

## Pre-Scrimmage

RGB that stands out. Something special about our robot is RGB, which is something Alex has wanted to do since he started Vex, even though our coach says "it's a waste of time". At least we <sup>were</sup> in style. There are two primary types of RGB strip, a 4 pad and a 3 pad. How the 4 pad work is there is 1 ground pad and 1 pad for each color (R, G, B). This means the whole strip will be that color, how ~~color~~ works is it mixes the primary colors, red, green, and blue ~~so~~ we input varying voltages but a vex brain can not do that. The old cortex motors also needed a analog signal too but used the 29 Motor Control. This controls a Pulse Width Modulation Signal (PWM) into a analog PWM by varying lengths of very fast on and off in a signal. Which the CPU can send. The issue is we need 1 for each color which takes up a lot of room and port, also is ~~illegal~~ legal because of ~~CR6.1B~~. The other option is a addressable WS2811 led strip where using a PWM we can control every individual led but it uses a special PWM protocol. Since it is essentially turning a ~~port~~ on and off we can implement this but the kernel makes it so that all user programs are run on 1 CPU 1 thread. (continued on pg 5)

Scrimmage  
We switched to tall bot to give us more room and also de scoring was not effective because of no double zone. We realized 2 essential things we needed to improve was ~~spurter~~ wings and intake, new more reliable wings and faster intake. I show a diagram showing how forces are generated wings.

## Formosa (cont'd) / 12

We travelled to Taipei to compete in Formosa, met many teams, and learnt lots from matches. We did okay during most of our matches.

One issue we encountered was that the sleds were unreliable and not smooth; this was because it had a very steep angle. We fixed this with our next iteration.

Before competitions we waited for competition to tighten and turnouts. Even with loctite, our collar locks still sometimes fell out and it is very annoying to retighten. The reason is that the two collar locks are in the corners. We overcame this by using spacers to move it to the center and easily access it.

Another issue was bracing, the dt metal could bend in and out in the front which caused friction in the drive, this was fixed by adding a bar that spanned the whole thing. The bigger bracing issue was the punchers, if it bent very much during the match, as it was mounted by 4 stand offs. Our new design mounts onto 4 bars via 10 contact points, with boxing allowing us to hold the robot from any point.

The wedge we had worked very well pushing balls robots to other side.

We also found out that we have quite some time after before pushing triballs into goals we would ram into bots but it was not very effective so we add a blaster that will pop up using 1 piston.

Fulbright

17/12

Fulbright was a very interesting competition, the vets made some poor calls (We should have won SF) but we were still very happy our robot worked very well.

Our puncher may not have been great but it was mostly reliable which gave us 69 points (one in auto skills without moving).

Our blocker was the thing that helped us win a lot of matches. We could stop very good penumbra shooters helping lower the score at the opponent.

Our elevation strategy was using the wings, this worked twice, one time our alliance thought we were stuck and pushed us off, and one time the ref said it didn't count. It was very simple and could have got a lot of points, next time we just need to give the drivers more practice.

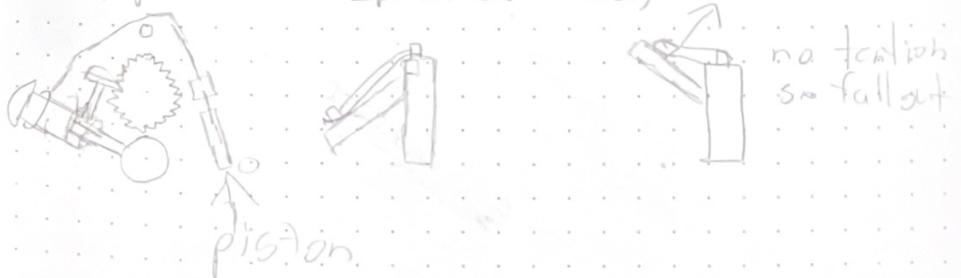
What we learnt was that there are lots of teams with good offence but not a lot with good defence, therefore we will want to focus on improving blocker and wings (we will still keep a decent puncher for skills). And in case our alliance can't match loads.

For our next version we are making a double stage to block robots that have a high arch. We will also ~~use~~ the two pistons to make the blocker less wobbly.

