mathbb{R}^{m \times n}$ $B \in \mathbb{R}^{n \times p}$

$m \times n \times p$

multiplication: $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{n \times p}$

$= A_i B_j$

$i^{\text{th}} \text{ row}$ $j^{\text{th}} \text{ col.}$

commutative in general.

In general $AB \neq BA$

$-$ Composition of linear transforms

$-$ n -step transition probs. of Markov chains

$I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}_{n \times n}$

Professor: So, you take 2 vectors or in this case as defined here, span is a set of all linear combinations of these vectors V_1 to V_n , so in other words, if I take just to, again I am trying to avoid going to 1 and 2 dimensions because like I said linear algebra is not limited to one or two dimensions, but that is all I can show you here on a white board.

But if I take only vector in two dimension space and I ask what is its span, it is basically this line going through the origin. That is all the vectors that you can represent as a linear combination of this one vector here, but if I take two vectors here. But if I take two vectors in the two dimensional space, then their span is actually their whole plane, as long as these two vectors are linearly independent.

By taking different linear combinations of this I can span the entire two dimensional space. If I take two vectors in three dimensional space and let us say I take one vector here and another vector here, then these two... in this three dimensional space, but they will not span the entire three dimensional space. It is the set of all points that are reachable by taking a linear combination of taking the sum of scaled versions of the two vectors.

Student: Sir, could you repeat the plane part, your voice was not audible for a brief moment?

Professor: So, all I was saying is that if I take, if anybody is able to see the three dimensional plane, three dimensional axis that I have drawn, I have drawn the x, y, z axis.

Student: Yes.

Professor: If you are able to see it, please confirm.

Student: Yes, x, y axis.

Professor: So, if I take two vectors one vector along the x axis, another vector along the y axis, you can see that if I take all possible linear combinations of these two vectors, I will span the two dimensional plane defined by the x and y axis. There will be now no component in the z direction. So, it will span a two dimensional subspace of the three dimensional space and that is true.

Even if I take any two non-coincidental vectors in the x, y plan, together they will span the entire x, y plane but they will never have any component along the z direction. Every vector I take which is a scaled version of the first vector will have 0 as its z component, so specifically if I take V_1 equal to say V_{11} and 1, V_{12} , 0 and I take V_2 equal to V_{21} , V_{22} , 0, it will not be 01 and 10.

Any linear combination I take of these two vectors $c_1 \mathbf{V}_1$, $c_2 \mathbf{V}_2$, will always be of the form $c_1 \mathbf{V}_1$ plus $c_2 \mathbf{V}_2$, $c_1 \mathbf{V}_1$ plus $c_2 \mathbf{V}_2$ and 0. So, this third component will always be 0. So, it will always lie in the x, y plane.

Student: Sir, we cannot see what you are writing?

Professor: It will come in a minute. So, \mathbf{V}_1 and \mathbf{V}_2 span the x, y plane.

TA: Sir, Vishnu has his hand raised.

Professor: Vishnu, go ahead.

Student: Can you hear me sir?

Professor: Yes, please go ahead.

Student: The thing is earlier you said that vectors can be represented as columns of matrix, right?

Professor: Columns of a matrix are vectors.

Student: Yes, columns of a matrix are vectors, right? So, is it compulsory to use columns or can we use rows also?

Professor: Yes, so...

Student: But in textbook it is mentioned as columns, mostly.

Professor: That is where they say there are three types of people in this world, the kind who think of vectors as column vectors, the kind who think of vectors as row vectors and that is a bit of a joke, but essentially vector when I stated could be a column vector or a row vector, the point is one of its (dimen), when we say vector we are thinking of a one dimensional vector, that it has one dimension which is where you have say n elements and it is a string of entries written along that dimension.

You can represent it either as a row or as a column and in fact, we will use both depending on the convenience, but definitely from, it is true that vectors are often, most common to think of vectors as column vectors.

Student: Okay.

Professor: So, in fact if I go back here in my definition, I used both. I used a row vector and I used a column vector.

Student: Sir, what are the third kind of people?

Professor: That is the joke.

Student: Okay.

Student: Sir, one more thing.

Professor: Yes.

Student: Sir, matrix is like linear combination between two vector spaces, linear transformation, right?

Professor: Yes.

Student: Sir, we have matrices like m by m , m by m or something... m by m by n .

Professors: Tensors, I am not discussing tensors just yet.

Student: Okay.

Professor: Matrix by definition in this course is going to be of 2, there are going to be 2 parts to it, m by n ; that is it.

Student: Okay sir.

Professor: I will not be, at least in the, for the most $(())$ (26:08).

Student: Thank you.

Professor: I will need another course to teach tensor mathematics. So, if there are no further questions we will stop here. Thanks for attending.

TA: Sir, Rashi has a question.

Professor: Go ahead please.

Student: Hello, sir. In this last example where you explained the two dimensions vectors along the axis x and y , so here we took these two vectors along the axis x and y , but if we take these two vectors along certain plane, I mean, one vector along x , y plane and one vector along some other plane say it x , y or y , z plane, then would it be three dimensional, would be able to span the three dimensional space?

Professor: What you think?

Student: I guess, we must have another vector to span three dimensional space.

Professor: Precisely, so that is one of the things I will show, which is that you cannot span three dimensional space using just two vectors, no matter how you choose those vectors. If I take two, three dimensional vectors, I can always find a vector in three dimensional space which cannot be $(\)_{(27:34)}$ as a linear combination of these vectors. It makes intuitive sense, right?

Student: Yes.

Professor: Because if I take the three dimensional space like this, it is hard for me to draw it here, but if I take some vector like this, another vector pointing in some other direction, these two guys together define a plane.

Student: Yes, sir.

Professor: And it is only vector that is in this plane that I can reach by taking linear combinations. There is always going to be a perpendicular direction which is 90 degrees to both these guys, and anything that sit in this 90 degree direction cannot be reached by taking linear combinations of these two vectors.

Student: Yes, sir. Okay, sir, thank you.

Professor: Welcome. So, I guess we will stop here for today.