Explaining driving (continued in page 13)

AIVP

The way a team is ranked in qualification is by Win Point which is crossed with each win but also when we get the autonomous win point which is completing a series of tasks but the challenge is we need to take a ball out of the match load zone and push a tv ball into our own goal, these two are on opposite sides which is normally done by two robots but we don't want to be reliant so the solution is shooting the ball half court. This is hard because the ball is smaller than goals so a lot of power is needed to push into it. A solo AIVP mech is when we shoot with more force than the motor can pull down where the puncher is initially pulled down by hand. ~~How we plan to accomplish this is we have a ratchet that only allows the puncher to go down but instead of a rubber band, it is attached to a piston which will tension when the match starts but will release to shoot. We will mount rubber bands so that when they go up they become loose and full of puncher until ziptied onto robot.~~



no teeth  
so fall at

RFB

(extends page 1)

The kernel makes it so that all user program runs on the CPU thread which is fast enough for most things but a neopixel needs a 800Hz signal or every 1.25 microsecond taking 8 bits per color per pixel so 24 bits for one pixel & this results in a slow changing led and slow complete animations. James Pearman helped implement this in SPK 20220726-10-00-00 which uses CPIOO and allows for rapid communication. We put a long led strip the whole way with power injection to keep both ends the same brightness. We created a gradient by splitting the three colors, found the difference between two desired colors, then divide by a amount of led we have. increment by this amount every led. We then make a animation by putting the gradient into a vector then use `leds.rotate()`.

## Autonomous Motion Algorithms

Autonomous movement is something that seems simple at first but it took us a long time through hundreds of hours of testing. The goal of autonomous is moving to a desired location consistently and accurately but the challenge is that there are many unpredictable factors like friction or being pushed.

The simplest method of movement is Dead Reckoning, it has no feedback and is very unreliable. Given our robot which travels 6 inches/second with 6 volts applied if we want to move forward 6 inches, we apply 6 volts for 1 second. There are a lot of reasons why this wouldn't work, the most obvious one being acceleration, the robot will not reach the 6 inches

speed immediately so will travel less than calculated. An improved version can take into account acceleration but even then, it stills many problems. A motor will not always spin the same speed with the same input voltage because of the battery level which will alter the torque of the motor. Additionally the friction will also differ on different floor surfaces. Therefore this algorithm is very rarely used.

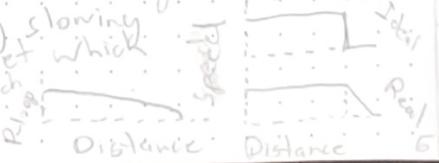
The way to measure the unpredictable factors is odometry, the simplest odometry uses encoders built inside of the V5 motors. The encoders of the v5 motors act as a rotation sensor for the motor, that output is square wave.

A tachometric encoder has a disk with many holes for light to shine through and then optical sensors measure the speed of oscillation.

There are also two optical sensors whose pulses can be compared to get the direction by seeing which pulse goes high first, this means encoders can help us give us feedback on how much we have to spin the motor.

A better way of dead reckoning is to go forward until the encoder spins the desired amount but a issue is the sensor might never output 0.0. so we can set a dead band by  $\text{abs}(\text{desiredError}) < 0.1$ .  
drive.spin()

The issue is this will overshoot because when we stop the wheels will still have momentum, this can be reduced by short braking which shorts circuit the motor by a shunt FET across windings but this will not always stop immediately and straining the motor. We can solve this by slowing down when near the target which is called a P-controller which stands for proportional.



desived = 611. inches

```
while True:
    error = desived - current
    speed = error * kp
    if speed > clamp(speed):
        speed = clamp(speed)
    drive.move(speed)
```

that we get by tuning. It decides at which distance the robot will start slowing down, if before this it will not yet more than the motor's max speed so we clamp the value to ±12 volts. This is a very simple controller that works well but there are many things that can still be improved. When approaching the target the error of a P controller will always be a non-zero value because the controller output puts such a low value that the robot does not overcome friction. We can solve this with a Integral term.

 By definition a "integral" is the area under a curve, in this case we add all the previous error. By accumulating the integral of past errors the integral adds a correction that grows over time in response to persistent steady-state error which eliminates the steady-state error while reducing settling time.

integral + error : There are two issues to this approach  
Speed = (error \* kp) +  
(integral \* ki)

if abs(error) <= 1 : if error is zero the integral is still non-zero so the speed is non-zero, this is mitigated by  
integral = 0 : setting the integral to zero if the error is within the accepted values. Integral  
if integral > huge : wind up is also an issue which is

when going a far distance and when approaching the set point, the integral is usually high so we set integral to zero every time it reaches a high value. A issue with a PI loop is oscillation and rapid unpredictable changes which is helped with a derivative term.

In a P-controller, the speed is proportional to the remaining distance so the robot will slow down when near the desived location. A addition value called kp is proportional gain which is a constant



Derivative is the rate of change or gradient of curve. We use a devitiae to dampen the robots movements based on what is expected to occur. It will be a signal with the direction opposite to the direction of travel with a greater magnitude for greater speeds, it typical will be out weighed by the proportional and integral components but if there is some deviation from normal the derivative will compensate. It is calculated derivative-error = previous by measuring the difference between previous error - error by the previous error. error = (error \* kp) + (integral \* ki) + This concludes the PID controller but we still need to figure out the constants. (derivative \* kd).

Tuning the controller is one of the most tedious parts of PID as it is time intensive. Here are the steps we:

Parameter Rises over time setting steady stability  
Time step: 1 Increase kp until slight oscillation  
kp Decrease N/A Decrease Worsen 2 Add to 0 until no oscillation  
ki Decrease Increase Decrease Worsen 3 Add to 0 until no steady-state error  
kd N/A Decrease Decrease N/A Improve  
it takes a lot of tuning to get the optimal char (expanded on pg 13).

This PID controller was invented by engineer Nicolas Minard for automatic ship steering in 1842 but is now used widely for many uses in our daily life like temperature control. We learnt PID from 'A Introduction to PID' by George Billard.

The method to implement PID is also very interesting, the simplest method as mention before is using the motor encoders, we calculate how much the wheels need to spin to move the distance using basic geometry and then calculating the rotation by measuring the distance between the turning center and wheel which gives the radius which can be used to calculate the arc length and rotation the motor spins. This PID is done by most teams on the cortex MB built into each motor using the built in function but the NO is slow and also non-tunable so we bypass the cortex and output a direct signal.

The issue with this is if we are pushed while moving, the robot movement would become inaccurate. We solve this with odometry.

Robot skills would champion robots created the concept of APG which we learnt when we met them in the VSAert worlds. The absolute positioning system is a system which keeps track of the robot's absolute position in cartesian coordinates and orientation. The core of this system is odometry which converts encoder signals into a series of movements. Often tracking wheels which are rubber banded so it contacts the floor constantly to reduce wheel slippage by the sensor which we have ordered haven't arrived yet so we use motor encoders. Normally people will have a back track wheel<sup>↓</sup> to track horizontal but since we can't do track wheels we use a we use <sup>movement</sup> radial wheels so the robot can't move horizontally. ΔS = movement

definitions. ΔL left movement

ΔR right movement

s distance from center to wheel

Δθ change in orientation

r<sub>m</sub> middle radius

r<sub>l</sub> left radius

r<sub>r</sub> right radius

$$\Delta L = r_L \Delta \theta \quad \Delta R = r_R \Delta \theta \quad // arc length formula$$

$$\Delta L = (r_m + s) \Delta \theta \quad \Delta R = (r_m - s) \Delta \theta$$

$$r_m = \frac{\Delta L + \Delta R}{2\Delta \theta} - s \quad r_m = \frac{\Delta R - \Delta L}{2\Delta \theta} + s \quad // rearrange$$

$$\frac{\Delta L}{\Delta \theta} - s = \frac{\Delta R}{\Delta \theta} + s$$

$$\Delta \theta = \frac{\Delta L - \Delta R}{2s} \quad // rearrange$$



Using the arc length formula

we can use the amount traveled by each wheel to form a arc and then calculate the angle of that arc by plugging it in each other.



$$\Delta S = \frac{\Delta L - \Delta R}{2}$$

$$\Delta X = \Delta S \cos(\Delta \theta)$$

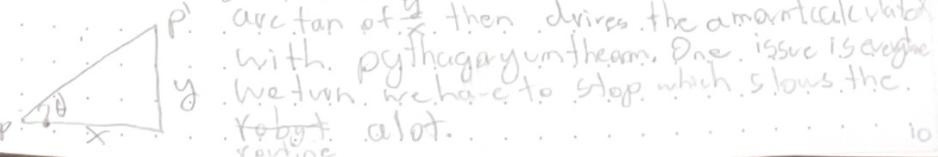
$$\Delta Y = \Delta S \sin(\Delta \theta)$$

$$P[X] = [X] + [\Delta X]$$

We then can figure out the amount moved relative using basic trigonometry then add this to our previously known coordinates to get our current coordinate.

From our testing, this form of motor encoder odometry is quite accurate and drifts very little if we stay under 280 rpm, but the turning odometry was quite inaccurate so we added a gyroscope. We then use a kalman filter which uses a prediction model to estimate probable values then will compare it with sensor values (encoder and gyro) will output the most likely value. We also use a GPS to start a match so the robot knows its initial position, we like the GPS as it is absolute positioning but has a very low refresh rate and is inaccurate while moving, how it works is it has a camera that sees the GPS strips then triangulates its own position.

The first way to use this odometry is to turn to a point with a angular PID then drive forward with a forward PID. This is what we used last year and it is very good. It is very accurate and does not drift which helped us a lot with skills in the use because their elevated field sits in the center which ruined many autonomies but it's worked well. For the amount we wanted to turn we used the virtual skills algorithm Alex made (max score last year) where it calculates theta with arctan of y then drives the amount calculated with pythagorean theorem. One issue is every time we turn we have to stop which slows the robot a lot.



```

// Turn to
deltaX = targetX - currentX
deltaY = targetY - currentY
targetTheta = M_PI - atan2(deltaX, deltaY)
angularError = std::fmodf((currentTheta - targetTheta) * 2 * M_PI, 2 * M_PI)

// Move to
deltaX = targetX - currentX
deltaY = targetY - currentY
linearError = std::hypot(deltaX, deltaY).

```

Move to is very simple as it is just pythagorean theorem but turn to took a while to derive, originally we used the function arc tangent. (last year) but this took a lot of lines. The issue with atan is that it's range is  $-\frac{\pi}{2} < \theta < \frac{\pi}{2}$ , which is not the full  $2\pi$ , this is due to it not recognizing opposing heading for example, since the inputs  $\frac{\pi}{2}$  and  $\frac{5\pi}{2}$  are the same. We had many proceeding if statements to check for quadrant handling. we found while reading documentation during fall break was atan2. This function can input  $-\pi < \theta \leq \pi$  as there are two inputs instead of one so it can tell between different quadrants.

One issue with turning and driving separately is we have to come to a complete stop everytime we switch directions so we then implemented curves. We saw a desmos graph from Purdue SIEBots with a algorithm called a boomerang controller which we could not find anywhere else but it seemed like a simple algorithm so we collaborated with several other VRC to implement this. We helped implement and test the initial version. How it works is we drive towards the end

a calculated current point  $(x_3, y_3)$   
when we drive towards it,  
(using PID)

$$h = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

$$x_3 = x_2 - h \sin(\theta_2) \cdot \text{dead} + \text{turn\_curve\_radius}$$

$$y_3 = y_2 - h \cos(\theta_2) \cdot \text{dead}$$

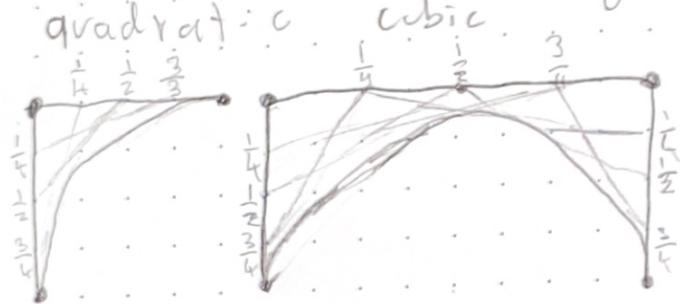
$$(1-t)(1-t)x_1 + tx_3 + t(1-t)x_2, (1-t)(1-t)y_1 + ty_3 + t(1-t)y_2$$

we recalculate the point as it will change when we turn to it and repeat. When repeated at 200ms it creates a very smooth curve. A quirk we found was that this algorithm leads the robot to follow a quadratic Bezier curve which is the same curve as Pure Pursuit. We have recently switched to Pure Pursuit for skills we feel the path is more customizable and it is also faster.

Right after the fulbright scrimmage we chose to switch to Pure Pursuit after much consideration. the benefit is that we can customize the path to however we want. Pure Pursuit is a algorithm proposed by Craig Courtevin 1992 and is commonly used by advance VRC teams. It takes a array with thousands of points which it drives following. It fixes errors using

a look ahead point which is calculated by drawing a circle with a radius of a constant called the look ahead dist. If set too low there will be lots of oscillation but if too high it will take while to get back on track. It will drive towards the intersection between the circle and the line. As mentioned we switched to this because of customizable paths and also minimal tuning. One of the cool things we designed was a web app to visualize and create the path then we can download it as a array of points then import it into the code. We make the path with a combination of lines, and cubic and quadratic bezier curves. We use two curve types because they have different advantages, the cubic bezier has two control points so it allows for more complex curves but takes a lot of time. The cubic bezier (similar to boomerang) is simple and fast.

The Bezier curve was created by Pierre Bézier  
the body work of Renault Cars and the spline  
commonly used for computer graphics and to  
as it has the ability to replicate any con-



the basic concept  
creating lines between  
the points to complete  
the curve.

We chose to use a boomerang controller for our  
autonomous because of PID's accuracy because  
we need to turn to tritballs accurately to intake  
while we will use Pure Pursuit for stunts as  
we are using wings and do not need accu-  
but want speed.

(continuation of pg 8) pid tuning

1 my 11, 12

blaster 3 continue pg 3

1 my 8 wing 9

wings continue pg 2

1 my 9 10